CHARACTERISTICS OF ROAD TRAFFIC ACCIDENTS IN KENYA

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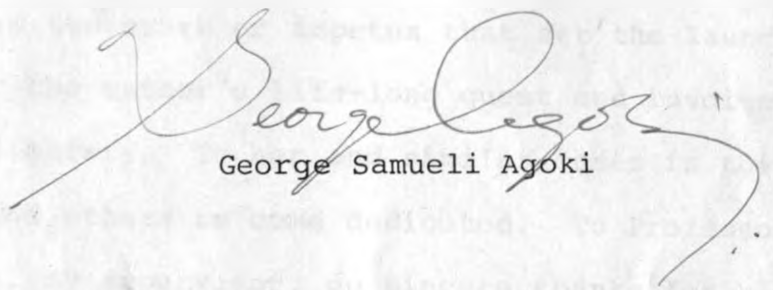
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A thesis submitted in fulfilment for the
degree of Doctor of Philosophy in the
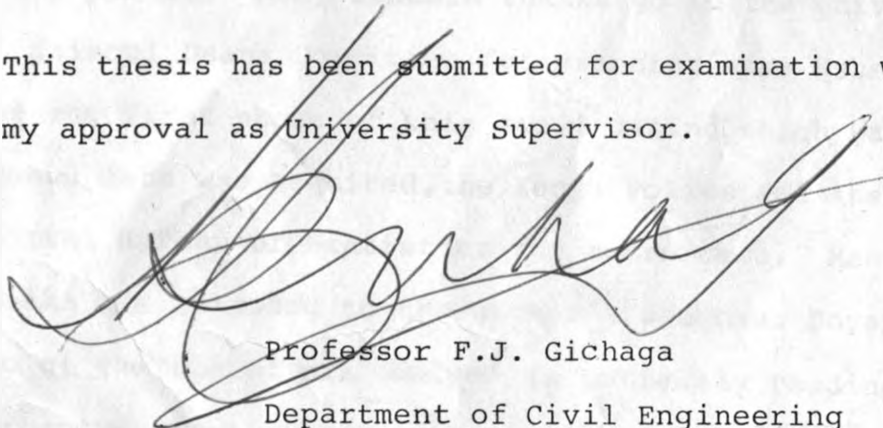University of Nairobi.

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DECLARATION

This thesis is my original work and has not been
presented for a degree in any other University.

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This thesis has been submitted for examination with
my approval as University Supervisor.

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Warm appreciation go to the author's father Johnson
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author's efforts, for their actions certainly helped
shape the events for the realization of this final
product.

SUMMARY

The objectives of this study were: to study Road Traffic Accidents (RTAs) in Kenya and to determine where possible their fundamental characteristics and causal factors related to their occurrence, to develop predictive models for Kenya at the national (macro) level to be used for the monitoring of RTAs and the performance of road safety improvement programmes and lastly to develop predictive models through some selected Kenyan roads at the road level (micro) to assist in the proper understanding of the behaviour of RTAs in relation to road design elements. In order to develop these predictive models, various mathematical models were used. These were: growth curve models, namely the logistic curve model and the logarithmic model; polynomial functions and finite differences techniques; harmonic analysis, generalised linear modelling and statistical methods for testing the fitness of the models developed. Macro level data for Kenya were collected from the Kenya Police records and from the Statistical Abstracts of the Central Bureau of Statistics of Kenya. Micro level data were collected through traffic volume counts, study of road geometry and pavement defects and by specially coded forms used for extracting data from the Traffic Police RTA records. The data were collected from the two carriageways of the Nairobi-Thika dual carriageway and the Kiganjo-Nanyuki single carriageway road. The police

forms were obtained from the police stations responsible
for the roads studied. The data were analysed to provide
characteristic patterns and evidence of the mathematical
techniques to be used in model development. Computer
facilities were used whenever necessary. The major
findings at the macro level were: the logistic model
is well suited in predicting the growth of RTAs and
related phenomena with time, the logarithmic trend
curve is well suited in predicting the growth in
the distribution of RTA responsibility and involvement
whilst the polynomial function is suited in predicting
the trend of RTAs in relation to motorization. The
major findings at the micro level were: the polynomial
functions are suited in predicting the effects of road
factors on RTA rates, the logistic curve is well suited
in predicting the growth of RTAs in relation to vehicle
flow, harmonic functions are suitable for predicting
variations in RTAs and vehicle flow with time of day
and the generalised linear model is beneficial when
trying to study the effects of traffic and geometrical
design elements on RTAs on an interactive basis. It
is recommended that there be continuous data collection
in the form of an accident data base. Such data will
then be used on a continuous basis for model calibra-
tion and monitoring of road safety improvement measures.

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CHAPTER 1 - INTRODUCTION

Right from the outset of Kenya's independence in 1963, major emphasis and efforts were directed towards the upgrading and expanding of the national road network in order to meet the transport demands of the aspired economic development targets, social and administrative requirements. Road development programmes took the form of reconstruction in order to improve the vertical and horizontal alignment standards and with all weather bitumenized or gravel surfacing simultaneously constructing new sections and extending these to reach areas of the country yet unaccessed.

Concomitant with the rapid transformation of the roads was a rapid increase in vehicle population and human population. With increased travel demand, more vehicles and better faster roads inevitably came the increased road traffic accident deaths, injuries and property damage.

The problem of road traffic accidents (RTAs) remains one of the major yet unsolved problems in Kenya in particular and in other countries of the world in general. It is a significant public health and engineering issue in general and traffic engineering specifically. This problem represents a serious

national loss as well as a loss to individual Kenyans
in terms of loss of life, injuries, loss of man-hours
and the consequent effects on overall efficiency.
RTAs are among the leading causes of death in Kenya
and in many parts of the world today. In Kenya, while
mortality and morbidity rates from transmissable
diseases are known to be declining due to better
medical and public health programmes, this is certainly
not so for RTAs mortality and morbidity rates. During
the period 1970-80 [58] statistics reported show
that the number of reported deaths from the major
notifiable infectious diseases decreased remarkably
while the death toll from RTAs rose. For example,
deaths from diarrhoeal disease, typhoid, smallpox
and malaria decreased from 4631, 2602, 948 and 731
to 182, 4, 0, 20 respectively while RTA deaths
increased from 944 to 1413. The carnage on Kenyan
roads   appears to be continuing into
the future unless and until some drastic measure is
taken. Whilst motor vehicles are an economic asset
to every nation and a necessity for mobility in
today's life pursuits they are simultaneously weapons
of wanton destruction of human lives.

The problem of RTAs is not very well understood
in many circles. The characteristics, the causes, the
interplay and counterplay of factors in RTAs causation
are least understood. Research in road safety in Kenya

will be one of the most significant contributions towards the solution of the RTAs problem in Kenya and elsewhere.

Research in road safety has been pursued by the use of empirical/basic research as well as trial-and-error methods. The trial-and-error method uses experimentation and observation to find out if a "promise-looking" device or idea does in fact increase road safety. The emperical/basic research method, more characteristic of scientific research, uses fundamental investigations which endeavour to build up an understanding of the phenomenon of RTAs. If an individual RTA is considered, the main questions asked are what was the cause and who or what was to blame. If these questions are asked concerning a large number of RTAs the answers could reveal factors which are significant in the causation of RTAs. But experience has shown that this is not entirely satis-factory for the purposes of building a picture of the most important causative factors. The objections include the fact that if one or two factors are named, this may be a matter of opinion in their choice, and many factors are inevitably left out because their relevance is questionable and impro-perly understood or assumed to be the normal state of affairs warranting no special consideration. For scientific purposes it is thought better to regard

the phenomenon of RTAs as a chance process. At any
time the road, the traffic, the user and the vehicle
in a set of circumstances have a certain probability
or chance of leading to a RTA. The statistician
thinks more of the probability and whatever affects
this probability rather than an individual RTA and
its causes. On the other hand, due to the fact that
every RTA depends on a serious disturbance in the
relationship between the road, the vehicles and the
road users, the disturbance is not a matter of chance
but traceable to an immediate causal factor. Thus,
by studying a large number of RTAs a better under-
standing of the characteristics and the nature of
occurrence of the disturbances giving rise to the
RTAs may be obtained. This may provide a better
practical approach from the point of view of the
improvement of the road safety situation. Methods of
defining RTA causes, so that relevant countermeasures
are found, are not very well established anywhere in
the world. No unified RTA theory exists. The basic
source of information so far has been various kinds
of in-depth investigations of RTAs. These investiga-
tions are extremely time - and money-consuming. Thus
in-depth RTA investigation studies are not justified
from a resource viewpoint.

For any given country there is necessity to
study the individual country's RTA characteristics

in order to understand the RTA problem better and
provide a means for developing and monitoring road
safety countermeasures. Such a research study may also
help in the contribution to the knowledge and theory
of RTAs and their causation.

The RTA problem may be summarized as the ever
increasing tendency of RTAs and their resultant effects,
lack of sufficient knowledge of RTAs characteristics
both at the national level (macro) and the road environ-
ment level (micro) as well as the lack of sufficient
knowledge of the characteristic effects of road users,
vehicles and the road. Due to the complex nature of
the RTA problem as well as their characteristics
it is necessary that the data collection be as compre-
hensive as possible so as to provide sufficient back-
ground for the understanding of RTAs within the entire
road traffic system. Consequently, the objectives for
this study become:

-      to study RTAs in Kenya in order to systematically
       determine their characteristic patterns,

-      to develop predictive models for RTAs in
       Kenya at the national as well as the road
       level.

In this way a framework for understanding the RTAs
problem in Kenya will be provided which will lead to
the development of realistic solutions aimed at
increasing road safety.

CHAPTER 2 - LITERATURE REVIEW

2.1 Macro Level

Prior to 1972 very little research had been carried out in the developing countries on the problem of RTAs. A reason why RTA studies have been neglected in these countries is given by Jacobs [1] as being that RTA rates have been considered to be low in comparison with countries in Europe and North America.

Using data for road fatalities, vehicles and population for the year 1938 from 20 countries, the majority of which were European, Smeed [1] derived a relationship which is given by the formula

$$F/V = 0.0003 \ (V/P)^{\frac{2}{3}} \qquad\qquad (2.1)$$

where, F = road fatalities

V = number of vehicles

P = population.

The equation was a good fit for data also collected by Smeed for 16 countries, mostly European, ranging over the period 1957-66 and also a good fit for data from 68 countries over the period 1960-67. Smeed [2], in 1949, had developed the formula

$$D = 0.0003 \ (NP^2)^{\frac{1}{3}} \qquad\qquad (2.2)$$

where, D = number of deaths in road accidents

in any year

N = number of licensed vehicles

P = population.

The statistics N, P and D had been reported by a number of Western European and North American countries. Another form of the Smeed [3] formula that has been used is

$$D/P = 0.0003 (V/P)^{\frac{1}{3}}$$ (2.3)

with the notations as above.

Using methods similar to those of Smeed, Jacobs and Hutchinson [1] carried out an analysis for 32 developing countries for which 1968 figures were available. The number of vehicles per 10,000 persons and the number of fatalities per 10,000 vehicles were calculated and compared. In order to derive a linear relationship the logarithmic values of the fatality rates were regressed against the logarithmic values of the vehicle-ownership rates and the following equations developed:

$$(F/P) = 0.00077 (V/P)^{\frac{3}{5}}$$ (2.4)

$$(F/V) = 0.00077 (V/P)^{-\frac{2}{5}}$$ (2.5)

with notations as above.

A number of countries over the 10-year period 1958-68
or as near to this period as possible had been chosen.
For comparison a number of developed countries were
included. , These relationships were found to be
statistically significant at the 1.0 per cent level.
The number of fatalities per head of population
increased with increase in motorization in all except
Cyprus where there was a slight decrease. The injuries
per head of population also rose with the rise in motori-
zation in all countries.

The change in fatality and injury rates per
licensed vehicle is thought to be a more meaningful
indication of the accident situation in any country
over a period of time. It has been observed that
there is a general tendency for both rates to decrease
with time. Out of the 29 countries studied, 15 showed
a decrease in fatalities per vehicle and 14 a decrease
in injuries per vehicle. This analysis agreed with
that carried out by Smeed [1] which showed decreases
in 15 out of 16 countries studied. Jacobs and
Hutchinson [1] state that in those countries not showing
this tendency unusual factors are operating. Kenya,
Zambia and Jamaica had considerable increases in the
number of fatalities per licensed vehicle. The reasons
for this were advanced as non-introduction of training
enforcement/regulations, improvement of vehicle standards

and road design whilst vehicle ownership was
increasing rapidly.

Various reasons for the fatality rate decreases
observed have been advanced the most important being:

-       a decrease in the proportion of two wheeled
        traffic on the roads, a category of vehicle
        with much the highest accident rate,

-       a general fall in pedestrian casualty rates
        which may in turn be due to improved pedestrian
        facilities,

-       an overall tendency towards higher levels
        of road-user education and training, better
        maintenance of vehicles and the road system.

Jacobs and Hutchinson [1] conclude that since the
Smeed equation predict a greater decrease in fatality
rate for an equivalent increase in vehicle ownership
than does the equation derived for developing countries,
it is possible that improvements in the safety of the
road system, the vehicle and the road-user are not
taking place as rapidly in the developing countries
as in the more developed. If this continues, the
accident situation is 'likely to become very serious
indeed in the developing world particularly in
situations of rapid motorization.

2.2 Micro Level

Data obtained on rural roads in Kenya and Jamaica were analysed separately by Jacobs [4]. From the analysis equations were derived which related RTAs per kilometre per annum to vehicle flow and RTAs per million vehicle-kilometres to the geometric parameters.

Regression analysis was used to establish and quantify the relationships between a dependent variable and one or more independent variables. The quantity under study was termed the dependent variable. The choice of the independent variables was such that they were 'sensibly' related to the dependent variable, simple to define and reasonably easy to measure for an engineer working in the field. As a first step investigation of which variables were most closely correlated with accident rate, simple regressions of accident rate on each of the road features individually, were performed. The equations derived were of the form

$$y = a + b_1 x_1 \qquad\qquad (2.6)$$

where y and x_1 are the dependent and independent variables respectively and a and b_1 the regression constant and coefficient respectively. Since many of the road design features are inter-related simple regression was thought to give a

misleading impression of the relationships that they
have with accident rate. Multiple regression, in
which the accident rate is expressed as a function
of several 'independent' variables simultaneously,
was thought to be a better guide. The equations
developed were then of the form

$$y = a + b_1x_1 + b_2x_2 + b_3x_3 + \ldots + b_nx_n \quad (2.7)$$

where, $y, x_1, x_2, x_3, \ldots x_n, b_1b_2 \ldots, b_n$ were
as above.

For these estimates to be acceptable it was necessary
to test the hypothesis that the value computed for
each regression coefficient was unlikely to have
arisen by chance. To check that this was true, the
standard error of each regression coefficient was
computed and tested for significance, variables
with non-significant coefficients being eliminated
from the analysis. The computer programme used
employed the technique called 'stepwise'regression
analysis whereby non-significant variables are
eliminated and tested with other combinations
replacing them where necessary. The relationships
significance levels used were 5 or 10 per cent
where 5 per cent probability is the level usually
accepted in statistical analysis. Due to the many
factors affecting accident rates, a relationship
found significant at 10 per cent level was also

considered satisfactory. The correlation coefficient r was given as well as the coefficient of determination r^2, where r^2 provides a measure of the proportion of variability in y that is accounted for by variability in the appropriate x value.

The number of injury RTAs per kilometre of road per annum occurring on rural roads in Kenya was regressed against the vehicle flow per hour occurring, on each test road section, averaged over a 12-hour period (7 am - 7 pm). The accident rate was found to be related to the vehicle flow at a significance level of 5 per cent. The equation derived for Kenya was

$$y = 0.116 + 0.009x \qquad (2.8)$$

where y = personal injury RTAs per km per annum

x = average vehicle flow/hour.

The regression equation of factors related to the accident rate, significant at the 5 per cent level, in Kenya was found to be

$$y = 1.45 + 1.02x_5 + 0.017x_3 \qquad (2.9)$$

where, y = accident rate per million vehicle-kilometres

x_3 = horizontal curvature (deg/km)

x_5 = junctions per kilometre

At the 10 per cent level of significance the
equation for Kenya was

$$y = 1.09 + 0.031x_3 + 0.62x_5 + 0.0003x_4 + 0.062x_2$$

$$(2.10)$$

where y, x_3, x_5 were as above and

$$x_4 = \text{surface irregularity (mm/km)}$$

$$x_2 = \text{vertical curvature (m/km)}$$

Thus in Kenya junctions per kilometre was found
to be the most significant independent variable with
$r^2 = 0.49$. The road studied was the Nairobi-Mombasa,
a two-lane single carriageway trunk road. There were
never more than two junctions per kilometre. An
addition of one junction per kilometre was associated
with an increase in the accident rate of over one
accident per million vehicle-kilometres.

In Kenya the horizontal curvature was found to
be significantly related to the accident rate with a
decrease of 35 degrees per kilometre reducing the
accident rate by one accident per million vehicle-
kilometres.

The effects of surface irregularity and vertical
curvature were considerably less than those of junctions
per kilometre and horizontal curvature. It was found

that the rougher the road the higher the number of
accidents per million vehicle-kilometres. An
improvement in roughness of 200 mm per kilometre was
associated with a reduction in the accident rate of
0.8 accidents per million vehicle-kilometres per
annum.

On the Nairobi-Mombasa road, there was very
little variation in the road width and the small
amount of variation did not provide a significant
relationship with accident rate.

In 1973 Silyanov [4] published the results of
a comparison of accident rates on roads of different
countries using data from Russia, Sweden, USA,
Australia, England, Hungary, West Germany, Czecho-
slovakia, France, Japan and Norway. The data used
was for personal injury RTAs.

It was found that the number of RTAs per
kilometre of road per year increases with an increase
in the hourly traffic flow with the relationship
given by the equation

$$n_N = 0.256 + 0.000408N + 1.36 \times 10^{-7} N^2 \qquad (2.11)$$

for $40 < N < 1600$

where, n_N = number of RTAs per km of road per year
on two-lane roads.

N = traffic flow (vehicles/hour).

The rise in the number of RTAs per vehicle-kilometer becomes markedly sharp when the width is less than 7 metres. The relationship was described by the formula

$$n_B = 1/(0.173B - 0.21) \qquad (2.12)$$

where, n_B = number of RTAs per million vehicle-kilo-metres

B = carriageway width (metres) assumed between 4 and 9 metres.

On the effect of the radius of horizontal curves the most dangerous curves were found to be those with radii less than 500 metres. The relationship was described as

$$n_R = 0.647 + 723/R - 649.5/R^2 \qquad (2.13)$$

where, n_R = number of RTAs per million vehicle-kilo-metres

R = the radius of horizontal curves in metres.

The longitudinal grade was found to greatly affect the RTA rate. The most dangerous effect of

the grade was found to be apparent for vehicle move-
ment down the grade. Silyanov [4] for the Russian
data found that 65 per cent of all the vehicles
involved on hills were moving downwards. The
number of RTAs was found to increase continuously
with an increase in grade, the increase being parti-
cularly sharp on grades of more than 3 per cent. The
formula developed was

$$n_i = 0.265 + 0.105i + 0.0229i^2 \qquad (2.14)$$

for $0.5 < n_i < 7$

where, n_i = number of RTAs per million vehicle-
kilometres

i = longitudinal grade expressed as percentage.

On sight distance many of the RTAs were found
to occur on road sections where sight distance is less
than 300 metres. The formula derived for the relation-
ship was

$$n_d = 1/(0.200 + 0.00111d + 0.0000009d^2) \qquad (2.15)$$

for $25 < d < 800$

where, n_d = number of RTAs per million vehicle-
kilometres

d = sight distance in metres.

Silyanov concludes that the practical application of the above relationships is for the detection of RTA black spots. Jacobs applied these relationships to study RTA rates in Kenya and Jamaica. His main finding was that RTA rates in developing countries, for similar levels of vehicle flow and geometric design, are considerably greater than in developed countries. It is noted however that some of the factors were not significantly related to RTA. Results from different countries also tend to contradict each other. No functional relationships between RTAs and road design elements have been developed for Kenyan roads to show how RTA rates vary characteristically with various design elements, pavement conditions and varying traffic conditions.

2.3 Road Traffic Accidents Characteristics

2.3.1 Road Traffic Accident Trends

Two deaths were recorded, in 1896 in Great Britain, as due to motor vehicles and one was recorded in the United States in 1899. From these small beginnings a terrible stream of road deaths and injuries has followed. Countries which have become highly industrialized and therefore motorized have suffered similarly to the extent that road accidents are the commonest cause of death in adolescents and young people, particularly males[20].

In most Western countries RTAs constitute the
commonest single cause of violent death. For example,
in Great Britain in 1963, after road deaths, death
by suicide was the next highest cause. Similar
trends were observed in Australia. In Canada road
death rates showed an equally formidable increase
over the period 1944-66 [17].

Smeed [17] has made the following observations
in the general trends in RTAs:

- a general tendency for road deaths per registered
  motor vehicle to decrease as motorization
  increases. Yerrel [41] cautions that it
  should not be assumed that as the number of
  vehicles per person increases in any given
  country, that such a country will simply
  and automatically follow a declining path
  of deaths per vehicle as if obeying some
  absolute law of nature.

- a general tendency for road deaths per head
  of population to increase as motorization
  increases.

- despite the very large differences in traffic
  conditions in different countries, the number
  of road deaths in a given country can to a
  large extent, be predicted from a knowledge
  of the population and number of vehicles
  [1, 2, 3].

- the number of pedestrian fatalities in a
  given country is largely determined by its
  population and is not very dependent on its
  degree of motorization.

- as motorization increases, there is a tendency
  for injuries to occupants of motor vehicles to
  increase in number relative to injuries to
  pedestrians.

- a general tendency for road-user behaviour
  to improve as motorization increases.

An analysis of personal injury RTAs in Great
Britain [3] showed the following RTAs trends:

- a steady increase of the total number of
  casualties for the period 1939-1960s.

- motorcyclists experiencing the greatest
  number of casualties, of the different classes
  of road users including children and adults.

- old people having the highest death rate.

- a very large number of casualties to pedal
  cyclists, motorcyclists and drivers on
  week-days during the hour 5-6 p.m.

- casualties to child pedestrians being more
  numerous in summer than in winter while
  adult pedestrian casualties were found
  to be relatively more frequent in winter.

- accident rates by night generally exceeding those by day.

- the total number of casualties tending to be greater in wet weather than in dry weather in winter.

- lowest RTA rates occurring on motorways and rural roads in open country and the highest in the centres of large towns. Death rates were found to be low in towns but high on unrestricted roads leading into large towns.

- of the different classes of vehicles, the motorcycle, per kilometre ridden, was found to be the most dangerous from the viewpoint of risk to the driver. The pedal cycle was found to be the next most dangerous. Also from the viewpoint of injury to the pedestrian, the motorcycle, per kilometre ridden, was found to be the most dangerous.

- risk of death, relative to the risk of injury was found to be the greatest for pedestrians.

In the Federal Republic of Germany, Froboese [30] gives the following RTAs trends in the period 1970-82:

- the number of casualties decreased slightly, that of deaths decreased considerably while the number of RTAs did not go down.

- despite the great increases in traffic volumes, the risk of becoming involved in a RTA decreased.

- motorways were found to be the safest roads for motorized traffic.

- the risk, for the occupants of passenger cars, of being injured decreased considerably.

- the number of RTAs involving pedestrians diminished.

- the number of RTAs involving cyclists increased noticeably.

- the RTA risk of the drivers of two-wheeled power-driven vehicles reached an alarming level.

- the number of children involved in RTAs, in particular the number of those killed greatly decreased. As pedestrians they were found to be less endangered, but as cyclists their risk increased.

- Beginner drivers were found to be especially exposed to danger and of particular danger to other road users.

- RTAs outside built-up areas were found to have particularly consequential effects, but the rate of RTAs within built-up areas was found to be higher.

-    RTAs involving only one vehicle and collisions
     with unprotected road users were found to be
     with the most serious consequences.

-    excessive speed and driving under the
     influence of alcohol were found to be
     the main causes of serious RTAs.

Sloth and Bach [28] have observed that the
number of RTAs and the number of fatalities have
followed a similar pattern in most industrialized
countries, until they reached unacceptable levels
during the late 1960s. For the developing countries,
Yerrell [41] has observed that death rates are very
often 20 times greater than those of Western Europe
or North America. For the period 1978-1980, for 35
developing countries Yerrell found a negative corre-
lation between fatalities per vehicle and the number
of vehicles per head of population, showing that the
smaller the number of vehicles relative to the
population the worse the death rate relative to those
vehicles.

2.3.2 Road User Characteristics

2.3.2.1 Driver

The driver's part in RTAs is a question of the
adequacy of his response to the road environment [3].
Driver characteristics result from the influence of

psychological and physiological characteristics on
the performance of the driving task and the interaction
of the driver with other road users and the road
environment. Drivers, like other road users, are both
recipients and causers of RTAs [20]. Various studies
concerning driver characteristics are summarised below.

In Great Britain in 1959 [20], drivers of cars,
commercial and passenger vehicles comprised 11.2 per
cent out of a total of all persons killed on the
roads. Using data from Belgium, Denmark, Great
Britain, Italy and Sweden [20] it was found that the
number of deaths of motor vehicle drivers is not
closely related to traffic density or to the number
of registered vehicles.

The age distribution of drivers killed showed
a peak in Great Britain below the age of 30 [20].
Serious injuries to car and taxi drivers in Great
Britain was found to be about 13 times, and slight
injuries about 44 times, the number of RTA deaths
amongst drivers. In the United States in 1959 [20]
drivers under the age of 25 were found to have a
considerably worse RTA ratio than that of all drivers.
The lowest fatal - RTA ratios were found to be for
those aged 50 to 60 (less than half those for drivers
under the age of 25). In a London transport study of
professional bus drivers for the period 1957-59 it

was found that [20] there was a relatively high RTA
rate in young and inexperienced drivers. Those under
30 with less than four years of service had nearly
four times as many RTAs as the best group, those that
were aged 60-64 with about 14 years of service. In
the United States it was found that driver rates in
fatal RTAs begin to rise at about the age of 65. In
Finland in 1958 [20] no correlation was found between
age and RTAs. Based on the number of RTAs per vehicle-
kilometres of travel age was found to affect RTAs only
in the over 65 age group in the United States [21].
Drivers under 25, representing 21 per cent of the
driving population in the United States, were found
to be involved in 34 per cent of the fatal RTAs.

In the United States [21] persons identified
as suffering from epilepsy, heart disease, diabetes
and mental illness were found to have a RTA rate
roughly twice that of the general public whereas
only 0.6 per cent of drivers fell into this category.
Drivers with physical defects in sight, hearing and
similar impairments were found to be involved in only
1.3 per cent of deaths and 0.6 per cent of all RTAs.
It was concluded that physical defects are not a
major contributor to RTAs.

Epidemiological studies [18, 20, 21, 26] show
that driving occur during periods when drivers are

under the active influence of drugs and alcohol. It
has been found that tranquilisers, barbiturates and
cannabis lead to impairment of driving skills [18].
Drinking drivers are one of the most serious causes
of all RTAs problems [21]. Physically and mentally
the drinking driver is RTA-susceptible. In the United
States the following alcohol related characteristics
were found [17]:

- drinking drivers responsible for fatal RTAs
  had higher Blood Alcohol Concentration (BAC)
  than those involved in non-fatal RTAs.

- as far as single vehicle RTAs were concerned,
  70 per cent of the drivers who died had been
  drinking beforehand. Further, in fatal single-
  vehicle RTAs, 49 per cent of the drivers were
  found to have BAC greater than 0.15 per cent
  whilst 20 per cent more had BAC between 0.05
  and 0.15 per cent.

- alcohol was a factor in about half of all RTA fata-
  lities.

- alcohol was related to RTAs in 30 to 70 per cent
  of instances.

- persons driving under the influence of alcohol,
  who were regarded as alcoholics, crashed their
  vehicles at higher speeds than did social
  drinkers.

- simple driving skills became impaired when
  BAC exceed 0.1 per cent.

- alcoholics were liable to incur six times as many RTAs and traffic violations as healthy drivers or drivers affected by medical illnesses uncomplicated by alcohol. In the United States 3 per cent of the drivers are alcoholics, and 4 per cent are "escape" drinkers [21]. Further, of the licensed drivers in the United States, all but 32 per cent drink [21].

- the presence of cirrhosis of the liver was found to be over 60 per cent in those persons who had substantial amounts of alcohol in their bodies at the time of death in RTAs.

- the ages of drinking drivers were found to be predominantly above 25 years.

Studies in Australia revealed the following characteristics concerning the effect of alcohol [17]:

- 40 per cent of drivers taken to hospital as a result of RTAs had BAC levels greater than 0.05 per cent.

- of the road fatalities 39.4 per cent were found to have BAC levels greater than 0.1 per cent.

- 60 per cent of the drivers killed in single vehicle RTAs had BAC levels greater than 0.1 per cent.

- drinking drivers tended to be older, mostly
   male and came from the lower occupation groups.

Similar studies in Great Britain [17] showed the
following characteristics about drinking and driving:

- of the drivers involved in RTAs 41 per cent had
   been drinking with 34 per cent having BAC levels
   greater than 0.05 per cent.

- drivers who had been drinking were most
   frequently involved in single vehicle RTAs.

- of the dead drivers 19 per cent had BAC
   greater than 0.1 per cent.

- the 30-39 age group had the highest BAC found
   to be greater than 0.15 per cent.

- 50 per cent of the drivers involved in RTAs
   between the hours of 10 p.m. and 4 a.m. had
   been drinking.

- the highest RTA rates were found to occur at
   week-ends after 10 p.m. when licensed alcohol
   selling premises closed.

In Canada the studies [17] showed that 28 per cent of
all drivers convicted of drunken driving were alcoholics
who had repeated RTAs and were less concerned with
careful driving after moderate or heavy intake than
were those alcoholics who managed to stay out of trouble.
The RTA repeater alcoholic believed that liquor had

no effect on his competence to drive. In Romania
[17] it was found that 28 per cent of the drivers
suspected of driving under the influence of alcohol
had been involved in RTAs. In Cechoslovakia [17]
persons with BAC greater than 0.15 per cent were
found to have a 124-fold greater risk of being
involved in RTAs when compared with those with
lower alcohol levels in their blood. In Finland
[17] some 14.6 per cent of road deaths were found
to be due to alcohol whereas in Poland the proportion
was found to be 15-19 per cent. On the whole most
studies [17] show that in the vast majority of cases
involving alcohol the subject is male and more
commonly in the 30-60 age group. The peak age
period for road deaths among drivers lies between
15 and 25. Alcoholism, particularly when compli-
cated by mental and physical illnesses, is an
affliction of the middle-aged, a clinical factor
of universal occurrence.

2.3.2.2. Pedestrian

The pedestrian as a factor carries much of the
responsibility for RTAs mainly for his own safety.
The main findings with respect to pedestrian
characteristics are as follows:

- in Great Britain RTAs to pedestrians account
  for about 40 per cent of the fatal RTAs and
  about 20 per cent in the United States and
  mostly occurring in urban areas [20].

- the age distribution of fatally injured
  pedestrians is uneven. From the walking age
  to age 10 and from age 65 upward pedestrians
  are at special risk [20]. In Great Britain [3]
  comparing the number of pedestrian casualties
  with the population in each age group the
  maximum risk occurred for the 5-9 year old
  with risk increasing with age for those over 40.
  Children under 10 were found to be likely
  victims of the light commercial vehicles while
  persons over 70 years of age were more frequently
  involved with motorcycles and pedal cycles as
  compared with other age groups.

- the total number of pedestrians injured is
  about 25 times the number killed. Pedestrian
  deaths increase at periods of peak travelling,
  in cities particularly, during working days [20].

- in the United States pedestrians fatally injured
  consisted of the elderly who had been drinking
  alcohol a little or not at all and a group of
  the middle-aged who had been drinking heavily
  [20]. Further studies in the United States
  showed that alcohol was causally involved in

- more than 30 per cent of all fatal pedestrian RTAs [18].

Further findings in Great Britain [3] indicate that:

- the number of pedestrian casualties increases at a lower rate than the traffic flow.

- about 67 per cent of pedestrian RTAs the pedestrian is likely to be crossing the road.

- the relative frequencies with which different types of vehicle collide with pedestrians varies with the crossing place.

- the proportion of pedestrian casualties whose injuries were due to being hit by motorcyclists was higher on uncontrolled crossings than elsewhere.

2.3.2.3 Pedal Cyclist

The pedal cyclist like the pedestrian is unprotected unless cycling on cycle tracks. Findings from Great Britain [20] show that pedal cyclists killed annually form about 11 per cent of the total RTA deaths. Those aged 7-15 and the elderly form the higher proportion of pedal cyclist deaths. For each cyclist death there are 75 injuries. The proportion of those killed to the injured pedal cyclists was observed to rise with increasing age.

2.3.2.4   Motorcyclist

The motorcyclist like the pedal cyclist is also
unprotected.   Moreover, the motorcycle is capable of
very high speeds implying greater risk and severe
injury.   In the United States [20] out of RTA deaths
17.3 per cent have been observed to be motorcyclists.
Motorcycle fatalities affect the younger age groups
heavily.   In Great Britain [20] about 70 per cent of
motorcycle deaths affected the age-group 18-40, the
majority being male.   Thus length of experience and
power of motorcycles were found to be the two most
important factors in RTAs to young motorcyclists.

In the United States studies [20] have shown
that about 1 per cent of registered vehicles are
motorcycles which are responsible for 1.3 per cent
of the fatal RTAs, showing that the degree of risk
to motorcycle riders increases.   Studies in Great
Britain [21] indicate that the death rate per
kilometre for motorcycles was over 20 times that
for motor vehicles while the personal injury rate
is 3 times as great.   In the United States the
corresponding figure was found to be 4:1.

2.3.3   Vehicle Characteristics

The proportion of RTAs in which defects in
vehicles are thought to be the primary contributor
has been shown to be relatively small [20,21,50].

In Great Britain one study [20] showed that only 2.5 per cent of casualties in RTAs were attributable to brake, tyre and steering defects. In the United States [21] in fatal RTAs 2-11 per cent were traced to vehicle defects. In one study in the United States [20] one out of five passenger cars and one out of four trucks inspected were found to require maintenance. Some striking differences in the statistics of vehicle defects in RTAs have been observed. The common belief has been that vehicle inspection is carried out in order to decrease the risk of   RTAs through badly maintained vehicles. However, studies have not been conclusive in directly linking poorly maintained vehicles to RTAs [21].

2.3.4   Road Characteristics

Various studies in different countries have shown a strong correlation between RTAs and road design, construction and surfacing. Studies carried out in the United States [55] showed that simply resurfacing and/or widening of substandard roads increases speeds and hence increases the number and severity of RTAs. It was found out that improvements in highway geometry and alignment should be under-taken as well. Road features such as right curves, left curves, upgrades and downgrades were found to have equal traffic exposure. Left curves had a greater

RTA rate than right curves. The explanation was
that there is a tendency for vehicles to depart
to the right side of the road due to the fact that
if a vehicle leaves the travel lane to the left, the
adjacent lane allows room for recovery. There was
also found a higher RTA rate on curves than on
tangent sections. Most of the single vehicle RTAs
involved departure on the outside of the curve rather
than the inside. The explanation was that a moving
object continues along a straight path unless
redirected and hence the vehicles tend not to turn
enough rather than turn too much. In these studies
it was further found out that amany factors, often
independent, influence RTAs. It was difficult to find
sites where all the important factors are similar for
generalization to be made. It was therefore proposed
that in future studies, other types of analyses
involving operational measurements and theoretical
simulations be used to determine the relationship
between highway design elements and road safety.

Additional research in the United States [12]
on highway features and in particular on downgrades
showed that downgrades were overrepresented as RTA
sites. The RTA rate for downgrades was found to be 63
per cent higher than for upgrades while left curves
were found to be most RTA prone. Injury rates were

higher on curves than on tangent sections, higher on downgrades than upgrades and lowest for level straight sections. There were also results which appeared to be contrary to the view that improved conditions are safer. The possible explanation for this was that the better the road, the more careless the driver. However, experience with substandard roads has shown that high design standards have led to improved transportation and road safety.

Studies in the United States [11] to develop and test RTA prediction techniques showed that highways in mountain terrain cause problems especially for large commercial vehicles. The data collected included horizontal curvature, vertical alignment, percentage grade and length of grade. The problems were explained to be poor manouvrability, poor performance on upgrades and braking on descending grades due to loss of vehicle control, loss of brakes and improper downshifting of gears. The results of the analysis showed that:

- there was an increase in the RTA rate as the slope increases.
- a series of curves of decreasing radius place greater demands on vehicle brakes than of increasing radius.

- the average slope of a downgrade is not as
  significant as horizontal curvature at specific
  subsections of the grade.
- most RTA prone locations occurred just downgrade
  from sections of increasing horizontal curvature.
- irregular curves with frequent discontinuities
  were more RTA prone than smooth curves with
  few discontinuities.

However, the study was found to be limited in that
there was poor agreement in some results due to the
small sample size used.

Tests on road safety in Great Britain [10] on
a motorway dual carriageway showed a marked relation-
ship between RTA rate and alignment. The results
showed that:

- there was a difference in RTA rate for the two
  carriageways. There was no direct evidence to
  suggest the reason for this but the possible
  reasons suggested were driver fatigue, effect
  of sunglare, differences in skidding resistance
  due to different surfaces and road gradients.
- relating RTA rate to the curvature and gradient
  of the carriageway, evidence was found of higher
  RTA rates on gradients on curved sections than on
  straight sections.

- the effect of travelling downgrade was found
  to cause a lot of RTAs. This effect was found
  to increase stopping distances in three ways:
  the vehicle achieves greater speed, a small
  component of the vehicle opposes braking and
  the friction between the tyre and the road
  is less

Studies in Denmark [29] on RTAs on different
types of roads and intersections as a function of
the traffic flow showed that:

- the lowest number of RTAs is still to be found
  on motorways.
- for a given increase in the traffic volume, the
  increase in the RTA load will be less. There-
  fore, it pays off to gather the traffic on
  a few safe roads instead of spreading it on
  a large road network.
- the RTA frequency is lower on road sections
  lined with marginal strips, cyclepaths and
  the like compared to road sections without
  marginal stripping and paths. Furthermore, it
  was concluded that a broad delimitation will
  yield a lower RTA rate on road sections up to
  7 metres compared to narrow types of delimita-
  tion.

- two-lane urban roads of 6-9 metres have about
  twice the number of RTAs of rural road sections
  per kilometre.

- the number of RTAs per kilometre decreases on
  two-lane rural roads when the road width is
  increased up to 7-8 metres as a maximum. There-
  fore, a further reduction in the number of RTAs
  cannot be expected by widening two-lane roads
  with a width of carriageway of 7-8 metres.

- at two-lane roads with ribbon development the
  number of RTAs per kilometre increases with
  increasing width of carriageway up to 9 metres.

- signalized four-armed intersections have the
  highest number of RTAs. In general, the three-
  armed intersections have lower RTA numbers than
  the four-armed intersections.

- the black-spot calculations pointed out 11 per
  cent of the total section length and 5 per
  cent of the intersections as black-spots.

2.3.5 Road Traffic Accidents Characteristics in Kenya From Previous Studies

An analysis of RTAs in Kenya for the year 1972
[53] showed that:

- the greatest number of RTAs occurred in Nairobi
  (40 per cent of the total). The number of

casualties was also greatest in Nairobi,
being 30 per cent of the total.

- the lowest RTA and casualty rates occurred in
Nairobi, however. These lower rates were
attributed to the fact that almost all the
vehicle-kilometres travelled took place in
built-up areas where vehicle speeds on average
are much lower than elsewhere in Kenya.

- in Nairobi almost 50 per cent of all RTAs
involved pedestrians.

- Central, Eastern and Western Provinces had a
large majority of casualties occurring in
rural areas on roads without 30 or 45 kilometres
per hour speed limits.

- in urban areas 14 per cent of all RTA casualties
were fatal and 28 per cent were serious. The
equivalent figures for Great Britain in 1971,
were 2.2 and 25.4 per cent. These differences
were found to be statistically significant
at the 5 per cent level.

- in rural areas over 16 per cent of all casualties
were fatal. This level of severity was found
to be consistent with observations in other
countries including Great Britain. It was
attributed to the fact that RTAs occur at higher
speeds in rural areas and that medical treatment
is less readily available there.

- 38 per cent of all road casualties occurred to car occupants, which is low compared with most European countries.

- 26 per cent of all road casualties involved pedestrians, which is similar to the situation in Great Britain, Yugoslavia and Spain but rather high compared with Germany, France and Italy. However, the rate of pedestrian casualties per head of population was found to be low in Kenya compared with Great Britain.) Pedestrian casualties per head of population was slightly higher in Nairobi than the average value for urban dwellers in Great Britain.

- 16 per cent of the total casualties were occupants of commercial vehicles. This was found to be high compared to most European countries where the percentage is of the order of 5.

- the proportion of two-wheeled motor vehicle casualties was under 5 per cent of the total, being very low compared to European figures.

- 45 per cent of all fatalities were pedestrians.

- of the total casualties 11 per cent are juvenile casualties. The equivalent figure for Great Britain is 18 per cent and for Ghana 23 per cent. Over 19 per cent of all juvenile casualties were fatal whilst the equivalent adult value was 14

per cent, the difference being statistically
different at the 5 per cent level. In Great
Britain the reverse is the case with 2-3 per
cent of the adult casualties being fatal and 1.4
per cent of the juvenile casualties (under 15
years of age being fatal.

- over 12 per cent of all casualties occurring
during daylight were fatal, whilst during
darkness the equivalent figure was 17 per cent.

- 75 per cent of all casualties were injured
during the day and 22 per cent during darkness.
In Great Britain a greater percentage of all
casualties occurred in darkness (36 per cent),
a difference attributed to proportionately more
driving being done during darkness than in Kenya.

- the incidence of RTA casualties in Kenya rose
sharply between 6 a.m. and 7 a.m., continuing
throughout the day to reach a peak at 5 p.m.
The casualty rate then decreases sharply until
the following morning. In Great Britain for
casualties a peak is reached at 5 p.m. but
another peak occurred at midnight , attributed
to a greater proportion of travel at night and
a greater incidence of drinking and driving
in Great Britain than in Kenya.

- the highest number of casualties occurred on
Saturdays and Sundays whereas in Great Britain

the highest numbers occured on Fridays and
Saturdays. The difference was found to be
statistically significant at the 5 per cent
level and differences in social patterns were
thought to be the main reason for the difference.
the greatest number of casualties occurred in
March, April and September. In Great Britain,
Sweden, Germany and Yugoslavia the number of
casualties were highest in November and December.
In Great Britain periods of a high number of
casualties do not coincide with high traffic
flows. The casualties are affected by short
hours of daylight and extreme weather conditions.
No evidence of a regular pattern of seasonal
traffic flow variation had been observed in
Kenya. Therefore, there was no evidence to
support the notion that traffic flows are
greatest in the month containing the highest
number of casualties. In Kenya the climate
follows a strong seasonal pattern with long
rains in March, April and May and short rains
in November. The first period coincides with
months when casualties are highest.
vehicle-pedestrian RTAs were the most frequent
being over 39 per cent of the total. In Great
Britain the equivalent figure is 29 per cent,
the most common RTA being the vehicle-vehicle

RTA which accounts for 45 per cent of the total
whilst in Kenya it is 18.2 per cent. Single
vehicle RTAs in Kenya were 27.4 per cent of the
total for the equivalent figure of 19.5 per cent
in Great Britain. Nearly 75 per cent of all
single vehicle RTAs were found to occur in rural
areas in Kenya whereas the equivalent figure in
Great Britain was 42 per cent. The differences
were statistically significant at the 5 per
cent level.

on surfaced roads, vehicle-pedestrian RTAs were
commonest whereas on murram and unsurfaced
roads single vehicle RTAs were the most common.
Over 77 per cent of all RTAs occurred on surfaced
roads and 18 per cent on murram roads. With
5 per cent of the total vehicle kilometres
travelled in Kenya taking place on murram roads,
the RTA rate per vehicle-kilometre is much
higher on murram roads than on surfaced roads.

cars and land rovers were involved in 60 per
cent of all injury RTAs reported. Motorcycles
had the lowest rate, a pattern commonly
found in most European countries. In Great
Britain however, public service vehicles have a
higher RTA rate than private cars since public
service vehicle journeys are made in heavy
traffic with high occupancy rates.

An epidemilogical review of RTAs in Kenya for
the period 1968-73 [23] found that the main factors
were behavioural in nature and related to the drivers
who accounted for 48 per cent of all RTAs, the pede-
strians who accounted for 24.3 per cent, vehicle
defects which accounted for 5.3 per cent and road
defects which accounted for 2 per cent of the total.

In a review of RTAs in Nairobi for the period
1968-72 [24] it was found that:

- the severity index was 8.9 per cent.
- the fatality rate per 10,000 persons was
  3.44.
- seasonal variation of RTAs appeared to be
  related to the climatic seasons by month
  of the year, day of the week and time of day.
- children, students and civil servants were
  greatly involved in RTAs and their casualties
  resulted in 29 per cent of all RTA deaths.
- the fatality figure for the males was 5 times
  the figure recorded for the females.
- vehicle drivers were responsible for 48 per
  cent of all RTAs and the pedestrians were
  responsible for 36.2 per cent of all RTAs.
- 93.9 per cent of all RTAs were caused by road
  users.

In a further study on RTAs involving children under 12 years in Nairobi [19] during the year 1975 it was found that:

- RTAs ranked 14th among the top killers in all age groups of children and 2nd in the older more exposed school age group of 5-12 years.
- among the adults RTA deaths accounted for 10.3 epr cent of all deaths and for all ages RTA deaths accounted for 5.9 per cent.
- among RTA deaths there appears to be a male preponderance consistent with findings in the rest of the world.
- 60 per cent of the children involved in RTAs were unaccompanied at the time of the RTA, of whom 60 per cent were school children.
- for every RTA death there were 5 serious injuries and 8-19 slight injuries.
- RTA cases accounted for 7 per cent of all surgical admissions.

A study undertaken to establish the level of traffic laws violation in selected roads in Nairobi [25] indicated how unsafe it is to travel in Nairobi with 3-5.8 per cent of the drivers failing to conform to the traffic lights requirements, 2.8 per cent of the drivers failing to stop at a mandatory stop sign and 24 per cent of the drivers failing to keep to the proper lanes on a roundabout.

The pattern of RTAs in Kenya has been summarized [51] as follows:

- the RTA rate is stable, while the fatality rate is increasing,
- fatal RTAs constitute 15 per cent of the total,
- 60 per cent of all RTAs occur in rural areas,
- 47 per cent of all RTAs occur on rural tarmac roads,
- 45 per cent of all RTA deaths are pedestrians,
- 25 per cent of RTAs occur after dark,
- 12 per cent of RTA deaths involve vehicle occupants,
- 43 per cent of vehicle - pedestrian RTAs occur on tarmac roads, being 36 times the rate elsewhere,
- 40 per cent of RTAs on rural roads involve single vehicles.

An inventory of the factors and activities related to road safety in Kenya in the late 1970s [26] revealed:

- lack of adequate traffic enforcement, which is leading to unconcern about traffic rules and regulations,
- careless and incorrect driving, in which traffic rules are ignored, leading to too high speeds, careless overtaking and lack of care for safety in light traffic.

- passenger transportation on open platforms of
  commercial vehicles and extremely overloaded
  matatus,

- the absence of pedestrian and bicycle lanes or
  separated tracks next to main roads in densely
  populated areas,

- shortcomings in the pedestrian and bicycle
  facilities in urban areas, such as the insuffi-
  cient and too narrow footpaths, too few pede-
  strian crossings and non-existent two level
  street crossings,

- the bad condition of some main highways, parti-
  cularly of the pavement.

- use of improperly maintained and imperfectly
  equiped vehicles,

- driving under the influence of drink,

- lack of traffic education and information,
  leading to the ignorance of traffic rules and
  regulations.

2.4   Appraisal of Previous Research

The Ministry of Transport and Communications
has besides the above studies carried out work on the
development of an accident data system , a study of
dangerous (black spot) locations and investigations
on some selected RTAs. The accident data system
is based on police RTA records and is aimed at

providing statistical information for analysis and
dissemination. The study of dangerous locations on
roads under the ministry was aimed at the development
of countermeasures for highway improvement and
consequently road safety improvement. The accident
investigation was short-lived and failed to generate
further research. Other research on study areas
being pursued by the ministry include the review of
the traffic legislation in Kenya, monitoring of
alcohol amongst drivers, speed checks and the deve-
lopment of road safety devices. Findings from these
activities are yet to be studied more comprehensively
in order to yield concrete findings and conclusions
which will facilitate further research and improve-
ment in safety.

The Smeed relationships [1,3] for the developed
countries and the Jacobs and Hutchinson's [1]
relationships for the developing countries have been
used for international comparisons of accident sta-
tistics. These so-called Smeed relationships are
merely statistical relationships which do not imply
in any way causal relationships. Firstly, international
comparisons are made difficult due to the absence of
common definitions of death. Secondly, whilst the
countries of Western Europe are uniform in many ways
developing countries, particularly those of Africa,

have wide diversity of conditions and limitations.
In particular there is diversity in areal extent,
populations, gross national products, road provision
per head of population as well as motorization. There-
fore the Smeed, Jacobs and Hutchinson's relationships
cannot adequately predict fatalities for all countries
with the kind of accuracy required for monitoring road
safety programmes. Moreover fatalities are not the
only parameter that needs to be predicted in RTAs
characteristics.

The Silyanov relationships are useful for the
detection of RTA black spots for similar traffic
conditions. Thus they cannot predict RTA rates for
roads of such countries as Kenya which have vastly
different traffic and accident characteristics from
those studied by Silyanov. Results obtained by the
application of these relationships tend to be contra-
dictory.

As an improvement to the foregoing work it is
necessary to study the accident situation comprehensi-
vely using data available from as far back as records
exist to facilitate the development of RTA characteri-
stics and trends over long periods. These trends
can then form the basis for prediction of the accident
situation in those countries and the possible future
directions. Moreover, the study of an individual

country's statistics will aid in a better under-
standing of the accident situation before remedial
measures can be developed. Further, the relevant
predictive models of the accident trends can be a
powerful monitoring tool for road safety improvements.
Also, a systematic study of RTAs on a given road or
roads over a considerable period of time with varying
traffic conditions and geometric design can help
in the developement of functional relationships
between the various design elements and the accident
rates related to the changes in those elements.

2.5   Study objectives

The objectives of this study therefore are:

-       to develop predictive models for RTAs in
        Kenya at the national (macro) level to be used
        for monitoring the performance of road safety
        improvement programmes,

-       to develop predictive models for RTAs on
        Kenyan roads at the micro level to aid in the
        proper understanding of the behaviour of RTAs
        and the various design elements,

-       to study RTAs in Kenya with a view to determining
        the fundamental and long term characteristics
        and where possible the causes or reasons for
        these observed characteristics.

CHAPTER 3 - THEORETICAL ANALYSIS

3.1   Functions in Road Traffic Accident Theory

Road traffic accidents theory is an analytical
study that should be concerned with relations that
exist, or can be assumed to exist, between quantities
which are numerically measurable. Variable quantities
in RTAs are, among others, RTA numbers, population of
vehicles and humans, growth rates associated with
those, road width, road curvature, road gradients
and other factors related to the vehicles and roads.
Some of these quantities are measurable in physical
or 'natural' units. It is sufficient that they are
measurable in some units. Mathematical methods are
possible in RTA analysis and relations are expressible
by means of mathematical functions.

The relations of RTAs, and their related functions
which seek to express their form, are usually of
unspecified or unknown form. For example, a RTA
function cannot automatically be said to be linear
or quadratic in form, although it is sometimes con-
venient to assume that it can be approximately re-
presented in one of such ways. RTA conditions and
the very nature of their occurrence impose certain
limitations on the form of the functions. Fortunately
by considering the problems of RTAs it is possible to

say that the function concerned has the mathematical

property of being single-valued and decreasing, or

may be represented by a certain mathematically

shaped curve. This representation is sufficient for

the profitable use of mathematics and the understanding

of the variations in the measurable quantities in RTAs.

Analytical methods are therefore applicable to

RTA problems and their analysis. However, mathematics

is a tool for analysis and not a master! In this

chapter the theoretical basis, which forms the back-

ground for analysis in chapter 4, is set out. The

necessary formulae and the curve equations used for

curve fitting are presented. These formulae form the

basis for the development of the predictive models.

The statistical techniques and formulae for testing

the fitness of the model equations are stated.

Finally, the method of the final analysis is stated.

The method of the final analysis used compares

predicted data with observed data with the aim of

establishing how well the developed models predict

observed RTA data as well as the related observed

road traffic data.

3.1.1 General Theory of Functionals [8]

The theory of functionals deals with functions

of a finite number of variables. The variations of

quantities depend on one or more other quantities

as well as functions which take a quantity as
dependent not only as a finite set of other quanti-
ties but upon one or more variable functions.

A variable quantity u is defined as taking
its value from the form assumed by a function

$$x = \phi(t) \tag{3.1}$$

To each function $\phi(t)$ corresponds a definite value
of u.   As the form of the function is changed
so is the value of u changed.   The dependence of
u upon $\phi(t)$ is called a functional and written as

$$u = F\{\phi\} \tag{3.2}$$

A function of a function is expressed as

$$u = F\{\phi(t)\} \tag{3.3}$$

This assumes that $\phi(t)$ is a given function of t and
that u is also a definite function of t.   The
functional takes ϕ as a variable function associating
one value of u with each whole function ϕ.   The
variable t does not itself appear in the deter-
mination of u in the functional relation.   Generally,
when u is a functional of several variable functions
$x = \phi(t)$, $y = \psi(t)$, $z = x(t)$; ... the function takes
the form

$$u = F\{\phi, \psi, x, ...\} \qquad (3.4)$$

The functions ϕ, ψ, x, ... may be functions of several variables instead of one variable t only. In Fig.3.1, the variable function $x = \phi(t)$ is shown by a variable curve C in the plane Oxt. As the form of the function changes $(\phi_1, \phi_2, \phi_3, ...)$, the curve C shifts and takes up different positions and shapes $(C_1, C_2, C_3, ...)$. If u is a functional of ϕ, then its value depends on the particular position taken by the curve C, and to the series of positions $(C_1, C_2, C_3, ...)$ there corresponds a series of definite values of u $(u_1, u_2, u_3, ...)$.

FIG. 3.1 VARIATION OF FUNCTION AND CURVE SHAPE

3.1.2 Calculus of Variations [8]

Frequently the functional $u = F\{\phi\}$ takes
the form of an integral. If f(t) is a function
varying in form as (t) varies then

$$u = \int_{t_P}^{t_Q} f(t)\,dt \qquad (3.5)$$

where t_P and t_Q are certain limits of integration.
The value of u depends on what particular form is
taken for $\phi(t)$ and hence for f(t). The problem
usually encountered is the determination of that
function ϕ which makes u a maximum or a minimum.

The problem to be solved is stated so that
only certain arcs of the variable curve $x = \phi(t)$
in the plane Oxt are needed. Definite conditions
are imposed upon the points which mark the ends
of the arcs. The conditions are that the arcs
should start and finish at two fixed points P and Q
in the plane. The boundary conditions impose
limitations on the field of possible variation
of the curve $x = \phi(t)$ and only those curves which
can be drawn from P to Q are considered. Analytically,
the function $x = \phi(t)$ can only be selected
provided that $\phi(t_P) = x_P$ and $\phi(t_Q) = x_Q$
where $P(t_P,x_P)$ and $Q(t_Q,x_Q)$ are the two fixed end-
points.

Although the boundary conditions are considered,
the field of variation of the function x = φ(t) is so
large that the analysis of the problem is practically
impossible unless the field of variation is limited
by a further device. For, example, it is possible
to take only those functions satisfying the boundary
conditions, which are continuous and possess continuous
derivatives up to any desired order. Further still,
the field could be severely limited by taking
functions of a particular type of curves of a
particular class. For example, only those functions
of the quadratic form represented by parabolas with
vertical axis may be taken. The function type or
curve class can be represented by a relation
involving certain parameters α, β, γ, ... The
larger the number of parameters the more general
is the function type or class of curve. Consequently,
the limitation on the field of the variable functions
is to replace the function x = φ(t) of variable
form by

$$x = \phi(t; \alpha, \beta, \gamma, \ldots) \qquad (3.6)$$

where φ is now of fixed form and the variation
of the function is replaced by the variation of the
parameters involved. Thus the function is limited
to a more restricted variation described by para-
meters in a function of fixed form. If the parameters

are few the restriction is very severe. For example,
if only three parameters are used the function takes
the form

$$\phi(t) = \alpha t^2 + \beta t + \gamma \qquad (3.7)$$

so that the variable curve is limited to the class
of parabolas with their axes vertical and parallel
to 0x. But by taking more parameters the field of
variation of the function type is more generalized.
If a sufficiently large but finite number of parameters
is selected, the restricted field of variation is made
different in a few vital respects from the complete
field. This is achieved by excluding from the latter
the more unusual kinds of functions. The problem of
the calculus of variations is thus reduced to a
problem of extreme values of an ordinary function
of several variables ϕ, β, γ,... Functionals are
changed back into functions. The step from functionals
to functions is reversed. A functional can be regarded
as a function of an infinity of variables. Approxi-
mations are made by taking a function of a large
number of variables, the parameters α, β, γ,...
The extent to which the simplified form approximates
the original depends on the number of parameters
taken. The vital point about the analysis is that
it is quite independent of how many parameters there
are, provided their number is finite. The solution

obtained is not perfectly general but provides
an approximation sufficiently descriptive for
practical purposes.

The function f(t), which gives the variable
u on integration, depends upon the variable
function x = ϕ(t) and the derivatives of ϕ.
For example, using only the first derivative ϕ'(t)
and the function ϕ(t)

$$f(t,x, \frac{dx}{dt}) = f\{t, \phi(t), \phi'(t)\}. \qquad (3.8)$$

This is a function of t given in the function of
functions form. The function ϕ is of variable form,
as well as the function f to be integrated. It
is further assumed that the boundary conditions
are such that the variable curve x = ϕ(t) pass
through two fixed points P and Q. In particular
setting P and Q as 0 and 1 respectively
the problem reduces to one of finding extreme values
of the integral

$$u = \int_{t_o}^{t_1} f(t, x, \frac{dx}{dt}) dt \qquad (3.9)$$

for all possible variations in the function x = ϕ(t),

such that $\phi(t_o)x_o$, $\phi(t_1) = x_1$ where (t_o, x_o) and
(t_1, x_1) are the fixed points.

To solve equation (3.9) the limitations above are
imposed on the variation of $\phi(t)$ taking the functions
in the form of equation (3.6) where ϕ is a fixed
function with a continuous derivative and α, β, γ,...
are parameters. Alloting arbitrary differential
increments $\delta\alpha$, $\delta\beta$, $\delta\gamma$,... to the parameters the
corresponding variations x and x' are first derived
in the function x and its derivative x' = dx/dt.
Thus

$$\delta x = \frac{\partial x}{\partial \alpha} \delta\alpha + \frac{\partial x}{\partial \beta} \delta\beta + \frac{\partial x}{\partial \gamma} \delta\gamma + \quad \ldots \quad (3.10)$$

$$\text{and } \delta x' = \delta\left(\frac{dx}{dt}\right) = \frac{\partial}{\partial\alpha}\left(\frac{dx}{dt}\right)\delta\alpha + \frac{\partial}{\partial\beta}\left(\frac{dx}{dt}\right)\delta\beta + \frac{\partial}{\partial\gamma}\left(\frac{dx}{dt}\right)\delta\gamma +\ldots$$

$$= \frac{d}{dt}\left(\frac{\partial x}{\partial\alpha}\right)\delta\alpha + \frac{d}{dt}\left(\frac{\partial x}{\partial\beta}\right)\delta\beta + \frac{d}{dt}\left(\frac{\partial x}{\partial\gamma}\right)\delta\gamma + \ldots$$

$$= \frac{d}{dt}\left(\frac{\partial x}{\partial\alpha}\delta\alpha + \frac{\partial x}{\partial\beta}\delta\beta + \frac{\partial x}{\partial\gamma}\delta\gamma +\ldots\right)= \frac{d}{dt}(\delta x)$$

$$(3.11)$$

All the variations here are ordinary differentials and
subject to the ordinary rules of differentiation.

where, "d" refers to variation in the variable t and

"δ" refers to variation in the parameters $\alpha, \beta, \gamma, \ldots$

The function $f(t,x,x')$ and the integral u can be considered as dependent on the parameters α, β, γ,... and the variations in their values are obtained as

$$\delta f = \frac{\partial f}{\partial x}\,\delta x + \frac{\partial f}{\partial x'}\,\delta x' = \frac{\partial f}{\partial x}\,\delta x + \frac{\partial f}{\partial x'}\,\frac{d}{dt}(\delta x) \qquad (3.12)$$

and

$$\delta u = \delta\left\{\int_{t_o}^{t_1} f(t,x,x')\,dt\right\} = \int_{t_o}^{t_1}(\delta f)\,dt = \int_{t_o}^{t_1}\left\{\left(\frac{\partial f}{\partial x}\,\delta x\right)\right\}dt$$

$$+ \int_{t_o}^{t_1}\left\{\frac{\partial f}{\partial x'}\,\frac{d}{dt}(\delta x)\right\}dt$$

Now

$$\frac{d}{dt}\left[\frac{\partial f}{\partial x'}\,\delta x - \int\left\{\frac{d}{dt}\left(\frac{\partial f}{\partial x'}\right)\delta x\right\}dt\right] = \frac{d}{dt}\left(\frac{\partial f}{\partial x'}\,\delta x\right) - \frac{d}{dt}\left(\frac{\partial f}{\partial x'}\right)\delta x$$

$$= \frac{d}{dt}\left(\frac{\partial f}{\partial x'}\right)\delta x + \frac{\partial f}{\partial x'}\,\frac{d}{dt}(\delta x) - \frac{d}{dt}\left(\frac{\partial f}{\partial x'}\right)\,\delta x = \frac{\partial f}{\partial x'}\,\frac{d}{dt}(\delta x).$$

Hence, except for the addition of an arbitrary constant.

$$\int\left\{\frac{\partial f}{\partial x'}\,\frac{d}{dt}(\delta x)\right\}dt = \frac{\partial f}{\partial x'}\,\delta x - \int\left\{\frac{d}{dt}\left(\frac{\partial f}{\partial x'}\right)\,\delta x\right\}dt.$$

so $\displaystyle\int_{t_o}^{t_1} \{ \frac{\partial f}{\partial x'} \cdot \frac{d}{dt}(\delta x) \} dt \;=\; \left[\frac{\partial f}{\partial x'}, \delta x \right]_{t_o}^{t_1} - \int_{t_o}^{t_1} \{ \frac{d}{dt}(\frac{\partial f}{\partial x'},) \delta x \} \; dt.$

The expression for the variation in u then becomes

$$\delta u \;=\; \left[\frac{\partial f}{\partial x'}, \; \delta x \right]_{t_o}^{t_1} + \int_{t_o}^{t_1} \{ \frac{\partial f}{\partial x} - \frac{d}{dt}(\frac{\partial f}{\partial x'},) \} \delta x \; dt. \qquad (3.13)$$

The problem in the calculus of variations is therefore reduced to the simple problem of integrating a differential equation.

3.2. Model Curves

3.2.1 Logistic Curve Model [5]

The variable x increases as t increases at a rate given by

$$\frac{dx}{dt} \;=\; bx \; (1 - \frac{x}{L}) \qquad\qquad (3.14)$$

where L and b are given constants.
The curve represents x as a function of t.
To find x as a function of t

$$dx \;=\; bx \; dt - \frac{bx^2}{L} \; dt$$

$$Ldx \;=\; Lbx \; dt - bx^2 dt$$
$$\;=\; x(L-x)b \; dt$$

$$\frac{L dx}{x(L-x)} = b\, dt$$

$$\int \frac{L\ dx}{x(L-x)} = bt + k.$$

But $\quad \int \frac{L\ dx}{x(L-x)} = \int (\frac{1}{x} + \frac{1}{L-x})\, dx = \log x - \log(L-x)$

$$= -\log \frac{L-x}{x}$$

So $\quad \log \frac{L-x}{x} = -bt - k$

$$\frac{L-x}{x} = e^{-bt} e^{-k} = a\ e^{-bx}$$

where,

$\quad a = e^{-k}$ is an arbitrary constant.

Rearranging,

$$x = \frac{L}{1 + a\ e^{-bt}} \cdot \qquad (3.15)$$

This is an S - shaped growth curve where x has the
initial value of zero at equal to minus infinity.
The curve is symmetrical about its inflection
point x = L/2. The constants should determine
where the curve will be in time and the steepness of
the sharply rising portion. When $\ln (L/x - 1)$ is
plotted against time a straight line is obtained
which can be extrapolated into the future. From
historical data setting

$$Y_i = \ln(L/x - 1) \qquad (3.16)$$

values of Y_i corresponding to time t_i are obtained
and the expression

$$\sum_i^N (Y_i - \ln a + b\, t_i)^2 \qquad (3.17)$$

is minimized to obtain a regression fit of Y on t.
The initial growth is slow and the upper portion
flattens as it approaches the limit although in some
cases this may not be physically achievable. The
logistic model above is to be used in developing
predictive models where road traffic data and RTAs have
shown growing (increasing) tendencies with the passage
of time particularly where S-shaped curves are observed.

3.2.2 Linear Model [5]

In the linear model x varies linearly with
t such that

$$\frac{dx}{dt} = \beta \qquad (3.18)$$

$$dx = \beta dt$$

$$\int dx = \int_o^t \beta dt$$

$$x = \beta t + \alpha \qquad (3.19)$$

where α and β are regression constants calculated
as

$$\beta \ = \frac{n\Sigma xt - \ x \ \Sigma t}{n\Sigma t^2 - (\Sigma t)^2} \qquad\qquad (3.20)$$

$$\alpha \ = \frac{\Sigma x - \beta.\Sigma t}{n}$$

where n is the number of pairs of variables.
This model is to be used for the final analysis of
each predictive model developed where predicted data
using the predictive models is compared with observed
road traffic data as well as RTAs data.

3.2.3 Parabolic Model [5]

In this model the variation in x is equated
to the acceleration of a new body in equilibrium to
a new equilibrium state. The accumulation of the
variable is analogous to the distance travelled
by the new mass. The rate of x's generation is
equivalent to speed and the second derivative of
x over time is equivalent to acceleration.
Mathematically stated this yields

$$\frac{d^2x}{dt^2} \ = \ a \ constant, \ g \qquad\qquad (3.22)$$

$$\frac{dx}{dt} \ = \ gt$$

$$x = \int_o^t gt\ dt = \frac{g}{2} t^2$$

which may be written more completely as

$$x = \alpha + \beta t + \gamma t^2. \qquad (3.23)$$

The parameters α, β, γ can be solved by the solution of the normal equations

$$\Sigma x = \alpha n + \beta \Sigma t + \gamma \Sigma t^2$$

$$\Sigma xt = \alpha \Sigma t + \beta \Sigma t^2 + \gamma \Sigma t^3 \qquad (3.24)$$

$$\Sigma xt^2 = \alpha \Sigma t^2 + \beta \Sigma t^3 + \gamma \Sigma t^4$$

The parabolic model is a curve with one bend. This model is to be used where observed data show a curve with a single bend.

3.2.4 Cubic and Higher Polynomial Models

Following similar methods as those used for the linear model and the parabolic model the cubic model is of the form [5]

$$x = \alpha + \beta t + \gamma t^2 + \lambda t^3. \qquad (3.25)$$

The cubic model is a curve with two bends. The parameters α, β, γ and λ can be solved by the solution of the normal equations

$$\Sigma x = \alpha n + \beta\Sigma t + \gamma\Sigma t^2 + \lambda\Sigma t^3$$

$$\Sigma xt = \alpha\Sigma t + \beta\Sigma t^2 + \gamma\Sigma t^3 + \lambda\Sigma t^4 \qquad (3.26)$$

$$\Sigma xt^2 = \alpha\Sigma t^2 + \beta\Sigma t^3 + \gamma\Sigma t^4 + \lambda\Sigma t^5$$

$$\Sigma xt^3 = \alpha\Sigma t^3 + \beta\Sigma t^4 + \gamma\Sigma t^5 + \lambda\Sigma t^6.$$

Polynomials of higher degrees may be solved by increasing the number of parameters and the number of normal equations correspondingly. Cubic and higher polynomial curves are to be used where road traffic as well as RTAs data show curves with more than one bend.

Finite difference methods [8] as numerical methods can also be used to get approximate solutions. The principle by which they operate is that very simple equations are adequate to describe the function of a variable over very short distances and times.

Finite-difference methods are applicable to functions for which values are available at equidistant points. Given a series of points

$$x_n = x_o + n.h \qquad (n = 0,1,\ldots,N)$$

with corresponding function values

$$f_n \ (n = 0,1,\ldots,N).$$

One of the many problems related to the analysis of
experimental data is the representation of data by
analytical formulae such as those above. Finite-
differences are useful in such analysis. The
simplest analysis on a table of values is to find the
difference between each pair by subtracting each
value from its successor in the table, second
differences by repeating a similar process on the
first differences and so on for higher orders. These
differences together comprise the finite differences
of the table. Considering a set of pairs of values
(x_i, y_i), where $i = a, 1, \ldots, n-1$, which can be
represented by points in the xy plane, the differences
between successive pairs of ordinates y_{i+1} and y_i
is denoted by Δy_i, where Δ is the difference
operator.

Thus,

$$\Delta y_i = y_{i+1} - y_i \qquad i = 0,1,2,\ldots,n-1. \quad (3.27)$$

The second forward differences are defined by

$$\Delta^2 y_i = \Delta y_{i+1} - \Delta y_i \qquad \text{and in general, the kth}$$

forward differences are

$$\Delta^k y_i = \Delta^{k-1} y_{i+1} - \Delta^{k-1} y_i. \qquad (3.28)$$

If the rth differences $\Delta^r y_i$ are constant then all the differences of order higher than r are zero. From equations 3.27 and 3.28 it follows that

$$y_i = y_o + \Delta y_o$$

$$y_2 = y_1 + \Delta y_1 = (y_o - \Delta y_o) + (\Delta y_o + \Delta^2 y_o) = y_o + 2 y_o + \Delta^2 y_o$$

$$y_3 = y_2 + \Delta y_2 = (y_o + 2\Delta y_o + \Delta^2 y_o) + (\Delta y_1 + \Delta^2 y_1)$$

$$= (y_o + 2\Delta y_o + \Delta^2 y_o) + (\Delta y_o + \Delta^2 y_o + \Delta^2 y_o + \Delta^3 y_o)$$

$$= y_o + 3\Delta y_o + 3\Delta^2 y_o + \Delta^3 y_o,$$

Thus

$$y_1 = (1+\Delta) y_o$$

$$y_2 = (1+\Delta)^2 y_o$$

$$y_3 = (1+\Delta)^3 y_o$$

in which $(1+\Delta)^k$ is an operator on y_o with the exponent on the Δ indicating the order of the difference. By induction

$$y_k = (1+\Delta)^k y_o \qquad k = 1,2,\ldots \qquad (3.29)$$

and expanding

$$y_k = y_o + k\Delta y_o + \frac{k(k-1)}{2!} \Delta^2 y_o + \frac{k(k-1)(k-2)}{3!} \Delta^3 y_o + \ldots$$

$$(3.30)$$

With the assumption that the values x_i in a given set of data (x_i, y_i), where $i = 0, 1, 2, \ldots, n$ are equally spaced with the spacing interval h

$$x_1 = x_o + h, \quad x_2 = x_o + 2h, \ldots, \quad x_n = x_o + nh.$$

Letting the data be represented by some formula $y = f(x)$ which for $x = x_o + kh$ yields $y_k = f(x_o + kh)$, noting that $f_k = y_k$ formula (3.30) yields for $r \leqslant k$,

$$y_k = y_o + \binom{k}{1} \Delta y_o + \binom{k}{2} \Delta^2 y_o + \ldots + \binom{k}{r} \Delta^r y_o \quad (3.31)$$

where the binomial coefficients $\binom{k}{r}$ are defined by

$$\binom{k}{r} = \frac{k(k-1)(k-2)\ldots(k-r+1)}{r!}. \quad (3.32)$$

Since the x_i are spaced h units apart,

$$x_k = x_o + kh \qquad k = 1, 2, \ldots, n$$

so that

$$k = \frac{x_k - x_o}{h}. \quad (3.33)$$

The expression (3.31) is a polynomial of degree r in k. On substituting in 3.30 for k from 3.32, a polynomial of degree r in x_k is obtained. Collecting like powers of x_k equation 3.30 takes the form

$$y_k = a_o + a_1 x_k + a_2 x_k^2 + \ldots + a_r x_k^r. \quad (3.34)$$

Accordingly, the polynomial in x

$$y(x) = a_o + a_1 x + a_2 x^2 + \ldots + a_r x^r \qquad (3.35)$$

assumes the values y_k when $x = x_k$. Thus when the
rth differences of the y_k are constant and the x_k
are equally spaced, the polynomial (3.35) represents
these data exactly. When rth differences in a given
set of data are not constant but differences are
negligible, the polynomial (3.35) represents the
data approximately. The finite difference technique
is to be used to test tabulated data observed in
order to ascertain the suitability of fitting
polynomial curves on to such data.

In order to determine the maximum values the
first order condition will generally give the value
or values of the independent variable for which $\frac{dx}{dt} = 0$.
The value of the independent value for which this is
the case is called the critical value for the function
in question. In order to ascertain whether the
critical value so obtained constitutes a relative
maximum, the second order condition for a relative
maximum is applied which is

$$\frac{d^2 x}{dt} < 0.$$

Thus for example for a cubic function $\frac{dx}{dt}$ which is a
parabola is solved by using the general quadratic

equation

$$\alpha + \beta t + \gamma t^2 = 0 \qquad \gamma \neq 0$$

$$\text{and } t = \frac{-\beta \pm \sqrt{(\beta^2 - 4\alpha\gamma)}}{2\gamma} \qquad (3.36$$

3.2.5 Exponential Model [5]

This model is based on the simple explanation that the variation in the variable x is proportional to the level of the variable at any given time

$$\frac{dx}{dt} \propto x \qquad (3.37)$$

$$\frac{1}{2}dx = \beta dt$$

$$\int \frac{1}{x}dx = \int_o^t \beta dt$$

$$\ell nx = \beta t + \alpha$$

$$x = \alpha e^{\beta t}. \qquad (3.38)$$

To solve for the parameters α and β equation (3.38) is transformed into

$$\ell nx = \ell n\alpha + \beta t \qquad \text{and } \alpha \text{ and } \beta$$

calculated by the expressions

$$\beta = \frac{n\Sigma(t\ell nx) - t\Sigma \ell nx}{n\Sigma t^2 - (\Sigma t)^2} \qquad (3.39)$$

$$\alpha = e(\frac{\Sigma \ell nx - \beta \Sigma t}{n}).$$ (3.40)

where   n   is the number of pairs of the variables.
This model as a growth curve is to be used in the
preliminary analysis of fitting observed data
particularly RTAs data at national level that has
been observed as increasing with time.

3.2.6 Logarithmic Model [5]

This model is based on the simple explanation
that the variation in the variable   x   is proportional
to the inverse of the time

$$\frac{dx}{dt} \propto \frac{1}{t}$$ (3.41)

$$dx = \frac{\beta}{t} dt$$

$$\int dx = \int_{o}^{t} \frac{\beta}{t} dt$$

$$x = \beta \ell nt + \alpha.$$ (3.42)

To solve for the parameters α and β the expressions

$$\beta = \frac{n\Sigma(x\ell nt) - \Sigma \ell nt \Sigma x}{n\Sigma(\ell nt)^2 - (\Sigma \ell nt)^2}$$ (3.43

$$\alpha \quad \frac{\Sigma x - \beta \Sigma \ell nt}{n}$$ (3.44)

are solved where, n is the number of pairs of
the variables. This growth model is to be used in
fitting data and developing predictive models for
observed data particularly RTAs data at national
level which has been observed as showing an increasing
trend over time.

3.2.7 Power Model [5]

This model is based on the simple explanation
that the variation in the variable x is proportional
to the level of the variable and the inverse of
time

$$\frac{dx}{dt} \propto \frac{x}{t} \qquad (3.45)$$

$$\frac{1}{x}dx = \frac{\beta}{t}dt$$

$$\int \frac{1}{x}dx = \int_{o}^{t} \frac{\beta}{t}dt$$

$$\ln x = \beta \ln t + \ln \alpha$$

$$x = \alpha t^{\beta} \qquad (3.46)$$

To solve for the parameters α and β the expression

$$\beta = \frac{n\Sigma(\ln t \ \ln x) - \Sigma \ln t \ \Sigma \ln x}{n\Sigma(\ln t)^2 - (\Sigma \ln t)^2} \qquad (3.47)$$

$$\alpha = \frac{\Sigma nx - \beta \Sigma \ln t}{n} \qquad (3.48)$$

are solved where n is the number of pairs of the variables. This growth model is to be used for fitting data at the preliminary analysis stage where observed data shows increasing tendencies with time.

3.3. Time Series [5,6]

A time series is a set of observational data taken at specified times more often than not at equally spaced intervals. Mathematically a time series is defined by the values Y_1, Y_2, ..., Y_n of a variable at times t_1, t_2, ..., t_n. Thus Y is a function of t, (with Y the dependent variable and t the independent variable)

$$Y = F(t). \tag{3.49}$$

Using the equation

$$y = (\frac{\Sigma xy}{\Sigma x^2})x \tag{3.50}$$

where

$$x = X - \bar{X} \tag{3.51}$$

$$y = Y - \bar{Y}. \tag{3.52}$$

The trend line has the equation

$$Y = A + BX \tag{3.53}$$

where the values of Y computed for various values of X are called trend values. The origin $X = 0$ is the base year and the units of X are 1 year.

If values of X are assigned to the years for which data is available beginning from the base year to the last so that $\Sigma X = 0$, the equation of the least square line can be derived as

$$Y = \bar{Y} + (\frac{\Sigma XY}{\Sigma X^2})X \qquad (3.54)$$

where X replaces x which was given by equation 3.51 i.e. $\Sigma(X-\bar{X}) = 0$ and Y = y the initial values.

$$\bar{Y} = (\frac{\Sigma Y}{N}) \qquad (3.55)$$

where N is the number of years. If the number of years (N) is even the equation must be modified. For N even, the column of x is doubled to yield $\Sigma X=0$ and the origin becomes January 1 (between July 1 of the two middle years. The resulting equation (3.54) has X with units of half years. To measure X in whole years instead of half years, X is replaced by 2X but the origin remains in January 1 as before. For N odd the middle year is the origin (X=0) and the values of Y refer to mid-year values i.e. as of July 1.

The components of a time series are the characteristic movements T, C, S and I and are related by the equation

$$Y = T \times C \times S \times I \qquad (3.56)$$

where, T is the long-term secular movements,

secular variation

or secular trend indicated by the trend curve,

C is the cyclical movements,

S is the seasonal movements,

I is the irregular, random movements or

residual influences.

The analysis of the factors T, C, S and I is
called the decomposition of a time series.

The amount of variation present in the time
series data can be reduced by the use of techniques
such as moving averages. The elimination of these
unwanted fluctuations is called the smoothing of a
time series.

Given a set of variables Y_1, Y_2, Y_3,...,Y_n a moving
average of order N is defined by the sequence of
arithmetic means

$$\frac{Y_1+Y_2+ \ldots +Y_N,}{N} \qquad \frac{Y_2+Y_3+\ldots+Y_{N+1},}{N} \qquad \frac{Y_3+Y_4+\ldots Y_{N+2}}{N}$$

$$\ldots \ (3.57)$$

The sums in the numerators are called moving totals
of order N. If the data is given annually, a moving
average of order N is called an N year moving
average. Any other unit of time can be used. If
weighted arithmetic means are used, the weight being

prespecified, the sequence is called a weighted
moving average of order N.

The estimation of the trend curve is achieved
through the following steps in the time series
analysis:

-    data collection,
-    graphing, noting qualitatively the presence of
     long-term trend, cyclical varations and
     seasonal variations,
-    construction of the long-term trend curve or
     line by use of least squares method or moving
     averages method
-    prediction and error evaluation.
Time series techniques are to be used where data
observed particularly RTAs data at national level
shows growing tendencies over time. The techniques
are to be used in fitting trend line (curve) equations
and data smoothing.

3.4   Harmonic Analysis [8]

This is the problem of representing a suitable
periodic function in a trigonometric series. The
problem reduces to one of fitting a finite trigono-
metric function to a set of observed values (x_i, y_i).

Letting the set of observed values (x_0, y_0),
$(x_1, y_1), \ldots, (x_{2n-1}, y_{2n-1})$, (x_{2n}, y_{2n}), $\ldots$ be such

that the values of y start repeating with y_{2n}

(i.e., $y_{2n}=y_0$, $y_{2n+1} = y_1$ etc.). The assumption is

that the x_i are equally spaced, $x_0=0$, and that

$y_{2n}=2\pi$. On the basis of these assumptions

$$x_i = \frac{i2\pi}{2n} = \frac{i\pi}{n}.$$

The trigonometric polynomial

$$y = A_0 + \sum_{k=1}^{n} A_k \cos kx + \sum_{k=1}^{n-1} B_k \sin kx \qquad (3.58)$$

contains the 2n unknown constants $A_0, A_1, A_2, \ldots, A_n, B_1,$

$B_2, \ldots, B_{n-1}$, which can be determined so that equation

(3.58) will pass through the 2n given points

x_i, y_i by solving the 2n simultaneous equations

$$y_i = A_0 + \sum_{k=1}^{n} A_k \cos kx_i + \sum_{k=1}^{n-1} B_k \sin kx_i \quad i=0,1,2,\ldots,2n-1.$$

Since $x_i = \frac{i\pi}{n}$, these equations become

$$y_i = A_0 + \sum_{k=1}^{n} A_k \cos \frac{ik\pi}{n} + \sum_{k=1}^{n-1} B_k \sin \frac{ik\pi}{n}$$

$$i=0,1,2,\ldots,2n-1. \qquad (3.59)$$

The solution of equations 3.59 is done by means of a scheme similar to that used for the determination of the Fourier coefficients. The coefficient of A_o is unity. Therefore, multiplying both sides of each equation by the coefficient of A_o and adding the results yields

$$\sum_{i=o}^{2n-1} y_i = 2n\, A_o + \sum_{k=1}^{n} (\sum_{k=1}^{2n-1} \cos \frac{ik\pi}{n}) A_k + \sum_{k=1}^{n-1} (\sum_{i=o}^{2n-1} \sin \frac{ik\pi}{n}) B_k .$$

Now,

$$\sum_{i=o}^{2n-1} \cos \frac{ik\pi}{n} = 0 \qquad k = 1,2,\ldots,n$$

and

$$\sum_{i=o}^{2n-1} \sin \frac{ik\pi}{n} = 0 \qquad k = 1,2,\ldots,n-1 .$$

Therefore

$$2nA_o = \sum_{i=o}^{2n-1} y_i . \qquad\qquad (3.60)$$

Multiplying both sides of each equation in 3.59 by the coefficient of A_j in it, and adding the results, yields

$$\sum_{i=o}^{2n-1} y_i \cos \frac{ij\pi}{n} = \sum_{k=1}^{n} (\sum_{i=o}^{2n-1} \cos \frac{ik\pi}{n} \cos \frac{ij\pi}{n})$$

$$A_k + \sum_{k=1}^{n-1} (\sum_{i=o}^{2n-1} \sin \frac{ik\pi}{n} \cos \frac{ij\pi}{n}) B_k$$

for $j = 1,2,\ldots,n-1,$

But

$$\sum_{i=0}^{2n-1} \cos \frac{ik\pi}{n} \cos \frac{ij\pi}{n} = 0 \qquad \text{if} \quad k \neq j$$
$$= n \qquad \text{if} \quad k = j$$

and

$$\sum_{i=0}^{2-1} \sin \frac{ik\pi}{n} \cos \frac{ij\pi}{n} = 0 \qquad \text{for all values of } k.$$

therefore,

$$n A_j = \sum_{i=0}^{2n-1} y_i \cos \frac{ij\pi}{n} \qquad j = 1, 2, \ldots, n-1. \quad (3.61)$$

To determine the coefficient of A_n the same procedure is followed, but

$$\sum_{i=0}^{2n-1} \cos \frac{ik\pi}{n} \cos i\pi = 0 \qquad \text{if} \quad k \neq n$$
$$= 2n \qquad \text{if} \quad k = n.$$

Hence,

$$2n \, An = \sum_{i=0}^{2n-1} y_i \cos i\pi. \qquad (3.62)$$

On multiplying both sides of each equation of (3.59) by the coefficient of B_k in it and adding, it is found that

$$nB_j = \sum_{i=0}^{2n-1} y_i \sin \frac{ij\pi}{n} \qquad (3.63)$$

$$j = 1, 2, \ldots, n-1.$$

Equations 3.60 to 3.63 give the constants
in equation 3.58 . A compact schematic arrangement
is used to simplify the labour of evaluating these
constants. The method is based on the equations
that determine the constants together with trigono-
metric relations such as

$$\sin{}^{\pi}/n = \sin \left(\frac{n-1}{n}\right)\pi = -\sin \left(\frac{n+1}{n}\right)\pi = -\sin \frac{(2n-1)\pi}{n}$$

$$\cos{}^{\pi}/n = -\cos \frac{(n-1)\pi}{n} = -\cos \frac{(n+1)\pi}{n} = \cos \frac{(2n-1)\pi}{n}.$$

For a six-ordinate scheme $2n=6$; the given
data being (x_i,y_i), where $x_i = \frac{i\pi}{3}$ $(i=0,1,2,3,4,5)$;
and equation 3.58 yields

$$y = A_o + A_1 \cos x + A_2 \cos 2x + A_3 \cos 3x + B_1 \sin x + B_2 \sin 2x.$$

With the notation given below

y_o	y_1	y_2		v_o	v_1	w_o	w_1
y_3	y_4	y_5			v_2		w_2
sum	v_o	v_1	v_2	p_o	p_1	r_o	r_1
diffe-rence	w_o	w_1	w_2		q_1		s_1

equations 3.60 to 3.63 , with n=3, yield

$$6A_o = \sum_{i=o}^{5} y_1 = y_o+y_1+y_2+y_3+y_4+y_5 = p_o+p_1$$

$$3A_j = \sum_{i=o}^{5} y_1 \cos \frac{ij\pi}{3} \qquad j=1,2$$

$$3A_1 = \sum_{i=o}^{5} y_i \cos \frac{i\pi}{3} = y_o+y_1\cos\frac{\pi}{3} +y_2\cos\frac{2\pi}{3} + y_3\cos\pi$$

$$+ y_4\cos\frac{4\pi}{3} + y_5\cos\frac{5\pi}{3}$$

$$= y_o+\frac{1}{2}y_1 - \frac{1}{2}y_2 - y_3-\frac{1}{2}y_4 + \frac{1}{2}y_5$$

$$3A_2 = \sum_{i=o}^{5} y_i\cos \frac{i2\pi}{3} = y_o+y_1\cos \frac{2\pi}{3}+ y_2\cos\frac{4\pi}{3}+ y_3\cos2\pi$$

$$+ y_4\cos\frac{8\pi}{3} + y_5\cos\frac{10\pi}{3}$$

$$= y_o-\frac{1}{2}y_1 - \frac{1}{2}y_2 + y_3 - \frac{1}{2}y_4 - \frac{1}{2}y_5$$

$$6A_3 = \sum_{i=o}^{5} y_1\cos i\pi = y_o+y_1\cos\pi+y_2\cos2\pi+y_3\cos3\pi$$

$$= y_o-y_1+y_2-y_3+y_4-y_5$$

$$3B_j = \sum_{i=o}^{5} y_1 \sin \frac{ij\pi}{n} \qquad j = 1,2$$

$$3B_1 = \sum_{i=o}^{5} y_i \sin \frac{i\pi}{3} = y_1 \sin \frac{\pi}{3} + y_2 \sin \frac{2\pi}{3} + y_4 \sin \frac{4\pi}{3} \sin \frac{5\pi}{3}$$

$$= \frac{\sqrt{3}}{2} y_1 + \frac{\sqrt{3}}{2} y_2 - \frac{\sqrt{3}}{2} y_4 - \frac{\sqrt{3}}{2} y_5$$

$$3B_2 = \sum_{i=o}^{5} y_i \sin \frac{2i\pi}{3} = y_1 \sin \frac{2\pi}{3} + y_2 \sin \frac{4\pi}{3} + y_4 \sin \frac{8\pi}{3}$$

$$+ y_5 \sin \frac{10\pi}{3} = \frac{\sqrt{3}}{2} y_1 - \frac{\sqrt{3}}{2} y_2 + \frac{\sqrt{3}}{2} y_4 - \frac{\sqrt{3}}{2} y_5$$

giving

$$6A_o = p_o + p_1, \quad 3A_1 = r_o + \frac{1}{2} s_1, \quad 3A_2 = p_o - \frac{1}{2} p_1$$

$$6A_3 = r_o - s_1, \quad 3B_1 = \frac{\sqrt{3}}{2} r_1, \quad 3B_2 = \frac{\sqrt{3}}{2} q_1$$

with the checks on the calculations given by

$$A_o + A_1 + A_2 + A_3 = y_o \quad \text{and} \quad B_1 + B_2 = \frac{\sqrt{3}}{3} (y_1 - y_5).$$

For a 24-ordinate scheme $2n=24$, $n=12$, given data (x_i, y_i) where $x_i = \frac{i\pi}{12}$ $(i = 0,1,2,\ldots,22,23)$ equation 3.58 becomes

$$y = A_o + \sum_{k=1}^{12} A_k coskx + \sum_{k=1}^{11} B_k sinkx.$$

To calculate all the coefficients and set out a scheme for their solution equations 3.60 to 3.63 are used as

$$24 A_o = \sum_{i=o}^{23} y_i, \quad 12A_1 = \sum_{i=o}^{23} y_i cos \frac{i\pi}{12}, \dots$$

Harmonic analysis techniques are to be used in representing suitable periodic functions in a trigonometric series for observed road traffic as well as RTAs data where such data show periodic variation for example, the variation of road traffic flow and the variation of RTAs over a period of 24 hours. The periodic functions will then be fitted to predict the variation of traffic or RTAs.

3.5 Errors in Prediction Models

3.5.1 Errors and Confidence Levels of Predictions [5,6,7]

It is desirable to assess the confidence levels in predictions and to have some measure of the proba-bility of error. This is necessary because predictive models, particularly when they are approximating functions, contain inherent errors. The assessment

of the errors is done by regression analysis and
correlation analysis which are measures of fitting curves
to the variation of errors and a measurement of the degree
of fit respectively. This is achieved by determining
the standard error of estimate and the calculation of
the correlation coefficient.

Just as the standard deviation measures the varia-
tion of the values of a variable about their arithmetic
mean, the standard error of the estimate is a prediction
of the scatter of the variables about a line. The
standard error of the estimate, symbolized by $S_{y.x}$ repre-
sents the standard deviation of the y's on the x's.

$$S_{y.x} = \left(\frac{\Sigma (y - \bar{y}_p)^2}{n-2} \right)^{\frac{1}{2}} \qquad (3.64)$$

where,

$(y - \bar{y}_p)$ are vertical deviations from the regression
line, $\bar{y}_p$ is the predicted value and

n-2 are the degrees of freedom.

If a large number of data are observed, calculating
each $\bar{y}_p$ point on the regression line and then squaring
the differences is very involving. For ease of compu-
tation the formula used is

$$S_{y.x} = \left(\frac{\Sigma y^2 - a(\Sigma y) - b(\Sigma xy)}{(n-2)} \right)^{\frac{1}{2}} . \qquad (3.65)$$

Theoretically the standard error of the estimate is
a valid measure in setting the confidence limits
about a predicted value $\bar{y}_p$ if the size of the
sample is large and the points on the scatter
diagram are normally distributed.

In the confidence limits

$\bar{y}_p \pm S_{y.x}$ encompasses the middle 68% of the
data points,

$\bar{y}_p \pm 1.96 \ S_{y.x}$ encompasses the middle 95% of
of the data points and

$\bar{y}_p \pm 3.S_{y.x}$ encompasses the middle 99.7%
of the data points.

The equation of the confidence level is

$$\bar{y}_p \pm t(S_{y.x}) = \left(\frac{1}{n} + \frac{(x-\bar{x})^2}{\Sigma (x-\bar{x})^2} \right)^{\frac{1}{2}} \qquad (3.66)$$

where n = number of observation

t = value from Students t - distribution
obtainable from statistical tables.

Generally, the standard error of estimate
is rather difficult to interpret. A standard error
of zero means that all of the variation is explained
by x. The proportion of the variation explained is
called the coefficient of determination r^2 and the
unexplained variation is called the coefficient of
non-determination denoted by (k^2). The value of the

coefficient of determination varies from 0 to 1. A
coefficient of determination zero indicates that none
of the variation in y is explained by the variable
x. A coefficient of determination 1 indicates that
100 per cent of the variation in y is explained
by x. The variation in y which is not associated
with x is measured by the standard error of the
estimate $S_{y.x}$. To convert the variation to a
coefficient $S_{y.x}$ is divided by the total variation
thus,

$$r^2 = 1 - \frac{\text{unexplained variance}}{\text{total variance}} = 1 - \frac{S_{y.x}^2}{S_y^2} . \quad (3.67)$$

The square root of the coefficient of determination
is called the correlation coefficient. A more
convenient formula for calculating the correlation
coefficient is

$$r = \frac{n(\Sigma xy) - (\Sigma x)(\Sigma y)}{\{[n(\Sigma x^2) - (\Sigma x)^2][n(\Sigma y^2) - (\Sigma y)^2]\}^{\frac{1}{2}}}$$

The value of r lies between + 1 (for perfect
positive relationships) and -1 (for perfect negative
relationships). A correlation coefficient of 0.9
and greater is considered quite significant.
To calculate the level of significance the value of t
in equation 3.66 may be calculated from the equation

$$t = \frac{r(n-2)^{\frac{1}{2}}}{(1-r^2)^{\frac{1}{2}}} \qquad\qquad (3.69)$$

where, r is the correlation coefficient,

r^2 is the coefficient of determination and

n   are the number of observations.

The statistical techniques above are to be used
in evaluating the quality of fitness of the predictive
models developed and the confidence levels of such
predictions.

3.5.2 <u>Method of Final Analysis on Predicted Data</u>

In order to predict the relative accuracies of
the models developed, each model is used as applicable
to predict the dependent variable whose variation is
being sought to be fitted to a curve. Simple linear
regression is used to compare the observed data with
the predicted data by each model. A similar test is
applied to models, developed by other researchers,
predicting the same variable. Perfect prediction by
any formula would result in a regression line whose
slope = 1, intercept = 0, a correlation coefficient
r of 1.0 and the critical coefficient or the
coefficient of determination r^2 = 1.   The less
accurate a prediction is, the more the regression
line varies from this ideal and the lower the corre-
lation coefficient.   A further measure of the data

scatter for each regression is obtained by calcu-
lating the standard error about the observed =
predicted line. The slope and correlation coeffi-
cient for various predictive models can be compared
since these are dimensionless.

3.6 Generalised Linear Models [59]

A generalised linear model is generally regarded
as consisting of two elements. These are the systematic
component and the random component. The systematic
component describes the way the predicted values of
the dependent variable relate to a set of independent
variables. In the ordinary least squares regression
the fitted equation has the form

$$\mu = a_o + a_1x_1 + a_2x_2 + a_3x_3 + \ldots \tag{3.70}$$

where μ is the value of the dependent variable
predicted by the regression line for a particular set
of the independent variables. The a's are the
regression coefficients. The generalised linear model
preserves the linear form of the right-hand side of
equation 3.70, but generalises the relationship between
the value of the linear predictor denoted by η and
the fitted value μ yielding the equations

$$\eta = a_o + a_1x_1 + a_2x_2 + a_3x_3 + \ldots \tag{3.71}$$

$$\eta = f(\mu). \tag{3.72}$$

Equation 3.72 is a relationship giving the link function.
For accident data, the dependent variable is the number
of accidents occurring at a particular site within a
given period. The number of accidents is normally
regarded as having a Poisson error structure. The
independent variables are those associated with traffic
flow and road geometry. By using a logarithm link
such that

$$\eta = \ln (\mu) \qquad\qquad (3.73)$$

where $\ln$ is the logarithm to the base e a multipli-
cative model becomes

$$A = e^k K^b e^{(\Sigma a_i x_i)} \qquad\qquad (3.74)$$

where A is the number of accidents, K is the vehicle-
kilometres of travel during the period under study,
a_i are the coefficients, x_i are the independent
variables and k is the constant.
The linear equation becomes

$$\ln A = k + b\ln K + a_1 x_1 + a_2 x_2 + a_3 x_3 + \ldots \quad (3.75)$$

The other component of the generalised model
is the random element. In least squares regressions
the observations are regarded as drawn from a population
with a mean equal to the value given by the regression
line. The variance is constant throughout the range

of the data. If the data is drawn from a normal population standard significance tests are applied and the least squares estimates of the regression parameters are maximum likelihood estimates. If the observations are drawn from a non-normal population, as in this case where in RTAs the error distribution is regarded as Poisson, the constant variance assumption is violated since the variance of the Poisson distribution equals its mean. The generalised linear model formula allows either a known or assumed error distribution of the dependent variable to be specified explicitly for the exponential family of distributions.

Once the link function and the error structure have been specified, the maximum likelihood estimates of the coefficients are calculated. The 'normal' equations are solved to give the coefficient estimates. These equations are similar to those for ordinary weighted least squares regression but the dependent variable is replaced by a modified variate given as

$$\eta + \delta(y-\mu)$$

where η is the linear predictor, y is the observed dependent variable, μ is the predicted value and δ is $\frac{d\eta}{d\mu}$ which is the derivative of the link function. The procedure is iterative. Each

cycle of the fit uses estimates of the various para-
meters from the previous cycle until convergence is
obtained. Convergence is obtained usually in 3 or 4
cycles.

For significance testing the Scaled Deviance
is used. Deviance is a likelihood ratio equal to
$-2 \ln\lambda$, where

$$\lambda \quad = \quad \frac{\max L}{\max L_f} . \tag{3.76}$$

Therefore,

Scaled Deviance $= -2[\ln(\max L) - \ln(\max L_f]$

where $\ln(\max L)$ is the maximised log-likelihood
for the model under review and $\ln(\max L_f)$ is the
corresponding value for the full model which exactly
fits all the data points $(\mu=y)$.

A generalised linear model is good if
Scaled Deviance (S.D)/degrees of freedom (d.f) is
near 1.

The generalised linear modelling technique will
be used through the computer program GLIM to obtain
unified RTA models for the single carriageway road
(Kiganjo-Nanyuki) and the dual carriageway road
(Nairobi-Thika).

CHAPTER 4 - ROAD TRAFFIC AND ACCIDENT
DATA COLLECTION AND ANALYSIS

In this chapter road traffic and road traffic
accident data collection and analysis are presented.
Road traffic data and analysis is presented followed
by road traffic accident data and analysis. Firstly,
road traffic data collection and analysis is presented
in three sections covering national traffic, dual'
carriageway (Nairobi-Thika Road) traffic and the single
carriageway (Kiganjo-Nanyuki Road) traffic. Secondly,
road traffic accident data collection and analysis is
also presented in three sections covering national road
traffic accidents, dual carriageway RTAs and the single
carriageway RTAs. The data analysis is based on the
threoretical analysis presented in Chapter 3.

4.1   Road Traffic Data Collection and Analysis

4.1.1 National Road Traffic

4.1.1.1. Data Collection

At the national level the data collected
relating to the Kenya transportation system (Fig.4.1)
included human population, cumulative number of vehicles,
vehicle composition and the total length of classified
roads.   These data were collected from the Central
Bureau of Statistics of Kenya (CBSK) for the years
1949-1983 (Appendix A.13).

FIG 4.1   KEY PLAN

4.1.1.2 Data Analysis

Data on human population was plotted as shown
in Fig.4.2. Using equation 3.15 and figures of
population for Kenya, as predicted by the United
Nations   [56] for the year 2025, models for growth
in population were developed to be used in relation
to the cumulative number of registered vehicles
to obtain degree of motorization. The United Nations
projects Kenya's population to be 103.738 and 53.314
million for high and low growth levels respectively.
These figures were used as the upper limits of growth
and the models developed were respectively

$$H_{p_1} = \frac{103.738}{1 + 18.797\ e^{-0.038t}} \qquad (4.1)$$

$$H_{p_2} = \frac{53.314}{1 + 9.254\ e^{0.042t}} \qquad (4.2)$$

for high and low growth rates, where

H_{p_1} is human population predicted using high
growth rate

H_{p_2} is human population predicted using low
growth rate

t   is time in years

e   is base of natural logarithms.

FIG.4.2 KENYA HUMAN POPULATION GROWTH.

On the basis of the method of analysis outlined in
3.5.2 and using equation 3.19 the regression
equations were

$$H_o = 1.004 \ H_{P_1} - 0.036$$

with r = 0.99, r^2 = 0.99 and standard error of 4.256,

$$H_o = 1.011 \ H_{P_2} - 0.112$$

With r = 0.99, r^2 = 0.99 and standard error of 4.271
where H_o is the observed human population data.   In
all the prediction models developed in this chapter
the relationship between observed and predicted data
for perfect prediction (section 3.5.2) is expected
ro result in a regression line whose slope is 1,
intercept is 0 and the correlation and determination
coefficients 1.

The predictive models for vehicles per popula-
tion (motorization) were based on the assumption that
the degree of vehicle ownership for Kenya will grow
and tend to the level experienced in Nairobi the
country's capital. According to the Nairobi Urban
Study Group [57] vehicle ownership was projected
to be 0.09 vehicles per person at the turn of the
century. From Fig.4.4 the overall national vehicle
ownership was found to be less than 0.02. Thus for
the purpose of model development the upper limit was
set at 0.1 vehicles per person. On this basis the two

FIG.4.3 KENYA MOTOR VEHICLE GROWTH.

upper limits for vehicles, corresponding to the
human population were derived as 5.3314 and 10.3738
million vehicles respectively for the year 2025.

Observed cumulative vehicles were plotted
(Fig.4.3) and using equation 3.15 the predictive
models for motor vehicles were developed as

$$V_{P_1} = \frac{10.3738}{1 + 281.817 \ e^{-0.064t}} \qquad (4.3)$$

$$V_{P_2} = \frac{5.3314}{1 + 144.667 \ e^{-0.065t}} \qquad (4.4)$$

where, V_{P_1} and V_{P_2} are predicted cumulative vehicles
using high and low growth rates respectively. The
comparison of observed and predicted data for
equations 4.3 and 4.4 were

$$V_O = 0.972 \ V_{P_1} + 0.003$$

with $r = 0.99$, $r^2 = 0.99$, standard error 0.062,

$$V_O = 0.973 \ V_{P_2} + 0.003$$

with $r = 0.99$, $r^2 = 0.99$ and standard error 0.061,
where V_O is observed cumulative vehicles.

GROWTH IN MOTORIZATION IN KENYA

$(V/P)_p = \dfrac{0}{1 + 12}$

COMPARISON OF OBSERVED
AND PREDICTED DATA
$(V/P)_o = 0.914(V/P)_p + 0.00075$

FIG.4.4 GROWTH IN VEHICLES PER PERSON (MOTORIZATION) IN KENYA

Using the figure of 0.1 vehicles per person for
the first upper limit for the degree of motorization
in Kenya, the observed vehicles per person were
plotted against time and the data smoothed by the
moving averages technique N = 5 years, using equation
3.57 to get rid of fluctuations (Fig.4.4). Using
equation 3.15 the predictive model for the growth
in motorization was developed as

$$(V/P)_p = \frac{0.1}{1 + 12.790\ e^{-0.028t}} \qquad (4.5)$$

where, $(V/P)_p$ is the predicted level of motorization
at time t in years. The comparison of observed and
predicted data for equation 4.5 was

$$(V/p)_o = 0.914\ (V/p)_p + 0.00075$$

with r = 0.95, r^2 = 0.90 and standard error of
0.00289, where $(V/p)_o$ is the observed degree of
motorization.

From the observed number of vehicles involved
in RTAs the highest percentage of RTAs vehicles with
respect to the cumulative vehicles was found to be
3.12 per cent. Using this figure and the upper
limit of cumulative vehicles at the lower growth
rate (equation 4.4) of 5.3314 million vehicles the
approximation of the upper limit of vehicles to be

involved in RTAs was estimated as 166 073 vehicles.
Using this limit and equation 3.15 the predictive
model for the growth of RTAs vehicles was developed
as (Fig.4.5)

$$(V_A)_p = \frac{166\ 073}{1 + 77.867\ e^{-0.052t}} \qquad (4.6)$$

where $(V_A)_p$ is the predicted number of vehicles
involved in RTAs.   The comparison of observed and
predicted data for equation 4.6   was

$$(V_A)_o = 0.882\ (V_A)_p + 451$$

with $r = 0.94$, $r^2 = 0.88$ and standard error of 1437.

Observed data on vehicle composition was plotted
and smoothed by moving averages.   The data was con-
verted into percentage vehicle composition for the
individual class or group of vehicle type in order
to determine their proportion in relation to the
cumulative vehicles.   After testing the data by
various models the curve of best fit was found to
be equation 3.42 , the logarithmic model.   The
model for the percentage composition of cars and
utilities was developed as (Fig.4.35)

$$((\%)_{CU})_p = 76.427 - 2.160\ \ell nt \qquad (4.7)$$

GROWTH IN RTA VEHICLES IN KENYA

$$(V_A)_P = \frac{166073}{1 + 77.8669\, e^{-0.05229\, t}}$$

O SMOOTHED DATA

COMPARISON OF OBSERVED AND
AND PREDICTED DATA

$$(V_A)_O = 0.882 (V_A)_P + 451$$

FIG. 4.5   GROWTH IN RTA VEHICLES IN KENYA

where, $((\%)_{CU})_p$ is the predicted eprcentage composition
of cars and utilities at time   t.   The composition
of observed and predicted data for equation   4.7
was

$$((\%)_{CU})_o = 0.994\ ((\%)_{CU})_p + 0.489$$

where $((\%)_{CU})_o$ is observed percentage composition,
with   r= 0.91, r^2 = 0.82 and standard error 1.785.
Using similar techniques as above the growth model
for the percentage composition of buses, lorries
and taxis was developed as (Fig.4.36)

$$((\%)_{BLT})_p = 11.873 + 0.239\ \ell nt \qquad (4.8)$$

where, $((\%)_{BLT})_p$ is the predicted percentage composi-
tion of buses, lorries and taxis at time t.   The
composition of observed and predicted data for
equation 4.8 was

$$((\%)_{BLT})_o = 1.158\ ((\%)_{BLT})_p - 2.070$$

where $((\%)_{BLT})_o$ is observed percentage composition,
with r = 0.38, r^2 = 0.14 and standard error 0.198.
The growth model for the percentage composition of
motorcycles was developed as (Fig.4.37)

$$((\%)_M)_p = 5.323 + 0.309\ \ell nt \qquad (4.9)$$

where $((\%)_M)_p$ is the predicted percentage composition

of motorcycles at year (time) t. The comparison of

observed and predicted data yielded the equation

$$((\%)_M)_o = 1.123 ((\%)_M)_p - 0.705$$

where $((\%)_M)_p$ is observed percentage composition,

with $r = 0.63$, $r^2 = 0.40$ and standard error of

.0.256.

The relationships developed as equations 4.1 to

4.9 were found to be significant at a level of

5 per cent.

4.1.2 Dual Carriageway (Nairobi-Thika Road) Traffic

The Nairobi-Thika Road was chosen for this

study because of a number of reasons. The most

important of these reasons are: as a dual carriageway

it has a higher capacity for traffic than any other

trunk road in Kenya thus providing a wide range of

variation in traffic volumes, speeds, vehicles compo-

sitions and geometric design, it passes through land

uses which vary from urban through semi-urban to

semi-rural which greatly influence traffic movement.

These reasons together with the fact that as a tarmac

road with a high level of service consequently making

it one of the roads in Kenya with a high potential for

RTAs made the choice of this road for study a natural or

The Nairobi-Thika Road is a Class A international trunk road according to the road classification of the Ministry of Transport and Communications (MOTC) of Kenya. This road is part of the trunk road designated A2 and is one of the two major trunk routes emanating from Nairobi on which the road system in Kenya's Central Province is focussed (Fig.4.1). The trunk route A2, commences in Nairobi skirts the eastern flanks of the Nyandarua via Thika and Muranga, serving as the through route for all traffic travelling in a north-south direction in the densely populated districts of Kiambu, Muranga, Nyeri and Kirinyaga. Due to the prohibitive deeply gullied topography of the area the main Nairobi-Thika Road is the sole connection between the numerous roads leading into the densely populated districts of Kiambu and Muranga. Nearly all the major urban centres of Central Province are located on or near this route and owe their development to the stimulus to trade provided by this bituminized road. The distance from Nairobi to Thika is some 38 kilometres making the total distance of the dual carriageway studied 76 kilometres. Each carriageway was built as a two lane 6.5 metre wide road. The design speed is 80 kilometres per hour.

4.1.2.1 Data Collection

The data collection points are shown in Fig.4.6.

FIG 4.6   NAIROBI – THIKA   ROAD   DATA   COLLECTION

END OF STUDY
KM 38 · 520
THIKA RIVER

BLUE   POST   HOTEL

THIKA

Km 35·000

JUJA   THIKA ROAD

Km 30·000

Km 25·000

LEGEND

○ TRAFFIC   COUNT   STATION

vehicle speeds, high volumes and poor lighting at
night. Fortunately, these limitations did not affect
the quality of the results since the enumerators were
trained well before hand to cope with such difficulties

The public transport counts were carried out
in order to provide the necessary background data for
RTAs analysis, to reveal the importance of this
transportion mode and to give specific background
information on the usage of matatus (mini-buses) as
opposed to buses. Two sections of the Nairobi-Thika
Road were chosen as representative. One located
near the Safari Park Hotel and another east of Ruiru.
The enumeration period was one week for each direction.
Day as well as night counts were made. Besides the
amount of buses and matatus, the number of passengers
were also recorded. A total of 8 enumerators
assisted by two policemen carried out the survey.
The policemen stopped the buses and matatus on a
queue and the enumerators quickly carried out the
counting of the passengers. The major constraint
was the delays experienced by these vehicles because
of stopping. This led to some annoyance to both
the operators as well as the passengers particularly
at the start of the enumeration period. Poor lighting
at night made the enumeration particularly of passe-
ngers difficult. These limitations however, did not
affect adversely the quality of the results.

Parallel with the vehicular traffic manual counts, pedestrians and pedal cyclists crossing the Nairobi-Thika Road were counted at some locations in order to give an indication of these road users level in order to provide the necessary background for understanding the RTAs related to these road users. The locations chosen were near the Drive-In-Cinema, near the Githurai junction and near the Juja junction. The enumaration near Githurai was done at half hour intervals. The enumaration for the other two sites was carried out as a whole from 6.30 a.m. to 6.30 p.m. to yield the approximate total number. The major limitation during this enumaration was the large number of pedestrians rushing across the carriageway at certain times, particularly at the start and end of the day.

4.1.2.2 Data Analysis

The directional vehicle movement counts showed that through traffic on Thika Road was dominating at intersections. Outer Ring Road and Garissa Road contributed most of the traffic onto Thika Road. Outer Ring had 40 per cent more than the through traffic. Outer Ring is a distributor road for traffic from Mombasa Road towards Thika and Nyeri area as well as serving the industrial corridor along the road. Garissa Road connects Thika Road to Thika Town an industrial and commercial town of rapid

development.   Ruiru urban area with industrial
activities equally contributed substantial traffic
especially lorries and buses.   Most of the buses from
Nairobi went through Ruiru Township towards Thika.

Although the enumeration days were different
for each site, the peak hours were occurring almost
at the same time.   The morning peak hour was between
7.30 a.m. and 8.30 a.m. whilst the afternoon peak
hour occurred between 4.30 p.m. and 5.30 p.m. with
some sections occurring between 5.00 p.m. and 6.00
p.m. (Fig.4.7).   The traffic variation at Ruiru
(Appendix A.1)differed from this general pattern.
It was observed that the heaviest morning traffic
was as late as 10.30 a.m. to 11.30 a.m.   This
enumeration point occurs more or less midway between
Nairobi and Thika.   Therefore due to its location
the peak hour is likely to be influenced by the
travel time from Nairobi to Ruiru and Thika to Ruiru.
Further, Ruiru has a number of industrial activities
which draw traffic from Nairobi and therefore such
traffic which leaves Nairobi and possibly Thika at
about 9.00 a.m. is likely to lead to this observed
traffic pattern.   The traffic from Nairobi reached
a peak between 6.30 p.m. and 7.30 p.m.

The vehicle counts revealed that the traffic
varied only slightly during the weekdays.   Sunday

FIG.4.7 PREDICTION OF AVERAGE TRAFFIC FLOW PER HOUR: NAIROBI - THIKA ROAD

had the lowest traffic volume at both locations, but
on Saturday the traffic at Ruiru on the Thika bound
carriageway reached a maximum. The reverse occurred
on Monday when traffic towards the city was above
average. This traffic pattern can be explained by
the recreational traffic. Lorries and buses had a
similar daily traffic variation pattern as the total
traffic. However, at peak hours the proportion
of buses and lorries was less than the off-peak
hours of the day. This was not surprising and was
confirmed by other surveys, because passenger cars
carrying people to and from work at those times were
dominating. During the week days, the percentage of
heavy traffic (lorries and buses) varied less near
Drive-In-Cinema (19-22%) than near Ruiru (19-30.5%)
of the total traffic.

The results of the public transport counts were
influenced by the four week period within which the
enumeration was done. During the first weekend of
August the high number of buses and matatu passengers
was found to be related to pay day. Buses were found
to carry the larger proportion of all passengers using
public transport. At Safari Park 66% - 79% of the
passengers were found to be bus users and 21%-34%
matatu users. At Ruiru the figures varied between 53%
and 71% for buses and 29%-47% for matatus. Comparing
the number of vehicles passing the enumeration sites,

matatus had a higher percentage. The survey revealed
that matatus are a vital factor in public transporta-
tion on the study road as well as the whole area. The
census indicated that many of the vehicles not only
matatus but buses as well, were overloaded with
passengers. The highest number of passengers recorded
in an ordinary matatu (weight 1070 kg) was 24 with
others carrying over 20. Ordinary matatus have a
passenger capacity varying between 13 and 20. Buses
with a capacity of 90 passengers had up to 200 passe-
ngers, but most of the buses did not exceed their
allowed capacity. From this analysis it was observed
that both buses and matatus are necessary in the
day to day passenger as well as goods transportation
at the same time their overloading being a significant
factor linked to RTAs (Appendix A.3).

Parallel with the vehicle counts, pedestrians
crossing Thika Road were recorded at some selected
locations. Near the Drive-In-Cinema where only the
total number was required some 1300 pedestrians
crossed the road in connection with the bus-stops
sited there. Many pedestrians were also observed to
be walking alongside the road on both sides as well
as in the median. The highest number of pedestrians
crossing the dual carriageway was recorded at the
Juja intersection, where some 4000 pedestrians and
250 bicyclists were crossing during the day. This

high number of crossings was related to the fact that
industrial use lies on one side of the road whilst
housing lies on the other coupled with the fact that
the shops are located close to the road on both
sides. In order to obtain the variation of pedestrian
and pedal cyclist crossings during the day the obser-
vations at Githurai had been done at hourly intervals
with a total of 2121 pedestrians and 83 (Appendix A.4)
pedal cyclists crossing the road. The crossings at
Githurai were made in connection with bus-stops and
the market located close to the road. Not surpri-
singly, these crossings are a significant factor
related to the RTAs on the study road.

In order to develop a predictive model, one
day's data, on traffic observed at Muthaiga, was
plotted and smoothed by moving averages (Fig.4.7).
Using the technique of harmonic analysis from 3.4
equations 3.60 to 3.63 were used and a six-
ordinate scheme developed to calculate the coeffi-
cients of a suitable periodic function in a trigono-
metric series to fit the observed data. To test the
effect of data smoothing two predictive models
were developed, one for unsmoothed data and an
improved model for the smoothed data. Both models
were analysed for fitness. For the unsmoothed
data the predictive model gave the following

expression:

$$q_{p_1} = 701.667 - 580 \cos t - 46.667 \cos 2t$$
$$+ 165 \cos 3t - 121.244 \sin t$$
$$- 144.338 \sin 2t \qquad (4.10)$$
$$40 < q \ 1800$$

where, q_{p_1} is the predicted number of vehicles per hour at time t. The comparison of observed and predicted data yielded the equation

$$q_o = 0.919 \ q_{p_1} + 36$$

with r = 0.91, r^2 = 0.83 and standard error of 458 where, q_o is the observed number of vehicles at time t.

Similarly, the equation for the smoothed data was developed as

$$q_{p_2} = 715.667 - 532.667 \cos t - 140.667 \cos 2t$$
$$+ 134.667 \cos 3t - 209.001 \sin t$$
$$- 162.813 \sin 2t \qquad (4.11)$$
$$40 < q < 1800$$

where, q_{p_2} is the predicted number of vehicles per hour at time t. The comparison of observed and predicted data yielded the equation

$$q_o = 0.945 \ q_{p_2} + 5$$

with $r = 0.93$, $r^2 = 0.86$ and standard error of 452

where, q_o is as above. Both models were tested

and found to be statistically significant at the

5 per cent level. The models can be improved by

adopting an ordinate scheme greater than six such

as 8 or better still using all data in a 24-ordinate

scheme. The effect of data smoothing can be seen in

the change in the slope and intercept of the linear

regressions obtained above by comparing the observed

and peridicted data. The slope improves from 0.919

to 0.945 and the intercept drops from 36 to 5 for

the two cases respectively. These two parameters

indicate, by the method of final analysis outlined

in section 3.5.2,that the calibration of the models,

particularly that obtained using smoothed data, is

quite acceptable and therefore the predictions are

close to the observed values since 0.945 is close to

1 and 5 (vehicles) is close to 0.

4.1.3 Single Carriangeway (Kiganjo-Nanyuki Road)Traffic

The Kiganjo-Nanyuki Road was chosen for this

study due to a number of reasons. The most important

of these reaons are: the continuation of the trunk

road A2 of which the Nairobi-Thika Road studied

earlier is part, it serves a typically rural area

where 60 per cent of all RTAs occur [15] and is

one of the rural tarmac roads which contribute 47

per cent of all RTAs in Kenya, it has a good geometric
design implying that the effect of geometric design
was likely to be of less significance compared to
the effect of junctions and pavement defects, it
offered a variety of pavement defects which at the
time of study were the major factors influencing
RTAs together with the many junctions and accesses
to the farms in this rural area.

The Kiganjo-Nanyuki single carriageway road is
of class A international trunk road classification
according to the MOTC. It is part of the trunk road
designated A2 and one of the major trunk routes from
Nairobi on which the road system in Kenya's Central
Province is focussed (Fig.4.1). The Kiganjo-Nanyuki
Road is located on the western side of Mount Kenya,
partly in the densely populated Nyeri District of
Central Province and partly in Laikipia District of
the Rift Valley Province. The road starts at the
junction of Kiganjo-Nanyuki Road A2 and the Nyeri-
Kiganjo secondary road C75, some 142 kilometres from
Nairobi and ends at Nanyuki. The total length of
the study road is 48 kilometres. It is a two lane
road of 6.1 metres width. The design speed is 80
kilometres per hour.

4.1.3.1 Data Collection

The data collection sites were located as shown

in Fig.4.8. The data collection comprised mainly of
classified traffic counts. These vehicle counts were
done manually using six persons. The manual counts
were carried out in order to obtain traffic volumes
on the main road, the turning vehicle movements at the
most important junctions and the daily traffic variation
and composition on the main road. The data were
further used in the analysis for developing predictive
models for road traffic, RTAs and as a background to
road traffic and road traffic accident characteristics.
Tally counters were used during the enumeration. The
enumeration phase lasted for two weeks Friday 21-1.83
to Thursday 27.1.83 and Tuesday 1.2.83 to Monday
7.2.83. The first week was used for recording the
classified manual counts between Naro Moru and
Nanyuki whilst the second week was used for the
Kiganjo-Naro Moru section. After a day's training
the enumerators performed satisfactorily. The enume-
rators worked in groups of twos for eight-hour shifts
with overlaps of up to one hour for changing the
teams. The counts included day and night traffic.
The major constraint in the enumeration was the
poor lighting conditions at night characteristic of
rural roads in Kenya. The enumerators were however,
equipped with torches and this limitation did not
adversely affect the results. The results were
plotted for each enumeration site (Appendix A.2).

START OF STUDY
KM 0 + 000

KIGANJO

KM 5 + 000

CHAKA

KM 10 + 000

KARICHENI

NYANGE

LOISOI

KM 15 + 000

THUNGARE

KM 20 + 000

AGUTHI

KM 25 + 000

RUARE

NARO MORU

WERUINI
MURIRO

KM 35 + 000

KM 40 + 000

BANTU
UTAMADUNI

GWARA

KM 30 + 000

RAGATI

KM 45 + 000

NANYUKI

END OF STUDY
KM 48 + 125

LEGEND

◯ TRAFFIC COUNT STATION

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FIG. 4.8   KIGANJO - NANYUKI ROAD DATA COLLECTION

Parallel with manual counts information on origin
and destination was obtained from heavy vehicles
to provide a background for RTAs contribution by
heavy vehicles.

4.1.3.2 Data Analysis

The traffic flow data obtained was compared
with the historical traffic data kept by the MOTC
and it was found that there was a general increase
in traffic in the area growing at annual rates
varying between 4.4 to 7.7 per cent. Through traffic
was found to dominate, being as high as 12 times that
of traffic turning at junctions. The classified manual
counts gave an indication of the daily traffic varia-
tions and distribution in the various vehicle types.
The analysis of the traffic composition revealed that
42 per cent of the traffic is passenger cars, 30 per
cent is light goods vehicles, 20 per cent is medium
goods vehicles, 5 per cent is heavy goods vehicles (
made up of 2 per cent oil-carrying and 3 per cent
non-oil-carrying), 4 per cent buses. Regrouping, it
was found that 29 per cent of the total traffic was
heavy commercial vehicles. Total night traffic did
not exceed 30 per cent of the total traffic. It was
noted that during the traffic enumeration the ban
on night driving of heavy vehicles was in force.
Matatus were found to be four times the number of

buses. Matatus were found to be making more frequent trips than buses plying the same routes.

Peak hour traffic varied at different sites generally recurring between noon and 4.00 p.m. along the whole study road. Due to the night ban on heavy traffic they were almost non-existent except those carrying essential services. Traffic varied only slightly during the weekdays. Sunday had the lowest traffic volume, particularly at Naro Moru. From information on origin and destination it was observed that of all the heavy traffic at Nanyuki, 75 per cent is through traffic. At Kiganjo it was found to be as high as 95 per cent. At Nanyuki 40 per cent of the through traffic had their origin/destination as Meru, with significant amounts going as far north as Isiolo and Marsabit. At Kiganjo, 30 per cent of the through traffic had its origin and destination as Meru, Isiolo and Marsabit. South bound heavy traffic (i.e. Nanyuki to Kiganjo) was found to be 78 per cent and 55 per cent of all through heavy traffic at Kiganjo and Nanyuki respectively.

Using equations 3.60 to 3.63 and a six-ordinate scheme of the harmonic analysis predictive models for traffic flow on the single carriageway were developed for both smoothed and unsmoothed data. The predictive model for the observed un-

smoothed traffic flow data was developed as (Fig.4.9)

$$q_{p_1} = 64.167 - 64.667 \cos t - 0.667 \cos 2t + 8.167 \cos 3t$$

$$-24-249 \sin t + 6.928 \sin 2t \qquad (4.12)$$

$$40 < q < 100$$

where, q_{p_1} is the predicted number of vehicles at time t. The comparison of observed and predicted data gave the equation

$$q_o = 1.019 \, q_{p_1} + 0.354$$

with r = 0.94, r^2 = 0.88 and standard error of 51 where, q_o is the observed number of vehicles at time t. Similarly, the equation for the smoothed data was developed as

$$q_{p_2} = 65.333 - 63.5 \cos t - 2.833 \cos 2t + 11 \cos 3t$$

$$-28.001 \sin t + 8.372 \sin 2t \qquad (4.13)$$

$$40 < q < 100$$

where, q_{p_2} is the predicted traffic flow at time t. The comparison of observed and predicted data yielded the equation

$$q_o = 1.011 \, q_{p_2} - 0.217$$

with r = 0.94, r^2 = 0.89 and standard error of 51

PREDICTION OF TRAFFIC FLOW PER HOUR

$$q_{P_1} = 64.167 - 64.667\cos t - 0.667\cos 2t + 8.167\cos 3t - 24.249\sin t + 6.928\sin 2t$$

$$q_{P_2} = 65.333 - 63.5\cos t - 2.833\cos 2t + 11\cos 3t - 28.001\sin t + 8.372\sin 2t$$

● UNSMOOTHED DATA   ○ SMOOTHED DATA

VEHICLES (q)

TIME (t) (HOURS)

COMPARISON OF OBSERVED AND PREDICTED DATA

$$q_O = 1.011 q_{P_2} - 0.217$$

$$q_O = 1.019 q_{P_1} + 0.354$$

OBSERVED q

PREDICTED q

FIG. 4.9 PREDICTION OF AVERAGE TRAFFIC FLOW PER HOUR: KIGANJO - NANYUKI ROAD

where q_0 is as above. To reduce the standard error

larger ordinate schemes than six can be used. The

slopes of the regression equations obtained by

comparing the predicted and observed data are very

nearly ideal (1.019 and 1.011). Their corresponding

intercepts are very nearly zero (0.354 and - 0.217).

Thus for low trafficked rural roads with little

flactuations in the traffic flow data smoothing does

not affect the prediction models developed. Both

models are quite acceptable for prediction purposes.

4.2   Road Traffic Accidents Data Collection and Analysis

4.2.1   National Road Traffic Accidents

4.2.1.1   Data Collection

National RTAs statistics were collected and

extracted from the Kenya Traffic Police records and

supplemented with those published in the annual

Statistical Abstracts prepared and published by the

CBSK. The data covered the period 1949-83. For

each year the data gathered included: injury RTAs;

road fatalities; drivers, motorcyclists, pedal cyclists

passengers and pedestrians (killed, seriously and

slightly injured); RTAs victims below/above age 16;

RTAs distribution by day/night; responsibility for

RTAs by class of road user (Appendix A.13).

4.2.1.2 Data Analysis

In order to develop mathematical models each plot of data was smoothed using the technique of moving averages. The curve shape revealed by the smoothed data was then determined. Using the smoothed data as the model data and choosing an appropriate function from Chapter 3 the predictive model, was developed and tested by the method of analysis of comparing observed and predicted data as outlined in 3.5.2. Finally the level of significance for each relationship was determined.

Injury Road Traffic Accidents

For injury RTAs growth models for RTAs and RTAs per 10^6 vehicle-kilometres were developed. Models predicting RTAs and RTAs per vehicle as functions of motorization were also derived. For injury RTAs the growth model was developed as a logistic curve which the shape of the smoothed data suggested (Fig.4.10). It was necessary to choose a limit for this model and the figure of 8049 which was the maximum observed was used for limit approximation. The model developed then was (using equation 3.15)

$$A_{p_1} = \frac{8049}{1 + 1.984\ e^{-0.137t}} \qquad (4.14)$$

where, A_{p_1} is predicted injury RTAs at year t. The comparison of observed and predicted data yielded the equation

$$A_o = 0.946 A_{p_1} + 399$$

where A_o is observed injury RTAs, with r= 0.91, $r^2 = 0.83$ and standard error of 1394. Smoothed data of the number of RTAs were plotted against motoriza-tion. Using the finite differences technique for equally spaced data the data suggested a polynomial curve of third degree (Fig.4.11). Using formulae in 8.26 the predictive model was developed as

$$A_{p_2} = 6.6981\text{x}10^{-8}(V/p)^3 - 0.82263376(V/p)^2$$

$$+ \quad 270.438684(V/p + 15342.4929 \qquad (4.15)$$

where, A_{p_2} is the number of injury RTAs and V/p is vehicles per 10^4 persons. The comparison between observed injury RTAs (A_o) and the predicted number of injury RTAs yielded the equation

$$A_o = 1.135 A_{p_2} - 605$$

with r = 0.92, $r^2 = 0.85$ and stnadard error of 1259. The predictions by the two models (4.14, 4.15) are quite consistent ($r^2 = 0.83$, 0.85 respectively) but model 4.14 predicts injury RTAs closer to the

INJURY RTAs AND MOTORIZATION IN KENYA

$$A_{p_2} = 6.6981 \times 10^{-6}(V/P)^3 - 0.82263376(V/P)^2 + 270.438684 V/P - 15342.4929$$

VEHICLES PER 10^4 PERSONS (V/P)

- •OBSERVED UNSMOOTHED DATA
- ○SMOOTHED DATA

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COMPARISON OF OBSERVED
AND PREDICTED DATA

$$A_o = 1.135 A_{p_2} - 505$$

PREDICTED RTAs

FIG 4·11   RELATION BETWEEN RTAs AND MOTORIZATION

observed values since the slope of the regression
equation (0.946) is closer to the ideal (1) than
1.135 and the intercept of 399 is closer to the
origin (0) than -605. This implies that the
polynomial model (4.15) tends to under-predict injury
RTA when compared with the logistic model (4.14).
Models 4.14 and 4.15 were found to be significant
at the 5 per cent level.

Injury RTAs per Vehicle

Injury RTAs per vehicle were plotted against
motorization. After data smoothing and testing for
polynomial fit the predictive model (Fig.4.12)

$$(A/V)_p = 0.656 \times 10^{-12} (V/P)^3 - 0.000\ 00484 (V/P)^2$$

$$+ 0.001059\ V/P - 0.0183 \qquad (4.16)$$

was developed where, $(A/V)_p$ is the number of injury
RTAs per vehicle predicted and V/P is degree of
motorization. The comparison of observed and
predicted data yielded the equation

$$(A/V)_o = 1.246 (A/V)_p - 0.0092$$

where, $(A/V)_o$ is observed injury RTAs per vehicle,
with $r = 0.80$, $r^2 = 0.64$ and standard error of
0.0034. The regression equation of observed values
compared with predicted values indicates good
correlation ($r=0.80$) but the model 4.16 explains only

RTAs PER VEHICLE AND MOTORIZATION

$$(A/V)_{p_2} = -0.0183 + 0.0011 \, V/P - 4.84 \times 10^{-6} (V/P)^2 + 0.656 \times 10^{-12} (V/P)^3$$

COMPARISON OF OBSERVED AND PREDICTED DATA

$$(A/V)_0 = 1.246 (A/V)_{p_2} - 0.009$$

FIG.4.12 RELATION BETWEEN RTAs PER VEHICLE AND MOTORIZATION

64 per cent of the variation in data. The regression
equation reveals that the intercept (-0.0092) is very
near the origin but from the slope (1.246) model
4.16 has a tendency of under-predicting. Model
4.16 was found to be significant at 5 per cent level.

Injury RTAs per 10^6 Vehicle-Kilometres

Predictive models were also developed along
the same lines for injury RTAs per 10^6 vehicle-
kilometres. The growth model was developed as
(Fig.4.13)

$$(A/K)_{p_1} = \frac{1.833}{1 + 0.085 \; e^{0.093t}} \qquad (4.17)$$

where, $(A/K)_{p_1}$ is the predicted RTAs per 10^6 vehicle-
kilometres at year t and 1.833 the highest obser-
ved RTAs per 10^6 vehicle-kilometres used here as an
approximation of the upper limit in the logistic
curve model of equation 3.15 . The comparison of
observed data $(A/K)_o$ and the predicted data by
equation 4.17 was found to be related by the
equation

$$(A/K)_o = 0.890 \; (A/K)_{p_1} + 0.168$$

with $r = 0.82$, $r^2 = 0.68$ and standard error of
0.195. The polynomial fit between injury RTAs per
10^6 vehicle-kilometres and motorization was deve-
loped as

GROWTH OF INJURY RTAs PER VEHICLE KILOMETRE

$$(A/K)_p = \frac{1.833}{1 + 0.085e^{0.093t}}$$

COMPARISON OF OBSERVED
AND PREDICTED DATA

$$(A/K)_o = 0.890(A/K)_p + 0.168$$

FIG. 4.13 GROWTH OF INJURY RTAs PER 10^6 VEHICLE-KILOMETRES IN KENYA

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$$(A/K)_{P_2} = 0.2728 \times 10^{-10} (V/P)^3 - 0.00019613 (V/P)^2$$

$$+ 0.042285 (V/P) - 0.6774 \qquad (4.18)$$

where $(A/K)_{P_2}$ is RTAs per 10^6 vehicle-kilometres predicted as a function of motorization. The comparison of observed data and predicted data by equation 4.18 yielded the equation

$$(A/K)_o = 0.956 (A/K)_{P_2} + 0.082$$

with $r = 0.76$, $r^2 = 0.58$ and standard error of 0.246. The two models 4.17 and 4.18 were found to be significant at the 5 per cent level. The regression equations for the predictions obtained by models 4.17 and 4.18 indicate that the predictions are close to the observed since the slopes 0.890 and 0.956 are close to unity and the intercepts 0.168 and 0.082 are close to origin (0). However, model 4.18 gives better predictions than model 4.17.

RTA Casualties

RTAs casualties are defined as the sum of road deaths and road injuries (serious plus slight). Predictive models were developed based on road casualties in Kenya for the period 1960-83 for which data was available. After data smoothing and plotting of casualties against time (years)

and using equation (3.15) the predictive model for
casualties was developed as (Fig.4.14)

$$C_{p_1} = \frac{14749}{1 + 3.772\ e^{-0.137t}} \qquad (4.19)$$

where, C_{p_1} is the predicted number of casualties at
time t and 14749 was the highest level of casualties
observed used here as an approximation of the limit.
The comparison between observed data (C_o) and
predicted data by model 4.19 gave the equation

$$C_o = 1.014\ C_{p_1} + 204$$

with $r = 0.94$, $r^2 = 0.88$ and standard error of
3137. Using formulae 3.26 and plynomial fitting
techniques as before the polynomial function pre-
dicting casualties as a function of motorization was
determined as

$$C_{p_2} = 0.1162 \times 10^{-6} (V/P)^3 - 1.10706866 (V/P)^2$$

$$+ 414.989047\ ^{V/}P - 26131.0196 \qquad (4.20)$$

where C_{p_2} is the predicted value of casualties at a
given level of motorization (V/P). Observed and
predicted values by equation 4.20 were compared
and yielded the equation

$$C_o = 1.090\ C_{p_2} - 733$$

GROWTH IN RTA CASUALTIES IN KENYA

$$C_{p_1} = \frac{14749}{1 + 3.772e^{-0.137t}}$$

$$C_0 = 1.014\,C_{p_1} + 204$$

COMPARISON OF OBSERVED AND PREDICTED DATA

FIG 4.14 GROWTH IN RTA CASUALTIES IN KENYA

with $r = 0.93$, $r^2 = 0.86$ and standard error of 2690.
Models 4.19 and 4.20 were found to be significant at
the 5 per cent level. Considering the slopes
of the regression equations 1.014 and 1.090, the
predictions by the two models are very close to the
observed values. However, model 4.19 has a better
calibration than model 4.20 considering that the
intercept 204 is closer to the origin than -733.
Both models 4.19 and 4.20 have very consistent
predictions considering the correlation and deter-
mination coefficients. Therefore, both models
are quite acceptable in predicting RTA casualties.

Casualties per Road Traffic Accident

Casualties per RTA (C/A) were plotted against
time (Fig.4.15), after data smoothing and using
equation 3.15 the predictive model was developed
as follows:

$$(C/A)_{P_1} = \frac{1.8}{1 + 0.584\ e^{-0.095t}} \tag{4.21}$$

where, 1.8 is the highest observed value used as the
approximate limit, $(C/A)_{P_1}$ is the predicted value of
casualties per RTA at time (year) t using equation
4.21 . Comparing observed values ($(C/A)_o$) and
predicted values the equation below was obtained

GROWTH IN CASUALTIES PER ACCIDENT IN KENYA

$$C/A_1 = \frac{1.800}{1 + 0.584\,e^{-0.095t}}$$

● OBSERVED UNSMOOTHED DATA
○ SMOOTHED DATA

COMPARISON OF OBSERVED AND
AND PREDICTED DATA

$(C/A)_0 = 1.219(C/A)_p - 0.304$

FIG. 4.15 GROWTH IN CASUALTIES PER ACCIDENT IN KENYA

$$(C/A)_o = 1.219 \, (C/A)_{P_1} - 0.304$$

with r = 0.87, r^2 = 0.75 and standard error of 0.159
The corresponding polynomial model was developed as
(Fig.4.16)

$$(C/A)_{P_2} = 0.11831 \times 10^{-10} (V/P)^3 - 0.00015 (V/P)^2$$

$$+ 0.04492 \, V/P - 1.7667 \tag{4.22}$$

where, $(C/A)_{P_2}$ is predicted casualties per RTA at
a given level of motorization (V/P). The observed
data were compared with predicted data and yielded
the equation

$$(C/A)_o = 1.355 (C/A)_{P_2} - 0.488$$

with r = 0.84, r^2 = 0.71 and standard error of 0.146.
Models 4.21 and 4.22 were found to be significant
at the 5 per cent level. Considering the coeffi-
cients of correlation and determination for the
regression equations for the models 4.21 and 4.22
the predictions are reasonably consistent (r = 0.87,
0.84; r^2 = 0.75, 0.71). However considering the
slopes (1.219, 1.355) which are greater than 1 and
the intercepts (-0.304, - 0.488) the models have a
tendency of over-predicting casualties per RTA
initially at low levels and under-predicting at high
values. Within the range of the observed data the

CASUALTIES PER RTA AND MOTORIZATION

$$(C/A)_{p_2} = 0.11831 \times 10^{-10}(V/P)^3 - 0.00015(V/P)^2 + 0.04492 V/P - 1.7667$$

C/A — CASUALTIES PER RTA

2.0

1.0

0

○ SMOOTHED DATA

90   100   120   140   160   180   200   220   240   260   280

VEHICLES PER 10^4 PERSONS   V/P

139

OBSERVED CASUALTIES PER RTA

2.0

1.5

1.0

0.5

0

0   0.5   1.0   1.5   2.0

PREDICTED CASUALTIES PER RTA

COMPARISON OF OBSERVED AND

PREDICTED DATA

$$(C/A)_0 = 1.355(C/A)_{p_2} - 0.488$$

FIG. 4.16 RELATION BETWEEN CASUALTIES PER RTA AND MOTORIZATION

predictions are quite close to the observed values.

Casualties per 10^4 Vehicles

Similarly for casualties per 10^4 vehicles
(C/V) the growth model using equation 3.15 was
developed as (Fig.4.17)

$$(C/V)_{P_1} = \frac{713.67}{1 + 0.964\ e^{-0.048t}} \qquad (4.23)$$

where, 713.67 is the highest value of C/V observed
used here to approximate the limit and $(C/V)_{P_1}$ is the
predicted value by equation 4.23 . Comparing
observed data $((C/V)_o)$ and predicted data yielded
the equation

$$(C/V)_o = 0.713\ (C/V)_{P_1} + 144.78$$

with $r = 0.52$, $r^2 = 0.27$ and standard error of
72.47. The polynomial model by similar techniques
as afore-mentioned was determined as (Fig.4.18)

$$(C/V)_{P_2} = 8.8191 \times 10^{-8} (V/P)^3 - 0.04464166 (V/P)^2$$

$$+ 12.597057\ V/_P - 337.6169 \qquad (4.24)$$

where, $(C/V)_{P_2}$ is the predicted C/V as a function
of motorization (V/P). Using equation 4.24
observed data compared with predicted data

C/V

GROWTH OF CASUALTIES PER 10^4 VEHICLES IN KENYA

$$(C/V)_{P_1} = \frac{713 \cdot 67}{1 + 0.964 e^{-0.048 t}}$$

o SMOOTHED DATA

COMPARISON OF OBSERVED
AND PREDICTED

$(C/V)_0 = 0.713 (C/V)_{P_1} + 145$

FIG 4.17 GROWTH OF CASUALTIES PER 10^4 VEHICLES IN KENYA

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CAUALTIES PER 10^4 VEHICLES AND MOTORIZATION

(Top graph labels)
1000 C/V
800
CASUALTIES PER 10^4 VEHICLES
600
400
200
0

o SMOOTHED DATA

$(C/V)_{P_2} = -337.613$

60 80 100 120 140 160 180 200
VEHICLES

COMPARISON OF OBSERVED AND PREDICTED DATA

$(C/V)_0 = 1.022 (C/V)_{P_2} - 8.841$

OBSERVED C/V
1000
800
600
400
200
0

0 200 400 600 800 1000
PREDICTED C/V

FIG 4.18 RELATION BETWEEN CASUALTIES PER 10^4 VEHICLES AND MOTORIZATION

resulted in the equation

$$(C/v)_o = 1.022 \ (C/v)_{P_2} \ - \ 8.841$$

with r = 0.79, r^2 = 0.62 and standard error of 71.414. Models 4.23 and 4.24 were found to be significant at the 5 per cent level. Considering the slopes of the regression equations (0.713, 1.022) and their intercepts (144.78, -8.841), model 4.24 is better calibrated than model 4.23 since the slope is close to unity and the intercept is close to 0. Further, considering the coefficients of correlation and determination (r = 0.52, 0.79; r^2 = 0.27, 0.62) model 4.24 fairs better than model 4.23. Model 4.23 has much lower consistency, as the scatter is quite considerable, than model 4.24.

Road Traffic Accident Deaths

Predictive models were also developed based on road deaths on Kenyan roads for the period 1949-83. After data smoothing and plotting of deaths against time (years) and using equation 3.15 the predictive model for deaths was developed as (Fig.4.19)

$$D_{P_1} = \frac{1720}{1 + 22.889 \ e^{-0.174t}} \tag{4.25}$$

GROWTH OF RTA DEATHS IN KENYA

$$D_{p_1} = \frac{1720}{1 + 22.889e^{-0.174t}}$$

• OBSERVED UNSMOOTHED DATA
○ SMOOTHED DATA

COMPARISON OF OBSERVED
AND PREDICTED DATA

$$D_0 = 0.994\, D_{p_1} - 22$$

FIG 4.19 GROWTH OF RTA DEATHS IN KENYA

where, D_{p_1} is the number of predicted deaths at time

t and 1720 was the highest level of deaths observed

during the period under study, used here as an

approximation of the limit. A higher value (like

2000) than 1720 would alter the limit of saturation

and increase the rate of growth of RTA deaths with

time. This was tried and the model obtained was

used to predict RTA deaths which were compared with

the observed value. It was found that the resulting

regression equation did not significantly differ

from the one obtained using 1720 as the upper limit.

The comparison between observed data (D_o) and pre-

dicted data by model 4.25 gave the equation

$$D_o = 0.994 \, D_{p_1} - 22$$

with $r = 0.97$, $r^2 = 0.95$ and standard error of 553.
Using formulae 3.26 and polynomial fitting techniques

as before the polynomial function predicting deaths as

a function of motorization was developed as (Fig.4.20)

$$D_{p_2} = 229.7235 - 11.960418 \, V/P + 0.14040617 (V/P)^2$$

$$+ 0.18958 \times 10^{-6} (V/P)^3 \qquad (4.26)$$

where D_{p_2} is the number of predicted deaths at a

given degree of motorization V/P. Comparing obser-

ved data against predicted data yielded the equation

$$D_o = 1.009 \, D_{p_2} - 10$$

with $r = 0.97$, $r^2 = 0.94$ and standard error of 545. Models 4.25 and 4.26 were found to be significant at the 5 per cent level. Considering the coefficients of correlation and determination of the regression equations ($r = 0.97$; $r^2 = 0.95$, 0.94) the predictions obtained by the two models are very consistent. Considering the slopes (0.994, 1.009) and the intercepts (-22, - 10) the calibration of the two models are very near perfect since these parameters are very close to 1 and 0 respectively. Both models are therefore acceptable as the predicted values are almost identical to the observed values.

RTA Deaths per 10^4 Persons

Road deaths per 10^4 persons (D/P) were smoothed and plotted against motorization (Fig.4.21). Using polynomial function fitting techniques as above the predictive model was developed as

$$(D/P)_{P_1} = 1.9788 \times 10^{-10} (V/P)^3 + 0.00001494 (V/P)^2$$

$$+ 0.007762 (V/P) - 0.4127 \qquad (4.27)$$

where $(D/P)_{P_1}$ is the predicted number of road deaths per 10^4 persons at a given level of motorization (V/P). Observed data $(D/P)_o$ was compared with the predicted data by equation (4.27) yielding

RTA DEATHS AND MOTORIZATION

$$D_{P_2} = 229.724 - 11.96 \, V/P + 0.14 \, (V/P)^2 + 0.19 \times 10^{-6} (V/P)^3$$

o SMOOTHED DATA

DEATHS (D)

VEHICLES/10^4 PERSONS (V/P)

COMPARISON BY AGOKI FORMULA

$$D_0 = 1.009 D_{P_2} - 10$$

OBSERVED D

PREDICTED D

FIG.4.20 RELATION BETWEEN RTA DEATHS AND MOTORIZATION

equation

$$(D/P)_o = 0.923(D/P)_{P_1} + 0.054$$

with $r = 0.93$, $r^2 = 0.87$ and standard error of 0.31.
Model 4.27 was found to be significant at the 5 per
cent level. The regression of predicted values
against observed values showed a consistent prediction
and strong correlation ($r = 0.93$, $r^2 = 0.87$). The
slope (0.923) and intercept (0.054) indicate slight
under-prediction for low ranges and slight over-
prediction for high ranges. The predictions,
nonetheless, are close to the observed values and
quite acceptable as the slope is close to 1 and the
intercept nearly 0.

RTA Deaths per 10^4 Vehicles

Further, road deaths per 10^4 vehicles (D/V)
were smoothed and plotted against motorization
(Fig.4.22). Again, using the polynomial function
fitting techniques as before the predictive model
was developed as

$$(D/V)_{P_1} = 1.4218 \times 10^{-8}(V/P)^3 - 0.00022871(V/P)^2$$

$$+ 0.577548 \, V/P - 5.4171 \qquad (4.28)$$

where $(D/V)_{P_1}$ is the predicted number of deaths
per 10^4 vehicles for a given level of motorization.

FIG. 4.21 RELATION BETWEEN DEATHS/10^4 PERSONS AND MOTORIZATION

The observed data $(D/V)_o$ was compared with the predicted data using equation 4.28 resulting into the equation

$$(D/V)_o = 0.769 \ (D/V)_{p_1} + 13$$

with $r = 0.77$, $r^2 = 0.59$ and standard error of 15. Model 4.28 was found to be statistically significant at the 5 per cent level. Considering the regression equation above with coefficients of correlation and determination 0.77 and 0.59 respectively the pre-diction is not very consistent since it has some scatter. The slope (0.769) and the intercept (13) suggest that at low ranges there is under-prediction of data and at high ranges there is over-prediction. This suggests need for further calibration. Comparing the predictions with those obtained by Jacobs and Smeed model 4.28 is a better fit for Kenyan data (Fig.4.22) and therefore quite acceptable.

RTA Deaths per 10^6 Vehicle-Kilometres

Similarly predictive models for deaths per 10^6 vehicle-kilometres (D/K) were developed. Firstly, a growth model was sought. Therefore after data smoothing and plotting (Fig.4.23) and using equation 3.15 the predictive growth model was developed as

FIG. 4.22 RELATION BETWEEN RTA DEATHS PER 10^4 VEHICLES AND MOTORIZATION

$$(D/K)_{p_1} = \frac{0.3415}{1 + 1.9986\ e^{-0.093t}} \qquad (4.29)$$

where, $(D/K)_{p_1}$ is the predicted number of road deaths
by the logistic curve model per 10^6 vehicle-kilometres
at time t and 0.3415 was the highest observed D/K
used here to approximate the limit. Comparing obser-
ved data $(D/K)_o$ against predicted data gave the
equation

$$(D/K)_o = 0.864\,(D/K)_{p_1} + 0.032$$

with $r = 0.84$, $r^2 = 0.71$ and standard error 0.060.
In order to relate road deaths per 10^6 vehicle-
kilometres smoothed data was plotted (Fig.4.24) and
using the polynomial function fitting techniques as
before the predictive model was developed as

$$(D/K)_{p_2} = 0.58935 \times 10^{-10}(V/p)^3 - 0.00000131(V/p)^2$$
$$+ 0.00241 V/p - 0.0273 \qquad (4.30)$$

where, $(D/K)_{p_2}$ is predicted data using equation
4.30 at a given level of motorization. Comparison
between observed data and predicted data yielded
the equation

$$(D/K)_o = 0.763\ (D/K)_{p_2} + 0.053$$

GROWTH IN RTA DEATHS PER 10^6 VEHICLE-KILOMETRES IN KENYA

$$(D/K)_{P_1} = \frac{0.3415}{1 + 1.9986e^{-0.093\,t}}$$

○ SMOOTHED DATA

DEATHS PER 10^6 VEH-KMs (D/K)

YEARS (t)

COMPARISON OF OBSERVED
AND PREDICTED DATA

OBSERVED D/K

$(D/K)_0 = 0.864\,(D/K)_{P_1} + 0.0318$

PREDICTED D/K

FIG. 4 23 GROWTH IN DEATHS PER 10^6 VEH-KMs IN KENYA

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with $r = 0.77$, $r^2 = 0.59$ and standard error of 0.059. The predictive models 4.29 and 4.30 were found to be significant at the 5 per cent level. Considering the coefficients of correlation and determination ($r = 0.84$, 0.77; $r^2 = 0.71$, 0.59) for the two regression equations above the predictions are fairly consistent although there is some scatter. The slopes (0.864, 0.763) and the intercepts (0.032, 0.053) suggest that at low ranges the values of D/K are under-predicted whilst at high ranges the values of D/K are over-predicted. However, since the slopes are reasonably close to 1 and the intercepts sufficiently close to 0 the predictions are acceptable for the data fitted. The slopes and intercepts suggest a possibility of improving the calibration of the models. Model 4.29 makes better prediction than model 4.30.

RTA Injuries

Predictive models were also developed based on RTAs injuries on Kenyan roads for the period 1949-83. After data smoothing and plotting of injuries against time (years) and using model equation 3.15 the predictive growth model for injuries was developed as (Fig.4.25)

$$I_{p_1} = \frac{13526}{1 + 12.095\, e^{-0.122t}} \qquad (4.31)$$

FIG 4.24 RELATION BETWEEN DEATHS PER 10^6 VEH-KMs AND MOTORIZATION

where, I_{P_1} is the predicted number of injuries at
time t and 13526 was the highest level of injuries
observed, used here as an approximation of the limit.
The comparison between observed data I_o and predicted
data as obtained by using model 4.31 yielded

$$I_o = 1.045\ I_{P_1} - 171$$

with r = 0.97, r^2 = 0.93 and standard error of 3430.
Using formulae 3.26 and the polynomial fitting
techniques as before the polynomial function pre-
dicting injuries as a function of motorization was
developed as

$$I_{P_2} = 1.2305 \times 10^{-6}(V/_P)^3 + 0.25629308(V/_P)^2$$

$$+ 60.580992 V/_P - 4849.778 \qquad (4.32)$$

where I_{P_2} is the predicted number of injuries at a
given level of motorization. Comparing observed
data against predicted data gave the equation

$$I_o = 0.961\ I_{P_2} + 344$$

with r = 0.94, r^2 = 0.89 and standard error of 3144.
Models 4.31 and 4.32 were found to be significant
at the 5 per cent level. Considering the coeffi-
cients of correlation and determination for the

regression equations above (r=0.97, 0.94; r^2 = 0.93, 0.89) the consistency in the predictions is very good. Further, the slopes (1.045, 0.961) and the intercepts (-171, 344) suggest that the calibration of the two models is good since the two parameters are sufficiently close to 1 and 0 respectively. The models are therefore quite acceptable for the data fitted.

RTA Injuries per 10^4 Persons

Then injuries per 10^4 persons (I/P) were smoothed and plotted (Fig.4.26) against time (years) and using model equation 3.15 the predictive growth model for injuries per 10^4 persons was developed as

$$(I/P)_{P_1} = \frac{8.483}{1 + 2.788\ e^{-0.086t}} \qquad (4.33)$$

where, $(I/P)_{P_1}$ is the predicted number of injuries at time t and 8.483 was the highest level of injuries per 10^4 persons observed, used here as an approximation of the limit. The comparison between observed data $(I/p)_o$ and predicted data as obtained by using model (4.33) yielded

$$(I/p)_o = 1.047(I/P)_{P_1} - 0.149$$

with r = 0.91, r^2 = 0.83 and standard error of 1.509. Using formulae (3.26) and the polynomial function

GROWTH IN RTA INJURIES IN KENYA

$$I_{p_1} = \frac{13526}{1 + 12.095e^{-0.122t}}$$

$$I_o = 1.045 I_{p_1} - 171$$

COMPARISON OF OBSERVED AND

PREDICTED DATA

FIG 4.25   GROWTH IN RTA INJURIES IN KENYA

fitting techniques the predictive model relating injuries per 10^4 persons and motorization (Fig.4.27) was developed as

$$(I/P)_{P_2} = 9.869 \times 10^{-10} (V/_P)^3 - 5.156 \times 10^{-4} (V/_P)^2$$

$$+ 0.174566 V/P - 7.9035 \qquad (4.34)$$

where, $(I/P)_{P_2}$ is the predicted number of injuries per 10^4 persons using model (4.34). Comparing observed and predicted data yielded the equation

$$(I/P)_o = 0.885 (I/P)_{P_2} + 0.764$$

with $r = 0.87$, $r^2 = 0.76$ and standard error of 1.742. Models 4.33 and 4.34 were found to be significant at the 5 per cent level. The coefficients of correlation and determination for the regression equations above indicate strong correlation between predicted and observed values as well as consistency in prediction. The slopes (1.047, 0885) and the intercepts (-0.149, 0.764) imply good calibration of the models. Model 4.33 however, is closer to the ideal (slope = 1, intercept = 0) than model 4.34 which tends to under-predict at low ranges and over-predict at high ranges of injuries per 10^4 persons.

GROWTH IN RTA INJURIES PER 10^4 PERSONS IN KENYA

$$(I/P)_R = \frac{8 \cdot 483}{1 + 2 \cdot 788 e^{-0 \cdot 085 t}}$$

$$(I/P)_O = 1 \cdot 047 (I/P)_R - 0 \cdot 149$$

COMPARISON OF OBSERVED AND
PREDICTED DATA

FIG 4 26 GROWTH IN RTA INJURIES PER 10^4 PERSONS IN KENYA

Injuries per 10^4 Vehicles

Similarly, a predictive growth model relating injuries per 10^4 vehicles (I/V) to time (years) was developed as (Fig.4.28)

$$(I/V)_{p_1} = \frac{664}{1 + 0.916 \ e^{-0.035t}} \qquad (4.35)$$

where, $(O/V)_{p_1}$ is the predicted number of injuries per 10^4 vehicles using model 4.34 . The observed data $(I/V)_o$ was compared with the predicted to yield the equation

$$(I/V)_o = 0.734(I/V)_{p_1} + 117$$

with $r = 0.44$, $r^2 = 0.19$ and standard error of 54. By polynomial function fitting techniques the predictive model relating injuries per 10^4 vehicles and motorization (Fig.4.29) was developed as

$$(I/V)_{p_2} = 6.0845 \times 10^{-8} (V/P)^3 - 0.0471716(V/P)^2$$

$$+ 12.4842 \ V/P - 337.0343 \qquad (4.36)$$

where, $(I/V)_{p_2}$ is the predicted number of injuries per 10^4 vehicles using model (4.36). Comparing observed and predicted data gave the equation

$$(I/V)_o = 1.103 \ (I/P)_{p_2} - 55$$

$$(I/P)_{p_2} = 9.8691 \times 10^{-10}(V/P)^3 - 0.00051556(V/P)^2 + 0.1745$$

o SMOOTHED DATA

$$(I/P)_0 = 0.885(I/P)_{p_2} + 0.764$$

COMPARISON OF OBSERVED AND PREDICTED DATA

FIG.4.27 RELATION BETWEEN INJURIES PER 10^4 PERSONS AND MOTORIZATION

GROWTH IN RTA INJURIES PER 10^4 VEHICLES IN KENYA

$$(I/V)_{P_1} = \frac{6\,64}{1 + 0.916e^{-0.035t}}$$

COMPARISON OF OBSERVED AND PREDICTED DATA

$$(I/V)_0 = 0.734(I/V)_{P_1} + 117$$

FIG 4.28 GROWTH IN RTA INJURIES PER 10^4 VEHICLES IN KENYA

with $r = 0.78$, $r^2 = 0.60$ and standard error of 56.
Models 4.35 and 4.36 were found to be significant
at the 5 per cent level. The regression equation
obtained by using model 4.35 for prediction shows
that the prediction is not very consistent as the
scatter is very considerable and correlation,
between observed and predicted values, weak
($r = 0.44$, $r^2 = 0.19$). However, the slope (0.734)
and the intercept (117 are fair indicating a fair
calibration. Model 4.36 has a better calibration
than model 4.35 considering that the slope (1.103)
is much nearer to 1 and the intercept (-55) is much
closer to 0 than the corresponding parameters for
model 4.35. Therefore, model 4.36 is more acceptable
for predicting injuries per 10^4 vehicles than model
4.35.

Injuries per 10^6 Vehicle-Kilometres

Injuries per 10^6 vehicle-kilometres
were smoothed, plotted (Fig.4.30) and modelled
using equation 3.26 yielding the predictive growth
model

$$(I/K)_{P_1} = \frac{2.657}{1 + 0.9897\ e^{-0.042t}} \qquad (4.37)$$

where, $(I/K)_{P_1}$ is the predicted number of injuries
per 10^6 vehicle-kilometres, 2.657 is the limit

RTA INJURIES PER 10^4 VEHICLES AND MOTORIZATION

o SMOOTHED DATA

$$(I/V)_{P_2} = -337.0343 + 12.4842 \, V/P - 0.047172 \, (V/P)^2 + 6.0845 \times 10^{-8} \, (V/P)^3$$

VEHICLES PER 10^4 PERSONS V/P

COMPARISON OF OBSERVED
AND PREDICTED DATA

$$(I/V)_0 = 1.103 \, (I/V)_{P_2} - 55$$

PREDICTED I/V

FIG 4.29 RELATION BETWEEN RTA INJURES PER 10^4 VEHICLES AND MOTORIZATION

approximated from the observed highest number of injuries per 10^6 vehicle-kilometres. The observed data $(I/K)_o$ was compared with the predicted data and yielded the equation

$$(I/K)_o = 0.924 \ (I/K)_{P_1} + 0.084$$

with $r = 0.69$, $r^2 = 0.47$ and standard error of 0.239. In order to relate injuries per 10^6 vehicle-kilometres to motorization, smoothed data was plotted (Fig.4.31) and modelled using polynomial curve fitting techniques as before yielding the equation

$$(I/K)_{P_2} = 2.2162 \times 10^{-10} (V/P)^3 - -0.00014687 (V/P)^2$$

$$+ \ 0.041751 (V/P) - 1.0039 \qquad (4.38)$$

where, $(I/K)_{P_2}$ is the predicted number of injuries per 10^6 vehicle-kilometres using equation 4.38 . Comparing observed and predicted data yielded the equation

$$(I/K)_o = 1.020 \ (I/K)_{P_2} - 0.052$$

with $r = 0.77$, $r^2 = 0.59$ and standard error of 0.242. Models 4.37 and 4.38 were found to be significant at the 5 per cent level. The models show a fair degree of consistency in their prediction ($r = 0.69$, 0.77; $r^2 = 0.47$, 0.59). There is however, some scatter in the plot of predicted values against

GROWTH IN INJURIES PER 10^6 VEHICLE-KILOMETRES IN KENYA

$$(I/K)_{P_1} = \frac{2.657}{1 + 0.9897e^{-0.042t}}$$

COMPARISON OF OBSERVED AND PREDICTED DATA

$$(I/K)_0 = 0.924(I/K)_{P_1} + 0.084$$

G 4.30   GROWTH IN INJURIES PER 10^6 VEH-KMs IN KENYA

observed data. Considering the slopes (0.924, 1.020)
and the intercepts (0.084, - 0.052) of the regression
equations above there is indication that the two
models are well calibrated as the two parameters
are very close to the ideal values (1,0) respectively.
The models are quite acceptable therefore.

Severity Index

Severity index is defined as the ratio of
road deaths to casualties. This ratio was converted
to percentage, smoothed and plotted against time
(years) (Fig.4.32). Using equation 3.26 the
predictive growth model was developed as

$$\rho_{P_1} = \frac{16.1}{1 + 0.995 \, e^{-0.051t}} \qquad (4.39)$$

where, ρ_{P_1} is the predicted severity index at time
t using equation 4.49 . The observed data ρ_o
was compared with the predicted data and yielded the
equation

$$\rho_o = 0.643\rho_{P_1} + 8.9$$

with $r = 0.53$, $r^2 = 0.29$ and standard error of
1.65. Using smoothed data, plotting it against
motorization (Fig.4.33) and on the basis of
polynomial function fitting techniques the

$(I/K)_{P_2}$ = -1·0039 + 0·04175 V/P - 0·0001469 (V/P)^2 + 2·2162 × 10^{-10} (V/P)^3

○ SMOOTHED DATA

I/K

INJURIES /10⁶ VEH - KMs

VEHICLES PER 10⁴ PERSONS (V/P)

COMPARISON OF OBSERVED
AND PREDICTED DATA

$(I/K)_0$ = 1·020 $(I/K)_{P_2}$ + 0·052

OBSERVED (I/K)

PREDICTED (I/K)

FIG 4.31 RELATION BETWEEN INJURIES PER 10⁶ VEH-KMs AND MOTORIZATION

GROWTH IN SEVERITY INDEX IN RTAs IN KENYA

$$\rho_{p_1} = \frac{16.1}{1 + 0.885e^{-0.051t}}$$

COMPARISON OF OBSERVED AND PREDICTED DATA

$$\rho_0 = 0.643\rho_{p_1} + 3.9$$

FIG 4.32 GROWTH IN SEVERITY INDEX IN RTAs IN KENYA

predictive model developed was

$$\rho_{P_2} = 0.106 \times 10^{-8} (V/P)^3 + 0.00121676 (V/P)^2$$

$$- 0.2271 \ V/P + 20.8768 \qquad\qquad (4.40)$$

where, ρ_{P_2} is the predicted severity index using model
4.40 . Comparing observed and predicted data
yielded the equation

$$\rho_o = 0.833 \rho_{P_2} + 1.655$$

with $r = 0.63$, $r^2 = 0.40$ and standard error of 1.508.
Both predictive models 4.39 and 4.40 were found to
be statistically significant at the 5 per cent level.
The corresponding regression eqations above reveal
considerable scatter, of the plot of predicted values
against observed values ($r = 0.53$, 0.63; $r^2 = 0.29$,
0.40), making the correlation a rather weak one
particularly for model 4.39. The slopes 0.643, 0.833
of the regression equations and the intercepts
(3.9, 1.655) suggest under-prediction and over-pre-
diction of data, by the models, at low ranges and high
ranges respectively. However, for the range of data
used the models could be improved in calibration as
the slopes and intercepts suggest.

Further National RTA Characteristics

Further national RTA characteristics were

modelled relating to the percentage distribution of
RTAs by day and night, percentage responsibility
for RTAs, percentage distribution of those killed
above/below age 16, percentage distribution of
RTAs victims killed and injured. For each of
these a predictive growth model was sought. In
order to develop such models the available data,
which covered the periods 1960-83 for some of them
and 1973-83 for others,were smoothed using the
technique of moving averages with N = 5 years
(equation 3.57). These data were then plotted and
models developed using trend curve fitting techniques
by trying each of the following models: linear model
(3.19), exponential model (3.38), logarithmic model
(3.42) and the power model (3.46). The model that
best described the trends was the logarithmic model.
Logarithmic time series trend curves were then
developed accordingly for each of the above mentioned
characteristics.

RTAs Distribution by Day and Night

For the growth of percentage distribution of
RTAs by day and night in Kenya, the predictive model
developed was (Fig.4.34)

$$((\%)_{A_d p}) = 81.073 - 5.656 \ln t \qquad (4.41)$$

with r = -0.76, r^2 = 0.58, where $((\%)_{A_d p})$ is the

$$\rho_{P_2} = 20.8768 - 0.2271 V/P + 0.00122 (V/P)^2 + 0.106 \times 10^{-8} (V/P)^3$$

COMPARISON OF OBSERVED
AND PREDICTED DATA

$$\rho_0 = 0.833 \rho_{P_2} + 1.655$$

FIG 4.33 RELATION BETWEEN SEVERITY INDEX AND MOTORIZATION

predicted percentage at year t. The comparison of
observed data $((\%)_{A_d})_o$ and predicted data yielded
the equation

$$\left((\%)_{A_d} \right)_o \;=\; 0.754 \left((\%)_{A_d} \right)_p \;+\; 17.004$$

with $r = 0.45$, $r^2 = 0.20$ and standard error of
4.674. The relationship described by equation 4.41
was found to be statistically significant at the
5 per cent level. Considering the coefficients of
correlation and determination ($r = 0.45$, or$^2 = 0.20$)
the prediction is not very consistent and the
correlation is weak revealing considerable scatter.
The slope (0.754) and the intercept (17.004) suggest
under-prediction at low ranges and over-prediction
at high ranges. The calibration could therefore be
improved with further data observation over the years.

RTAs Responsibility

Predictive models relating to the growth in
percentage RTAs responsibility for classes of vehicle
types, road user categories or groups of these were
developed and tested for fitness of the trend curve
and statistical significance. In order to compare
percentage responsibility with percentage composition
for each category, the percentage composition data
had also been smoothed and plotted on the same
graph. For RTA responsibility by cars and utilities

the trend curve was developed as the model given by
the equation (Fig.4.35)

$$\left((\%)_{cu} \right)_p = 74.161 - 1.923 \ln t \qquad (4.42)$$

with r = -0.70, r^2 = 0.49, where $(\%)_{cu}{}_p$ is the
predicted percentage responsibility by cars and
utilities for year t as predicted by equation
4.42 . Comparing observed data $(\%)_{cu}{}_o$ and
predicted data yield the equation

$$\left((\%)_{cu} \right)_o = 0.887 \left((\%)_{cu} \right)_p + 14.661$$

with r = 0.43, r^2 = 0.18 and standard error of 1.59.
Equation 4.42 was found to be statistically
significant at 5 per cent level. The coefficients
of correlation (r = 0.43) and determination
$(r^2$ = 0.18) show that the prediction is not very
consistent and the correlation is weak due to the
considerable scatter. The slope (0.787) and the
intercept (14.661) indicate under-prediction at low
ranges and over-prediction at high ranges. The cali-
bration could be further improved with additional
data observation over the years

     For RTA responsibility by buses, lorries and
taxis the trend curve developed is represented by

GROWTH IN RTA DISTRIBUTION BY DAY AND NIGHT

$$((\%)_{A_d})_P = 81.073 - 5.656 \ln t$$

o SMOOTHED DATA

COMPARISON OF OBSERVED AND PREDICTED DATA

$$((\%)_{A_d})_O = 0.754((\%)_{A_d})_P + 17.004$$

FIG. 4.34 PERCENTAGE DISTRIBUTION OF RTAs BY DAY AND NIGHT IN KENYA

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the equation (Fig.4.36)

$$\left((\%)_{b\ell T'} \right)_p = 18.268 + 2.652 \, \ell n t \qquad (4.43)$$

with r= 0.78, r^2 = 0.60, where $(\%)_{b\ell T'_p}$ is

the predicted percentage responsibility by buses,

lorries and taxis by equation 4.43 . Comparing

observed data $\left((\%)_{b\ell T'} \right)_o$ and predicted data

yielded

$$\left((\%)_{b\ell T'} \right)_o = 0.767 \left((\%)_{b\ell T'} \right)_p + 5.894$$

with r = 0.53, r^2 = 0.29 and standard error of 2.19.

Model 4.43 was found to be statistically

significant at the 5 per cent level. The coeffi-

cients of correlation (r = 0.53) and determination

(r^2=0.29) reveal that the correlation between

predicted and observed values is fair but the

consistency of prediction is not very strong as

there is considerable scatter. The slope (0.767)

and the intercept (5.894) indicate need for further

calibration with additional data. Model 4.43 has a

tendency to under-predict at low ranges and to over-

predict at high ranges as seen from the slope and

intercept.

The growth in percentage responsibility of

GROWTH IN COMPOSITION AND RESPONSIBILITY FOR CARS AND UTILITIES

$$((\%)_{CU})_P = 76.427 - 2.160 \ln t$$

$$((\%)_{cu})_P = 74.161 - 1.923 \ln t$$

△ SMOOTHED DATA] COMPOSITION
○ SMOOTHED DATA] RESPONSIBILITY

$$((\%)_{CU})_0 = 0.994 ((\%)_{CU})_P + 0.489$$

COMPARISON OF OBSERVED AND
PREDICTED DATA

$$((\%)_{cu})_0 = 0.787 ((\%)_{cu})_P + 14.661$$

FIG.4.35 PERCENTAGE COMPOSITION AND RESPONSIBILITY FOR RTAs OF CARS AND UTILITIES IN KENYA

motorcycles was developed as the growth model
(Fig.4.37)

$$\left((\%)_m\right)_p = 6.956 - 0\text{-}552 \; \ell nt \qquad (4.44)$$

with $r = 0.042$, $r^2 = 0.18$ where $\left((\%)_m\right)_p$ is the
predicted percentage responsibility of motorcycles.
The comparison of observed $(\%)_{m_o}$ and predicted
data yielded the equation

$$\left((\%)_m\right)_o = 0.868 \left((\%)_m\right)_p + 0.727$$

with $r = 0.30$, $r^2 = 0.09$ and standard error of 0.453.
Model 4.44 was found to be statistically significant
at 10 per cent. The coefficients of correlation
($r = 0.30$) and determination ($r^2 = 0.09$) suggest
very weak correlation between predicted and observed
values, very considerable scatter and therefore lack
of consistency in prediction. However, the model
calibration (slope = 0.869, intercept = 0.727)
indicate that the predictions are close to ideal
(slope = 1, intercept = 0).

The growth model representing pedal cyclists
percentage responsibility (Fig.4.38) was developed
as the equation

$$\left((\%)_{b'}\right)_p = 11.748 - 2.003 \; \ell n \; t \qquad (4.45)$$

GROWTH IN COMPOSITION AND RESPONSIBILITY FOR BUSES, LORRIES AND TAXIS

$$((\%)_{blT'})_P = 18.268 + 2.652 \ln t$$

$$((\%)_{BLT})_P = 11.873 + 0.239 \ln t$$

$$((\%)_{blT'})_O = 0.767 ((\%)_{blT'})_P + 5.894$$

COMPARISON OF OBSERVED
AND PREDICTED DATA

$$((\%)_{BLT})_O = 1.158 ((\%)_{BLT})_P - 2.070$$

FIG 4.36 PERCENTAGE COMPOSITION AND RESPONSIBILITY FOR RTAs OF BUSES LORRIES TAXIS IN KENYA

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GROWTH IN COMPOSITION AND RESPONSIBILITY FOR MOTORCYCLES

$((\%)_M)_P = 5.323 + 0.309 \ln t$

$((\%)_m)_P = 6.956 - 0.552 \ln t$

$((\%)_M)_O = 1.123 ((\%)_M)_P - 0.705$

$((\%)_m)_O = 0.868 ((\%)_m)_P + 0.727$

COMPARISON OF OBSERVED
AND PREDICTED DATA

FIG 4.37 PERCENTAGE COMPOSITION AND RESPONSIBILITY FOR RTAs OF MOTORCYCLES IN KENYA

with $r = 0.94$, $r^2 = 0.89$ where, $((\%)_{b'})_p$ is the

predicted percentage for pedal cyclists. Comparing

the observed data $((\%)_{b'})_o$ and the predicted yielded

the equation

$$((\%)_{b'}) = 0.979 \, ((\%)_{b'})_p + 0.040$$

with $r = 0.76$, $r^2 = 0.57$ and standard error of 1.655.

The level of significance for model 4.45 was found

to be 5 per cent. The coefficients of correlation

$(r = 0.76)$ and determination $(r^2 = 0.57)$ show that

there is consistency in the prediction. The slope

(0.979) and the intercept (0.040) are close to the

ideal values (1,0 respectively). This model is

therefore quite acceptable as the consistency and

calibration are good. This further implies that the

responsibility for RTAs by pedal cyclists have strongly

followed the trend given by equation 4.45.

For the growth in handcarts and animals per-

centage responsibility the trend curve was developed

as (Fig.4.38)

$$\left((\%)_{ha}\right)_p = 3.526 - 0.4 \, \ell nt \qquad (4.46)$$

with $r = -0.52$, $r^2 = 0.27$ where, $\left((\%)_{ha}\right)_p$ is the

predicted percentage responsibility for handcarts and

animals in year t. Comparing observed data $(\%)_{ha_o}$

and predicted data yielded

$$\left((\%)_{ha}\right)_o = 0.596 \quad \left((\%)_{ha}\right)_p + 0.983$$

with $r = 0.15$, $r^2 = 0.02$ and standard error of 0.331.
Model 4.46 was found to be statistically significant
at 25 per cent level. The coefficients of correlation
($r = 0.15$) and determination ($r^2 = 0.02$) show that the
predicted and observed values are not well correlated.
This implies that the prediction is not consistent
and the scatter is very considerable. This further
implies that the responsibility for RTAs of animals
and handcarts has remained stable. The slope (0.596)
and the intercept (0.983) suggest a need for further
calibration.

Finally, the growth model for pedestrian together
with passenger percentage responsibility was developed
as (Fig.4.38)

$$\left((\%)_{wp}\right)_p = 7.016 + 9.073 \, \ell nt \tag{4.47}$$

with $r = 082$, $r^2 = 0.67$ where, $\left((\%)_{wp}\right)_p$ is the pre-
dicted percentage using equation 4.47 . Comparing
observed data $\left((\%)_{wp}\right)_o$ and predicted data gave the
equation

$$\left((\%)_{wp}\right)_o = 0.736 \quad \left((\%)_{wp}\right)_p + 7.496$$

with r = 0.53, r^2= 0.28 and standard error of 7.498.
The model relationship described by equation 4.47
was found to be statistically significant at 5 per
cent level. The coefficients of correlation (r = 0.53)
and determination (r^2= 0.28) suggest fair correlation.
Further, the consistency of prediction is affected
adversely by the very considerable scatter. The
slope (0.736) and the intercept (7.496) show that
the model (4.47) has a tendency of under-predicting
at low ranges and over-predicting at high ranges.
The responsibility of pedestrians and passengers
for RTAs has also followed the trend suggested by
model 4.47 fairly closely. Thus the responsibility
by pedal cyclists, pedestrians and passengers, have
shown definite tendencies in Kenya.

Distribution by Age of RTA Victims killed

The growth model for the percentage distribution
of those killed above age 16 (Fig.4.39) was developed
using the logarithmic model as described above
(equation 3.42). The model is described by the
equation

$$\left((\%)_{D_{16+}} \right)_p = 86.051 - 2.884 \quad \ell nt \qquad (4.48)$$

with r = -0.72, r^2 = 0.52 where, $\left((\%)_{D_{16+}} \right)_p$ is the

predicted percentage using equation 4.48 . Comparing

GROWTH IN RESPONSIBILITY FOR PEDAL CYCLISTS, ANIMALS, HANDCARTS, PEDESTRIANS AND PASSENGERS

$$((\%)_{wp})_P = 7.016 + 9.073 \ln t$$

$$((\%)_{b'})_P = 11.748 - 2.003 \ln t$$

$$((\%)_{ha})_P = 3.526 - 0.400 \ln t$$

COMPARISON OF OBSERVED AND

PREDICTED DATA

$$((\%)_{wp})_O = 0.736((\%)_{wp})_P + 7.496$$

$$((\%)_{b'})_O = 0.979((\%)_{b'})_P + 0.040$$

$$((\%)_{ha})_O = 0.596((\%)_{ha})_P + 0.983$$

FIG 4.38 PERCENTAGE   RESPONSIBILITY FOR RTAs OF PEDAL CYCLISTS ANIMALS AND HAND -
CARTS, PEDESTRIANS AND PASSENGERS IN KENYA

observed data $\left((\%) D_{16} \right)_o$ and predicted yielded the equation

$$\left((\%) D_{16+} \right)_o = 0.211 \left((\%) D_{16+} \right)_p + 64.905$$

with $r = 0.05$, $r^2 = 0.003$ and standard error 1.547. The relationship in equation (4.48) was found not to be statistically significant. Considering that the predicted values and the observed values are not correlated ($r = 0.05$, $r^2 = 0.003$) model 4.48 requires more data for recalibration. The slope (0.211) and intercept (64.905) are far from ideal.

Distribution by Age of RTA Victims Injured

The growth model for the percentage distribution of those injured above age 16 (Fig.4.39) was found to be statistically significant at the per cent level of 20. The model was developed as

$$\left((\%) I_{16+} \right)_p = 91.01 - 2.468 \, \ell nt \qquad (4.49)$$

with $r = -0.81$, $r^2 = 0.66$ where, $\left((\%) I_{16+} \right)_p$ is the predicted percentage using equation 4.49 . The observed data $\left((\%) I_{16+} \right)_o$ compared with predicted data gave the equation

$$\left((\%)_{I_{16+}} \right)_o = 0.856 \left((\%)_{I_{16+}} \right)_p + 13.117$$

with $r = 0.40$, $r^2 = 0.16$ and standard error of 1.808.
Considering the coefficients of correlation ($r = 0.40$)
and determination ($r^2 = 0.16$) the scatter of the
predicted against observed values is very considerable
and the correlation weak. This implies that the
prediction by model 4.49 is not very consistent and
requires additional data observation. The calibration
is otherwise tending to the ideal considering that
the slope (0.856) and the intercept (13.117) are
approaching 1 and 0 respectively. With additional
data the characteristic trend curve could be improved. .

Distribution by Class of Road User of RTA Victims Killed

The growth in percentage distribution of RTA
drivers killed was developed into the model represented
by (Fig.4.40) the equation

$$\left((\%)_{D_{D'}} \right)_p = 14.695 - 0.541 \ln t \qquad (4.50)$$

with $r = -0.44$, $r^2 = 0.19$ where, $\left((\%)_{D_{D'}} \right)_p$ is the
predicted percentage distribution obtained by using
equation 4.50 . Comparing observed $\left((\%)_{D_{D'}} \right)_o$
and predicted data yielded the equation

GROWTH IN DISTRIBUTION BY AGE OF KILLED AND INJURED

$((\%)_{16+})_P = 91 \cdot 010 - 2 \cdot 468 \ln t$

$((\%)D_{16+})_P = 86 \cdot 051 - 2 \cdot 884 \ln t$

△ SMOOTHED DATA (INJURED)

○ SMOOTHED DATA (KILLED)

PERCENTAGE (%)

YEARS (t)

$((\%)_{D_{16+}})_O = 0.211 ((\%)D_{16+})_P + 54.905$

COMPARISON OF OBSERVED AND PREDICTED DATA

$((\%)_{I_{16+}})_O = 0.856 ((\%)_{I_{16+}})_P + 13.117$

OBSERVED %

PREDICTED %

FIG. 4.39   RTAs PERCENTAGE DISTRIBUTION BY AGE OF KILLED AND INJURED IN KENYA

$$\left((\%)_{D_{D'}} \right)_o = 3.147 \left((\%)_{D_{D'}} \right)_p - 30.145$$

with $r = 0.45$, $r^2 = 0.20$ and standard error of 0.398.
The relationship in equation 4.50 was found to be
statistically significant at the 10 per cent level.
The correlation coefficient ($r = 0.45$) and the
coefficient of determination ($r^2 = 0.20$) show that
there is considerable scatter and therefore the
prediction is not very consistent. The slope (3.147)
and the intercept (-30.145) indicate that the calibra-
tion of the model is far from ideal and requires
additional data to be obtained over the future. The
tendency shown by equation 50 is therefore a weak
one.

The percentage distribution of motorcyclist
killed was developed as the growth model (Fig.4.40)

$$\left((\%)_{D_M} \right)_p = 1.778\ \ell nt - 0.277 \qquad (4.51)$$

with $r = 0.90$, $r^2 = 0.80$ where, $\left((\%)_{D_M} \right)_p$ is the
percentage distribution predicted by equation 4.51.
Comparing observed values $\left((\%)_{D_M} \right)_o$ and predicted
values gave the relationship

$$\left((\%)_{D_M} \right)_o = 0.63 \left((\%)_{D_M} \right)_p + 0.884$$

with $r = 0.36$, $r^2 = 0.13$ and standard error of 1.303.
Equation 4.51 was found to be statistically signi-
ficant at the 20 per cent level. The trend shown
by equation 4.51 is a strong one ($r = 0.90, r^2 = 0.80$).
However the prediction is not very consistent
($r = 0.36$, $r^2 = 0.13$) as there is very considerable
scatter. The slope (0.63) and the intercept (0.884)
are tending to the required slope (1) and intercept
(0). With further data observation the calibration
could be improved even more.

The growth in percentage distribution of pedal
cyclists killed was developed as the model (Fig.4.40)

$$\left((\%)_{D_{B'}}\right)_p = 10.397 - 3.049 \ \ell n \ t \qquad (4.52)$$

with $r = 0.38$, $r^2 = 0.15$ where, $\left((\%)_{D_{B'}}\right)_p$ is the
predicted value of the percentage distribution of
pedal cyclists killed. Comparing observed data
$\left((\%)_{D_{B'}}\right)_o$ and predicted data yielded the equation

$$\left((\%)_{D_{B'}}\right)_o = 0.42 \ \left((\%)_{D_{B'}}\right)_p + 3.161$$

with $r = 0.38$, $r^2 = 0.15$ and standard error of
2.235, the relationship being significant at the
20 per cent level. The trend shown in equation
4.52 is weak. This has led to weak correlation
between predicted and observed values as shown by

the coefficients of correlation (0.38) and determina-
tion (0.15) for the regression of predicted values
against observed values. The considerable scatter
has led also to very poor consistency in prediction
resulting in poor calibration as seen by the slope
(0.42) being far from 1 and the intercept (3.161)
being far from 0. This implies need for additional
data over the coming years.

The growth in percentage distribution of pede-
strians killed in RTAs was developed as the model
(Fig.4.40)

$$\left((\%)_{D_W} \right)_p = 39.74 + 2.023 \ln t \qquad (4.53)$$

with $r = 0.87$, $r^2 = 0.77$ where, $\left((\%)_{D_W} \right)_p$ is the
percentage distribution predicted of pedestrians
killed. Comparing observed $(\%)_{D_W\ o}$ and predicted
data yielded the equation

$$\left((\%)_{D_W} \right)_o = 0.925 \left((\%)_{D_W} \right)_p + 3.197$$

with $r = 0.45$, $r^2 = 0.20$ and standard error of
1.484, the relationship being significant at
10 per cent level. The trend shown in model equation
4.53 was the strongest in terms of the distribution
of killed RTA victims. This shows the significant
unfortunate role played by pedestrians in RTAs in

Kenya on the one hand and on the other, the need for concentrating on pedestrian safety for the reduction of RTAs in Kenya. The slope (0.925) and the intercept (3.197) show that the calibration is very near the ideal (1, 0 respectively). The consistency can be improved however by further data observation as the scatter is very considerable ($r = 0.45$), $r^2 = 0.20$).

The growth in percentage distribution of passengers killed in RTAs was determined as the model equation (Fig.4.40)

$$\left((\%)_{D_P}\right)_P = 34.240 + 0.434 \ \ln t \qquad (4.54)$$

with $r = 0.37$, $r^2 = 0.14$ where, $\left((\%)_{D_P}\right)_P$ is the predicted percentage in equation 4.54 . The observed data $\left((\%)_{D_P}\right)_O$ and the predicted data were compared and yielded the equation

$$\left((\%)_{D_O}\right)_O = 1.339 \ \left((\%)_{D_P}\right)_P - 11.235$$

with $r = 0.21$, $r^2 = 0.04$ and standard error of 0.319. The relationship described by equation 5.54 was found to be significant at 30 per cent level. The regression equation of the predicted and observed values indicates by slope (1.339) and intercept (-11.235) that the

FIG 4.40 PERCENTAGE DISTRIBUTION OF PERSONS KILLED IN RTAs IN KENYA

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calibration is fair. However, the coefficient of
correlation (0.21) and determination (0.04) indicate
very considerable scatter and therefore very little
consistency in prediction. This is expected since the
trend in 4.54 is rather weak. The model could be
improved with additional future observations.

Distribution by Class of Road User of
RTA Victims Injured

Growth models were developed for the percentage
distribution of road users injured in RTAs (Fig.4.41)
using techniques as described above. For injured
drivers the model representing the distribution
growth is

$$\left((\%)_{I_{D'}}\right)_p = 19.19 - 2.251 \ \ln t \qquad (4.55)$$

with $r = -0.75$, $r^2 = 0.56$ where, $\left((\%)_{I_{D'}}\right)_p$ is the
predicted percentage distribution in equation (4.55).
The comparison of observed data $(\%)_{I_{D'}\ o}$ and
predicted data was found to be represented by the
equation

$$\left((\%)_{I_{D'}}\right)_o = 1.516 \ \left((\%)_{I_{D'}}\right)_p - 8.112$$

with $r = 0.59$, $r^2 = 0.35$ and standard error of 1.65,

GROWTH IN DISTRIBUTION OF INJURED PERSONS

$((\%)_{I_P})_P = 52.089 + 0.367 \ln t$

$((\%)_{I_{D'}})_P = 19.190 - 2.251 \ln t$

SMOOTHED DATA

$((\%)_{I_W})_P = 29.642 - 4.720 \ln t$

$((\%)_{I_{B'}})_P = 4.838 + 0.723 \ln t$

$((\%)_{I_M})_P = 2.216 + 1.131 \ln t$

COMPARISON OF OBSERVED
AND PREDICTED DATA

$((\%)_{I_P})_O = 3.448((\%)_{I_P})_P - 127.706$

$((\%)_{I_{D'}})_O = 1.516((\%)_{I_{D'}})_P - 8.112$

$((\%)_{I_W})_O = 0.486((\%)_{I_W})_P + 10.677$

$((\%)_{I_{B'}})_O = 6.175 - 0.057((\%)_{I_{B'}})_P$

$((\%)_{I_M})_O = 4.202 - 0.052((\%)_{I_M})_P$

FIG.4.41 PERCENTAGE DISTRIBUTION OF PERSONS INJURED IN RTAs IN KENYA

the relationship being significant at the 5 per cent
level. The trend shown by model 4.55 is fair as well
as the correlation (0.59) of the predicted values
against observed, but the consistency of prediction
is low as the scatter is very considerable ($r^2 = 0.35$).
The slope (1.516) and intercept (-8.112) indicate
over-prediction at low ranges and under-prediction
at high ranges. The model could improve in calibra-
tion with additional future data.

For motorcyclists the model developed is

$$\left((\%)_{I_M} \right)_p = 2.216 + 1.131 \, \ell nt \qquad (4.56)$$

with $r = 0.83$, $r^2 = 0.69$ where $\left((\%)_{I_M} \right)_p$ is the pre-
dicted percentage distribution by equation 4.56 .
Comparing observed data $\left((\%)_{I_M} \right)_o$ and predicted
data gave the equation

$$\left((\%)_{I_M} \right)_o = 4.202 - 0.052 \left((\%)_{I_M} \right)_p$$

with $r = -0.02$, $r^2 = 0$, standard error of 0.673 and
the relationship in equation (4.56) being found not
significant statistically. Although the trend shown
by model equation 4.56 is good ($r=0.82$, $r^2 = 0.69$),
there is no correlation between predicted values
and observed values ($r = -0.02$, $r^2 = 0$) as the scatter
is very considerable. This suggests further calibration

with additional data.

The model for the growth in percentage distribution of pedal cyclists injured in RTAs was developed as

$$\left((\%)_{I_{B'}} \right)_p = 4.838 + 0.723 \; \ln t \qquad (4.57)$$

with $r = 0.67$, $r^2 = 0.44$ where, $\left((\%)_{I_{B'}} \right)_p$ is the predicted value by equation 4.57 . Comparing observed data $\left((\%)_{I_{B'}} \right)_o$ and predicted data gave the equation

$$\left((\%)_{I_{B'}} \right)_o = 6.175 - 0.057 \; \left((\%)_{I_{B'}} \right)_p$$

with $r = -0.01$, $r^2 = 0$ and standard error of 0.529. The relationship in equation 4.57 was found not to be statistically significant. The model equation shows a fair trend. Due to the scatter of the data the slope (-0.057) and the intercept (6.175) are far from ideal. This suggests further data observation and recalibration.

The growth in percentage distribution of pedestrians injured in RTAs was modelled as

$$\left((\%)_{I_W} \right)_p = 29.642 - 4.72 \; \ln t \qquad (4.58)$$

with r = -0.86, r^2 = 0.75 where, $\left((\%)_{I_W} \right)_p$ is the

predicted value of the percentage distribution by

equation 4.58 . Comparing observed $\left((\%)_{I_W} \right)_o$

data and predicted data yielded the equation

$$\left((\%)_{I_W} \right)_o = 0.486 \left((\%)_{I_W} \right)_p + 10.677$$

with r = 0.30, r^2 = 0.09 standard error of 3.46,

the relationship in equation 4.58 being statisti-

cally significant at 20 per cent level. The trend

shown by model equation 4.58 is good but due to

the considerable scatter in the predicted values

against observed values the consistency of

prediction is poor. The slope (0.486) and intercept

(10.677) indicate much under-prediction at low

ranges and over-prediction at high values. This

implies need for further calibration with additional

data.

Finally, the growth in percentage distribution

of passengers injured in RTAs was modelled as

$$\left((\%)_{I_P} \right)_p = 52.089 + 0.367 \quad \ln t \qquad (4.5$$

with r = 0.18, r^2 = 0.03 where, $\left((\%)_{I_P} \right)_p$ is the

predicted percentage distribution of passengers

injured as predicted by the model in equation 4.59 .

The observed data $\left((\%)_{I_P} \right)_o$ was compared with the predicted data to yield the equation

$$\left((\%)_{I_P} \right)_o = 3.448 \left((\%)_{I_P} \right)_p - 127.706$$

with $r = 0.22$, $r^2 = 0.05$ and standard error of 0.268. The relationship described by the model equation 4.59 was found to be statistically signi-ficant at 30 per cent level. The trend shown by equation 4.59 is very weak and consequently because of much scatter the slope (3.448) and the intercept (-127.706) of the regression equation vary considera-bly from ideal. This suggests need for additional data.

4.2.2 Dual Carriageway Road Traffic Accidents

4.2.2.1 Data Collection

Data collection was based on police records of past RTAs based on the Police Form 41 (Appendix A.5). Forms 1 and 2 (Appendix A.6) were used for data acquisition. The study road falls under four Police Stations of Muthaiga, Ruiru, Juja and Thika. The information coded included road condition, traffic regulations and the environment, sociological and psychological conditions of the persons directly involved in RTAs.

The study area was zoned. The road network
was classified into road classes A, B, C, D and E.
The local road network consisting of feeder roads
and accesses and other locations consisting of
parking lots, yards, petrol stations and the like
were also classified. The nodes were located on
junctions between the trunk road and the secondary
roads. Each node was numbered in relation to its
zone. The number consisted of 4 digits where the
first three referred to the zone number and the last,
to the node in that zone (Appendix A.6/3). One zone
could only have nine nodes. If a junction was complex
or was a roundabout, a special area was introduced.
The special areas got their identification number
starting for example, with 1000. A special area
designated 1011 meant that it was the first special
area and the first node in that area being partitioned
into 9 subnodes only.

To define the location of a RTA the node
numbers were used. If the RTA ocurred at a junction
(node) only one number, the node number, was coded.
The section of road between two nodes was partitioned
into a grid of 8 squares horizontally and 9 vertically
in order to locate the RTA spot more precisely. The
spot was then coded using the horizontal and vertical
digits. The RTA spot at a junction was located pre-
cisely using an equally partitioned grid of 9 squares

horizontally and vertically. The coordinates of the
RTA spot were then coded.

The description of a RTA started with choosing
the primary elements, where an element was defined as
any vehicle or road user involved in the RTA in question
of which the primary elements were the principal or
main participants. In a single vehicle RTA there would
be only one primary element whereas in a vehicle-vehicle
or vehicle-pedestrian RTA for example there would be a
maximum of two primary elements. Thus, there was a
possibility of a maximum of two primary elements.
These primary elements were further supposed to be
the ones involved in the initial and last moments of
the RTA. Such primary elements would then as a result
of the RTA have incurred some degree of damage as a
result of the collision. Secondary elements were
defined as any other vehicles or road users involved
in the RTA in question other than the primary elements
defined above. Such secondary elements were deemed to
have been involved because they were indirectly made
to participate in the RTA in question due to their
presence in the traffic situation. Such involvement,
in the RTA in question, by the secondary elements
was defined as the disturbance phase of the RTA under
study.

4.2.2.2   Data Analysis

During the period January 1977 to July 1980, 725 RTAs were recorded on the study road. Of these, 702 were used for systematic analysis. Out of the total, 450 (64%) of the RTAs involved injuries and 130 (20%) were fatal. Although 167 persons were killed and 868 were injured. Therefore an average of 2.3 persons are killed or injured in every injury RTA on this road. The average annual increase in RTAs for the period 1977-79 was 16.5%. More than half of the fatal RTAs involved pedestrians. The RTA category which occurred most frequently apart from those involving pedestrians, were RTAs between vehicles travelling in the same direction (38%). Single vehicle RTAs had a share of 34% of the total.

The monthly distribution of RTAs (Fig.4.42) does not seem to indicate that certain seasons are significantly more RTA-prone than others. The two peak periods are May and December. These two coincide with the long rains in May when schools are closed and the Labour day public holiday (May 1), together with December when schools are closed and the Jamhuri celebrations (December 12) and of course Christmas and New Year celebrations.

The daily RTA trend (Fig.4.43) shows that week-ends have considerably more RTAs than the rest of the

FIG.4.42   NAIROBI-THIKA ROAD: MONTHLY ACCIDENT DISTRIBUTION

FIG.4.43   NAIROBI-THIKA ROAD: ACCIDENT DISTRIBUTION BY DAY OF WEEK

week. Of all RTAs 36% happened on Saturdays and
Sundays, but only 24% of the weekly traffic falls on
those two days. The time of the day with the most
RTAs was found to be between 5.00 p.m. and 8.00 p.m,
with 28% of the daily total (Fig.4.44). This is
partly due to the traffic peak between 5.00 p.m. and
6.00 p.m., the bad lighting conditions on the road
at that time and drunken driving. Almost, half of
the pedestrians killed in RTAs were killed at night.

Fig.4.45 shows the distribution by age of the
RTA elements (drivers, pedestrians) directly involved.
Although many did not reveal their ages, the figures
were nonetheless indicative and conclusive of the
fact that the most RTA involved age-group is ages
26-40. This group alone had an involvement per cent
of 36. The ages least involved according to this study
was found to be that over 56 years whose involvement
rate was only 2.4%.

Geometric factors have been responsible for many
of the RTAs studied. Some RTAs west of Utalii and
near Broadway Store/Allsopps had been indirectly caused
by the longitudinal gradient of the vertical alignment.
Due to the high proportion of heavy and very slow
speed vehicles on these sections, RTAs resulted
when they were overtaken by high speed traffic using
the inner lanes. Increased RTA rates were recorded

FIG 4.44   NAIROBI-THIKA ROAD ACCIDENT DISTRIBUTION BY HOUR OF DAY

FIG.4.45   NAIROBI-THIKA ROAD ACCIDENT INVOLVEMENT BY AGE GROUP

in both crest and sag curves as compared to the
tangent portions of the vertical alignment. The
vertical alignment and especially crest curves had
sight distances far below those allowed for the
traffic speed on the road [14]. For example, near
Clayworks, on the Nairobi bound carriageway, the sight
distance was found to be just over 100m and with a
bus stop and the access to Clayworks just behind the
crest curve, some RTAs recorded there, must have been
related directly to the alignment of the road. The
access to Clayworks had no provision for a decelera-
tion lane for vehicles turning into Clayworks. Many
RTAs were also caused indirectly by broken down
vehicles. Without provision for parking off the
traffic lanes, broken vehicles were often forced to
stand in either the inner or outer lanes. As a
result on-coming vehicles had no chance of stopping
in time and hit other vehicles when overtaking, due
to high speeds and short sight distances. Some
sections with high slopes without guardrails to pre-
vent falling towards the ditch had a high and serious
RTA rate. At two sections east of Githurai and east of
Kenyatta University, the narrow bridges were a signi-
ficant factor in RTA causation. When overtaking at or
near the bridge rails drivers got the psychological
feeling that the bridge rails were appearing too
close to the traffic lane. Then when trying to get
back to the outer lane, the vehicle hit others often

resulting in overturning. In some RTAs the bridge
rails were hit causing serious damage to the vehicles
and injury to persons. Besides the geometric layout,
road surfacing in general caused many RTAs. This was
especially observed in the outer lane, which was, at
the time of the study, cracked both on the surface
and edges. The bad surfacing together with high
speeds resulted in many RTAs. Drivers got aware of
the cracks or potholes only too late and in trying to
avoid them with a quick turning movement, lost control
over the vehicle either overturning on the carriageway
or into the ditch. At some locations the openings
in the central reserve had many RTAs. Both the paved
and track opening had bad sight distances in relation
to the traffic on the main road. These very short
sight distances coupled with the difficulty to judge
high speeds from a distance were causally related
to RTA occurrence there. At one of the most RTA
rated spots, the entering carriageway from Nairobi
towards Roysambu roundabout proved particularly
dangerous. There appeared to be no obstacles to sight
and at first glance the geometry looked satisfactory
but the manouvre at the roundabout led to RTAs. The
approaching carriageway was observed to have a crest
curve just before the roundabout. The geometry together
with underestimating speed is likely to have been the
cause of RTAs. Almost every RTA studied here involved
overturning of vehicles towards the centre of the round-
about.

        Vehicular factors were equally significant
in RTA causation. The mechanical condition of the
vehicles was found to be causing more RTAs than 8%.
In both the single vehicle RTAs and the vehicle-vehicle
RTAs between vehicles in the same direction, the causes
were related to the careless behaviour of drivers
combined with the bad mechanical conditions of the
vehicles. Burst tyres, faults in the breaking and
steering systems often resulted into loss of control,
overturning of vehicles and collisions between vehicles.
RTAs resulting into vehicles running off the carriage-
way were significant in frequency and severity. The
single vehicle RTAs were found to be more than twice
as likely to be fatal as other vehicle RTAs. Most
of these single vehicle (RTAs involved either over-
turning on the carriageway or on the roadside. Speed-
ing, improper driving, overloading, vehicle shape and
the unsatisfactory road conditions were significant
causation factors. For example, matatus were involved
in some fatal and serious RTAs with an indication of
overloading. Overloading left the front wheels with
light pressure on the road. Consequently, when hitting
a bump or pothole or making a sudden movement to avoid
an obstacle the vehicle easily overturned.

        Road user characteristics were observed to be
an important causal factor in RTAs. Most of the
fatalities were caused by pedestrians crossing the

Pedestrians had a high night to day RTA ratio. Only
closest to Muthaiga was there any lighting. At other
locations it was extremely difficult to discover
pedestrians on the carriageway in darkness.

Traffic signing was found to be generally very
poor. Many of the RTAs could be traced to lack of
warning signs to the dangerous locations. Reduce
speed signs were non-existent at the time of the
study (Appendix A.11).

In order to model RTAs independent variables
were chosen from the above analysis of the causative
factors influencing RTAs. The independent variables
chosen were longitudinal gradient, sight distance,
carriageway width, junctions, horizontal curve radius,
super-elevation, vehicle flow and time of day. All
coded RTAs were plotted along the study road. Using
traffic data obtained from the traffic counts vehicle-
kilometres were calculated for the period of study.
The dependent variable was then developed as RTAs per
10^6 vehicle-kilometres. For each of the above inde-
pendent variables the observed RTAs per 10^6 vehicle-
kilometres were recorded. RTAs per 10^6 vehicle-
kilometres were then plotted against each of the
independent variables. In order to obtain model
data it was necessary to smooth the data both for
the dependent and the independent variables. Data

smoothing was achieved by meaning, repeated meaning
by interpolation and use of moving averages. The aim
of data smoothing was to obtain values of the dependent
variable from group data representing the worst possible
RTA situation and the best possible RTA situation.
From these two extreme cases the most probable unsafe
condition representing the dependent variable was
chosen as the model data. This smoothed data was
then replotted against the independent variable with
ranges indicated where necessary. The plots indicated
certain curve functions. These were then tried for best
fit using models in section 3.2 and tested by the
methods of analysis outlined in section 3.5 (Appendix A.14).

Longitudinal Gradient

For longitudinal gradient a histogram of RTAs and
downgrade as well as upgrade gradient was constructed
(Fig.4.46). This revealed that flatter gradients are
more RTA prone. RTAs decrease with increase in gra-
dient. The data from both carriageways was combined.
A plot of the smoothed data for the upgrade gradients
was made (Fig.4.47). The curve shape was observed as
a quadratic polynomial function fitting techniques
using finite differences were used for points at
equal intervals to confirm the curve shape. Using
formulae 3.24 the model was developed as

FIG.4.46   NAIROBI – THIKA ROAD: NUMBER OF ACCIDENTS AGAINST GRADIENT

RTAs AND UPGRADE GRADIENT

$a_{g_{p_1}} = 0.9866 + 1.10666g - 0.18401g^2$

o smoothed data

COMPARISON BY AGOKI FORMULA

$a_{g_o} = 0.811a_{g_{p_1}} + 0.455$

COMPARISON BY SILYANOV FORMULA

$a_{g_o} = 0.025a_{g_{p_2}} + 1.956$

FIG.447 RELATION BETWEEN RTAs/10^6 VEHICLES-KILOMETRES AND UPGRADE GRADIENT

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$$(a_g)_{P_1} = 0.9866 + 1.10666g - 0.18401\ g^2 \qquad (4.60)$$

$$0 < g < 6$$

where, $(a_g)_{P_1}$ is the predicted RTAs per 10^6 vehicle-
kilometres and g is the upgrade gradient in per
cent. Comparing observed data $(a_g)_o$ and the predi-
cted data yielded the equation

$$(a_g)_o = 0.811\ (a_g)_{P_1} + 0.455$$

with $r = 0.68$, $r^2 = 0.46$ and standard error of 0.781.
The relationship given by equation 4.60 was found
to be significant at 20 per cent level. For down-
grade gradients the model developed in the same way
as above became (Fig.4.48)

$$(a_g)_{P_3} = 2.993 + 0.11g - 0.05165\ g^2 \qquad (4.61)$$

$$0 < g < 8$$

where, $(a_g)_{P_3}$ is the predicted number of RTAs per
10^6 vehicle-kilometres and g the downgrade gradient
in per cent. The observed data and predicted data by
equation 4.61 were compared and yielded the equation

$$(a_g)_o = 1.018\ (a_g)_{P_3} - 0.095$$

RTAs AND DOWNGRADE GRADIENT

$a_{g_{p_3}} = 2.993 + 0.11g - 0.05165g^2$

o smoothed data

COMPARISON BY AGOKI FORMULA

$a_{g_0} = 1.018a_{g_{p_3}} - 0.095$

COMPARISON BY SILYANOV FORMULA

$a_{g_0} = 3.584 - 1.180a_{g_{p_4}}$

FIG.448   RELATION BETWEEN RTAs/VEHICLE-KILOMETRES AND DOWNGRADE GRADIENT

with r = 0.97, r^2 = 0.94. The relationship (4.61) was found to be significant at the 5 per cent level.

Sight Distance

Using formulae 3.24 as before the models related to sight distance were developed separately for each of the two carriageways. For the Nairobi-to-Thika carriageway the model was developed as (Fig.4.49)

$$(a_s)_{P_1} = -8.745 + 0.079815 - 0.000136 \ s^2$$

(4.62)

$$150 < 5 < 400$$

where, $(a_s)_{P_1}$ is the predicted number of RTAs per 10^6 vehicle-kilometres and S is the sight distance measured in metres. Comparing observed data $(a_s)_o$ against predicted data yielded the equation

$$(A_s)_o = 0.723 \ (a_s)_{P_1} + 0.826$$

with r = 0.83, r^2 = 0.69 and a standard error of 0.996. For the Thika-to-Nairobi carriageway the equation of the model was developed as (Fig.4.50)

$$(a_s)_{P_2} = -1.8551 + 0.03467 \ S - 0.00006 \ s^2$$

... (4.63)

$$150 < 5 < 500$$

RTAs AND SIGHT DISTANCE

$a_{S_1} = -8.745 + 0.0798S - 0.00014S^2$

o smoothed data

COMPARISON BY AGOKI FORMULA

$a_{S_0} = 0.723 a_{S_{p_1}} + 0.826$

COMPARISON BY SILYANOV FORMULA

$a_{S_0} = 5.609 - 1.734 a_{S_{p_2}}$

FIG 4.49 RELATION BETWEEN RTAs/10^6 VEH-KMs AND SIGHT DISTANCE: NAIROBI-TO-THIKA CARRIAGEWAY

RTAs AND SIGHT DISTANCE

$a = -1.855 + 0.035S - 0.00006S^2$

o smoothed data

COMPARISON BY AGOKI FORMULA

$a_{s_0} = 0.550 a_{s_{p_1}} + 1.057$

COMPARISON BY SILYANOV FORMULA

$a_{s_0} = 3.463 - 0.473 a_{s_{p_2}}$

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FIG 4.50 RELATION BETWEEN RTAs/10^6 VEH-KMs   AND   SIGHT DISTANCE: THIKA-TO-NAIROBI CARRIAGEWAY

where, $(a_s)_{P_2}$ is the predicted number of RTAs per 10^6

vehicle-kilometres predicted by equation 4.63 .
Comparing observed and predicted data as before
yielded the equation

$$(a_s)_o = 0550 \ (a_s)_{P_2} + 1.057$$

with r = 0.74, r^2 0.55 and standard error of 0.441.
Both models for sight distance were found to be
significant at 5 per cent level.

Carriageway Width

After data smoothing RTAs per 10^6 vehicle-
kilometres and plotting them against carriageway
width (Fig.4.51) the relationship took the shape of
a linear function. Using model 3.19 the model
relating RTAs per 10^6 vehicle-kilometres a_w and
carriageway width (w) in metres was developed as
the equation

$$(a_w)_{P_1} = 32.6439 - 4.2348 \ W \qquad (4.64)$$

$$5.8 < W < 6.5$$

with r = -0.99, r^2 = 0.99 where, $(a_w)_{P_1}$ is the predi-
cted data using equation 4.64 . The observed data

RTAs AND CARRIAGEWAY WIDTH

$$a_{W_1} = 32.644 - 4.235W$$

o smoothed data

COMPARISON BY AGOKI FORMULA

$$a_{W_0} = 0.883 a_{W_{p_1}} + 1.603$$

COMPARISON BY SILYANOV FORMULA

$$a_{W_0} = 15.920 a_{W_{p_2}} - 11.293$$

FIG 4.51 RELATION BETWEEN RTAs/10^6 VEH-KMs AND CARRIAGEWAY WIDTH: NAIROBI THIKA ROAD

$(a_w)_o$ and the predicted data were compared yielding the equation

$$(a_w)_o = 0.883 (a_w)_{p_1} + 1.603$$

with $r = 0.88$, $r^2 = 0.78$ and standard error of 1.316. The relationship was found to be significant at the 5 per cent level. It was found necessary to combine data from both carriageways to yield a sensible relationship.

Junctions

Models relating RTAs per 10^6 vehicle-kilometres to junctions per kilometre (j) were developed for each of the carriageways. After data smoothing the plot for the Nairobi-to-Thika carriageway (Fig.4.52) suggested a linear model which by using equation 3.19 was developed as

$$(a_j)_{p_1} = 5.068 + 1.672j \qquad (4.65)$$

$$0 < j < 4$$

with $r = 0.92$, $r^2 = 0.84$ where $(a_j)_{p_1}$ is the predicted data using equation 4.65 . Comparing observed data $(a_j)_o$ and predicted data gave

$$(a_j)_o = 1.001 (a_j)_{p_1} - 0.004$$

RTAs AND JUNCTIONS

(b.) RTAs/10⁶ VEH-KMs — JUNCTIONS/KM (j)

$a_{rp} = 5.068 + 1.672j$

COMPARISON BY AGOKI FORMULA
OBSERVED RTAs/10⁶ VEH-KMs — PREDICTED RTAs/10⁶ VEH-KMs
$a_{jo} = 1.001a_{jp} - 0.004$

o smoothed data

COMPARISON BY JACOBS FORMULA
OBSERVED RTAs/10⁶ VEH-KMs — PREDICTED RTAs/10⁶ VEH-KMs
$a_{jo} = 1.639a_{jp} + 2.691$

FIG.4.52 RELATION BETWEEN RTAs/10⁵ VEH-KMs AND JUNCTIONS/KM: NAIROBI-TO-THIKA CARRIAGEWAY

with $r = 0.92$, $r^2 = 0.84$ and standard error of 2.644.
For the Thika-to-Nairobi carriageway (Fig.4.53) the
model developed was

$$(a_j)_{P_2} = 3.0099 + 2.9239j \qquad (4.66)$$

$$0 < j < 6$$

with $r = 0.99$, $r^2 = 0.99$ where, $(a_j)_{P_2}$ is the predicted
data by equation 4.66 . Comparing observed and pre-
dicted data yielded the equation

$$(a_j)_o = 0.97(a_j)_{P_2} - 0.696$$

with $r = 0.93$, $r^2 = 0.86$ and standard error of 6.316.
Both models for junctions effect were found to be
significant at the 5 per cnet level.

Horizontal Radius

The influence of horizontal alignment is often
thought to affect the vehicles that turn off the
road to the right or left. All such RTAs were
studied carefully. They were found to be
132 (19%) out of the total 702 RTAs studied. Of the
132 studied for horizontal alignment performance, 37%
occurred where a vehicle turned off to the left on a
straight road section (radius = ∞), as opposed to
24% which turned off to the right on a straight section.

FIG 4.53 RELATION BETWEEN RTAs/10⁶ VEH-KMs AND JUNCTIONS/KM. THIKA-TO

COMPARISON BY JACOBS FORMULA

$$a_{j_0} = 2.781 a_{j_{p_2}} - 1.808$$

OBSERVED RTAs/10^6 VEH-KMs

PREDICTED RTAs/10^6 VEH-KMs

- NAIROBI CARRIAGEWAY

Since on Kenyan roads vehicles are driven on the left
side it may be argued that this is the main reason
for a greater percentage turning off to the left. A
vehicle turning off to the right has a whole lane
width in which to regain in a corrective manoeuvre
assuming no other vehicles are occupying the opposite
lane. On curves, 11% of the RTAs involved vehicles
turning off to the left on a right curve whilst 7%
turned to the right on a right curve. On left curves
8.3% of the vehicles turned off to the left and 9.8%
to the right. The severity of injury in relation to
the horizontal alignment was found to be 0.9 injury
RTAs per kilometre on straights and 1.9 injury RTAs
per kilometre on curves. The fatality rate was found
to be 0.2 fatal RTAs per kilometre on straights and
curves. It is seen that in spite of a greater number
of RTAs on straights, the injury rate on curves is
higher. In similar studies [12] only driver injury
is considered so as to avoid variation due to the
number of occupants in the vehicle. The above rates
have been treated the same. For saloon cars and vans
there was a strong indication that RTA occurrence is
higher at right curves than either straights or left
curves. Of these RTAs studied in detail, 53.8%
involved saloons or lighter vehicles, 25% involved
vans, 12% involved lorries and 9% involved buses.
Generally, it can be inferred that the frequency
of occurrence decreases with increase in vehicle masses.

For lighter vehicles the pattern observed here is to be expected given that a moving object tends to continue in motion in a straight line until redirected and vehicles are ordinarily travelling on the left. Evidently, lighter vehicles are travelling at higher speeds than heavy ones. Hence speed is a significant factor in contributing to RTA causation. For the same phenomenon there were, only one RTA involving a bus, one involving a lorry on a right curve and two involving lorries and three buses on left curves.

The horizontal alignment effect was analysed using two independent variables. These were the horizontal curve radius and the superelevation of the horizontal alignment. Further, in order to distinguish the effect of upgrade and downgrade vertical curves, the horizontal curves were divided into those occurring on crests and those occurring at sags. Firstly, the RTA density per kilometre for each horizontal curve was calculated and a histogram constructed (Fig.4.54). This showed that small radii curves were more RTA prone than larger radii curves. However, RTAs per kilometre rose again with increased radius. To confirm this observation the histogram for RTAs and superelevation was constructed (Fig.4.57). This revealed that as superelevation tended to zero (i.e. horizontal radius tended to infinity) RTAs increased and as superelevation increased (i.e. horizontal radius

FIG.4.54   NAIROBI-THIKA ROAD: GRAPH OF NUMBER OF ACC

decreased) RTAs decreased. These observations were used in plotting RTAs per 10^6 vehicle-kilometres against horizontal curve radius. For upgrade curves the smoothed plot (Fig.4.55) revealed two bends in the curve. By polynomial fitting techniques and confirming the shape by the finite differences technique formulae 3.26 were used and the model developed as

$$(a_R)_{P_1} = 3.346 - 0.0009R + 0.77 \times 10^{-7} R^2$$

$$- 1.361 \times 10^{-12} R^3 \qquad (4.67)$$

$$130 < R < 2500$$

where, $(a_R)_{P_1}$ is the number of predicted RTAs per 10^6 vehicle-kilometres using equation 4.67 and R is radius in metres. Observed data $(a_R)_o$ was compared with predicted data and yielded the equation

$$(a_R)_o = 1.011 (a_R)_{P_1} - 0.055$$

with r = 0.67, r^2 = 0.44 and standard error of 0.774. For downgrade curves the smoothed plot (Fig.4.56) revealed a quadratic function. Using formulae 3.24 and confirming by finite differences technique the relationship was modelled as

$$(a_R)_{P_2} = 6.8699 - 0.0031R + 0.000\ 000\ 397 R^2 \quad \ldots (4.68)$$

$$130 < R < 6000$$

RTAs AND 'HORIZONTAL RADIUS

o smoothed data

$a_{R_1} = 3.346 - 0.0009R + 0.77 \times 10^{-7} R^2$
$- 1.361 \times 10^{-12} R^3$

COMPARISON BY AGOKI FORMULA

$a_{R_0} = 1.011 a_{R_{p_1}} - 0.055$

COMPARISON BY SILIANOV FORMULA

$a_{R_0} = 0.194 a_{R_{p_2}} + 2.112$

FIG 4.55 RELATION BETWEEN RTAs/10^5 VEH-KMs AND HORIZONTAL RADIUS (UPGRADE CURVES): NAIROBI - THIKA ROAD

RTAs AND HORIZONTAL RADIUS

$$a_{R_1} = 6.8599 - 0.0031R + 0.000000397R^2$$

RTAs/10^6 VEH-KMs

HORIZONTAL RADIUS (m)

o smoothed   data

COMPARISON BY AGOKI FORMULA

$$a_{R_0} = 0.884 a_{R_{p_1}} + 0.555$$

OBSERVED RTAs/10^6 VEH-KMs

PREDICTED RTAs/10^6 VEH-KMs

COMPARISON BY SILYANOV FORMULA

$$a_{R_0} = 0.827 a_{R_{p_2}} + 2.270$$

OBSERVED RTAs/10^6 VEH-KMs

PREDICTED RTAs/10^6 VEH-KMs

FIG.4.56 RELATION BETWEEN RTAs/10^5 VEH-KMs AND HORIZONTAL RADIUS (DOWNGRADE CURVES): NAIROBI-THIKA ROAD

where, $(a_R)_{P_2}$ is the predicted value of RTAs per 10^6

vehicle-kilometres using equation (4.68). Observed

data compared with predicted yielded the equation

$$(a_R)_o = 0.844 \ (a_R)_{P_2} + 0.555$$

with $r = 0.88$, $r^2 = 0.77$ and standard error of 1.843.

Both models (4.67) and (4.68) were found to be

significant at 5 per cent level.

Superelevation

The superelevation of upgrade curves was

separated from that of downgrade curves. Observed

RTAs per 10^6 vehicle-kilometres were smoothed as

afore-mentioned and plotted against superelevation.

For the upgrade curves the model was developed using

formulae (3.24) and polynomial function fitting

techniques. The model was developed as (Fig.4.58)

$$(a_\alpha)_{P_1} = 2.2729 + 0.6\alpha - 0.098\alpha^2 \qquad (4.69)$$

$$0 < \alpha < 5$$

where, $(a_\alpha)_{P_1}$ is the predicted value of RTAs per 10^6

vehicle-kilometres and α is superelevation in per

cent. Comparing observed data $(a_\alpha)_o$ and predicted

data yielded the equation

FIG.4.57 NAIROBI-THIKA ROAD: ACCIDENT FREQUENCY AGA

RTAs AND SUPER-ELEVATION

$$a_{\alpha_{p_1}} = 2.2729 + 0.60397\alpha - 0.0098028\,\alpha^2$$

RTAs/10^6 VEH-KMs

SUPERELEVATION (UPGRADE CURVES) (%)

o smoothed data

RTAs AND SUPER-ELEVATION

$$a_{\alpha_{p_2}} = 1.0693 + 1.31316\alpha - 0.220446\,\alpha^2$$

RTAs/10^6 VEH-KMs

SUPERELEVATION (DOWNGRADE CURVES) (%)

COMPARISON BY AGOKI FORMULA

$$a_{\alpha_0} = 0.990\,a_{\alpha_{p_1}} + 0.142$$

OBSERVED RTAs/10^6 VEH-KMs

PREDICTED RTAs/10^6 VEH-KMs

COMPARISON BY AGOKI FORMULA

$$a_{\alpha_0} = 0.945\,a_{\alpha_{p_2}} + 0.072$$

OBSERVED RTAs/10^6 VEH-KMs

PREDICTED RTAs/10^6 VEH-KMs

FIG.4.58   RELATION BETWEEN RTAs/10^6 VEHICLE - KILOMETRES AND SUPERELEVATION: NAIROBI - THIKA ROAD

- 233 -

$$(a_\alpha)_o = 0.99 (a)_{p_1} + 0.142$$

with r = 0.7, r^2 = 0.5 and standard error of 0.43.
Similarly, the model for downgrade curves was developed
as

$$(a_\alpha)_{p_2} = 1.0693 + 1.3132\alpha - 0.22\alpha^2 \tag{4.70}$$

$$o < \alpha < 4.5$$

where, $(a_\alpha)_{p_2}$ is the predicted value by equation
4.70 . Observed and predicted values were compared
yielding the equation

$$(a_\alpha)_o = 0.945 (a_\alpha)_{p_2} + 0.072$$

with r = 0.91, r^2 = 0.83 and standard error of
0.718. Both sueprelevation models were significant
at 5 per cent level.

Vehicle Flow

To model the effect of vehicle flow on RTAs
it was necessary to combine observed data for Kiganjo-
Nanyuki Road (single carriageway trunk road) and
each of the carriageways of the dual carriageway road
(the Nairobi-Thika Road). This was found necessary

because applying data smoothing techniques adopted
for this study, the range of vehicle flow on the
single carriageway yielded only a single model data
point. It was observed that RTAs rise slowly at first
with increase in vehicle flow to reach a saturation
level where RTAs are many but a majority of which are
non injury. Since the RTAs studied here were mainly
injury RTAs, they represent the earlier portion of
a logistic curve model. Therefore, the equation
3.15 was the basis for modelling. The result of
the analysis of the combined Kiganjo-Nanyuki road
and the Nairobi-to-Thika carriageway (Fig.4.59) was
the model

$$(a_q)_{P_1} = \frac{18.89}{1 + 24.128\ e^{-0.002q}} \qquad (4.71)$$

$$30 < q < 1430$$

where 18.89 is the highest observed level of RTAs/km/
annum used as approximate limit,$(a_q)_{P_1}$ is the pre-
dicted RTAs per kilometre per annum and   q   is
average vehicle flow per hour. Observed data
$(a_q)_{o}$ compared to predicted data yielded the relation-
ship

$$(a_q)_{o} = 0.966\ (a_q)_{P_1} + 0.065$$

RTAs AND VEHICLE FLOW

$$a_{q_p} = \frac{18.89}{1 + 24.128e^{-0.002q}}$$

• single carriageway smoothed data
o dual   "   "   "

COMPARISON BY AGOKI FORMULA

$a_{q_o} = 0.966a_{q_{p_1}} + 0.065$

COMPARISON BY TRRL FORMULA

$a_{q_o} = 0.415a_{q_{p_2}} + 0.126$

COMPARISON BY SILYANOV FORMULA

$a_{q_o} = 6.782a_{q_{p_2}} - 1.351$

FIG.459 RELATION BETWEEN RTAs/KM PER ANNUM AND VEHICLE FLOW: COMBINED KIGANJO-NANYUKI ROAD AND NAIROBI-TO-THIKA CARRIAGEWAY

with r = 0.97, r^2 = 0.94 and standard error of 0.937.
The result of the analysis of the combined Kiganjo-
Nanyuki Road and the Thika-to-Nairobi carriageway
(Fig.4.60) was the model

$$(a_q)_{P_2} = \frac{24.55}{1 + 38.983\ e^{-0.003q}} \qquad (4.72)$$

$$30 < q < 1340 .$$

where $(a_q)_{q_2}$ is the predicted RTAs/km/annum predicted
by equation (4.72) and 24.55 being the approximation
for the limit taken here as the highest value observed.
Comparing observed and predicted data yielded the
equation

$$(a_q)_o = 0.921\ (a_q)_{P_2} + 0.11$$

with r = 0.96, r^2 = 0.91 and standard error of 1.679.
Both models were significant at 5 per cent level.

RTAs per Hour

Using average vehicle flow for the two carriage-
ways of the Nairobi-Thika Road, all the RTAs occurring
each hour of the day for the 3½ years of RTA study
the mean RTAs per 10^6 vehicle-kilometres for each
hour of the day was calculated (Fig.4.44). Mean RTAs

FIG.4.60 RELATION BETWEEN RTAs / KM PER ANNUM AND VEHICLE FLOW: COMBINED KIGANJO-NANYUKI ROAD AND THIKA-TO-NAIROBI CARRIAGEWAY

per 10^6 vehicle-kilometres were plotted against time
of day and the data smoothed (Fig.4.61). By harmonic
analysis techniques (3.4) curves were fitted for the
unsmoothed and smoothed data. For the unsmoothed
data the model was developed as

$$(a/K)_{P_1} = 0.392 - 0.207 \cos t - 0.055 \cos 2t + 0.033 \cos 3t$$

$$- 0.095 \sin t - 0.147 \sin 2t \qquad \ldots (4.73)$$

where $(a/K)_{P_1}$ is predicted data from the unsmoothed
curve at time t. Comparing observed data $(a/K)_o$
and predicted data yielded the equation

$$(a/K)_o = 1.358 \ (a/K)_{P_1} - 0.026$$

with r= 0.77, r^2 = 0.59 and standard error of
0.201. For the smoothed data the model developed
was

$$(a/K)_{P_2} = 0.538 - 0.267 \cos t - 0.085 \cos 2t + 0.033 \cos 3t$$

$$- 0319 \sin t - 0.0095 \sin 2t \qquad ..(4.74)$$

where, $(a/K)_{P_2}$ is predicted value using equation
4.74 . The comparison of observed and predicted
data yielded the equation

$$(a/K)_o = 0.979 \ (a/K)_{P_2} - 0.020$$

PREDICTION OF RTAs BY TIME OF DAY

$$(a/K)_{P_1} = 0.392 - 0.207\cos t - 0.055\cos 2t + 0.033\cos 3t - 0.095\sin t - 0.147\sin 2t$$

$$(a/K)_{P_2} = 0.538 - 0.267\cos t - 0.085\cos 2t + 0.033\cos 3t - 0.319\sin t - 0.095\sin 2t$$

● UNSMOOTHED DATA   ○ SMOOTHED DATA

COMPARISON OF OBSERVED AND PREDICTED DATA

$$(a/K)_0 = 1.358 (a/K)_{P_1} - 0.026$$

$$(a/K)_0 = 0.979 (a/K)_{P_2} - 0.020$$

G.4.61 PREDICTION OF MEAN RTAs PER 10^6 VEH-KMs:
NAIROBI - THIKA ROAD

with r = 0.87, r^2 = 0.75 and standard error of 0.315.
Both models were significant at the 5 per cent level.

4.2.3   Single Carriageway Road Traffic Accidents

4.2.3.1   Data Collection

The methodology for RTAs data collection on the
single carriageway was identical to the one used
for the dual carriageway as outlined in section
4.2.2.1 and on the basis of the Police Form P41.   The
study road fell under three Police Stations at Kiganjo,
Naro Moru and Nanyuki.

4.2.3.2   Data Analysis

A total of 94 injury and fatal RTAs was
recorded during the study period January 1979 to
December 1982.   Non-injury RTAs were not studied.
Of the total injury RTAs 28% were fatal.   Nearly
10% of the fatal RTAs occurred at junctions but
many more were associated with junctions and accesses.
The annual increase for 1979-80 was 32% and that for
1981-82 was 38%.   An annual decrease of 16% was
recorded for 1980-81.   Of the fatal RTAs 20% involved
pedestrians.   The most frequent RTAs involved over-
turning (21%), head-on collision (21%) and turning
off the road onto the roadside (21%).   The second
category most frequently observed involved vehicles
driving in the same direction (14%) and pedestrians

crossing the carriageway (13%).

The monthly distribution of RTAs (Fig.4.62) indicated that March, May, September and December were the most RTA rated months for the four quarters of the year respectively. As observed earlier May is the month for the long rains and Labour day public holiday. December is the Christmas and New Year festivities as well as school holidays.

The daily RTA trend showed that (Fig.4.63) weekends had considerably more RTAs than the rest of the week. Of the total RTAs 59% occurred on Friday, Saturday and Sunday. For the remaining days Tuesday had a higher proportion than the others (17%). This trend was observed to coincide with high traffic volumes on Friday and Saturday. The worst times of the day (Fig.4.64) were observed to be 12.00 noon and 5.00 p.m., coinciding with the traffic peaks. These hours alone had 14% and 15% of the fatal RTAs respectively.

The road factors on this road found to be causally related to RTAs were road work or cleaning, damaged carriageway, loads on carriageway. The road environment contributed to 19% of the RTAs directly.

Vehicular factors on this road contributed to

FIG.4.62   KIGANJO - NANYUKI ROAD MONTHLY ACCIDENT DISTRIBUTION

FIG.4.63   KIGANJO - NANYUKI ROAD ACCIDENT DISTRIBUTION BY DAY OF WEEK

FIG.4.64   KIGANJO - NANYUKI ROAD ACCIDENT DISTRIBUTION BY HOUR OF DAY

FIG.465 KIGANJO-NANYUKI ROAD. DISTRIBUTION OF ACCIDENT VEHICLES BY AGE

FIG.4.66 KIGANJO-NANYUKI ROAD ACCIDENT INVOLVEMENT BY DRIVING EXPERIENCE

21% of the RTAs directly. Light vehicles like cars,
pick-ups and vans constituted 45% of the RTA elements.
Matatus constituted 11%, heavy vehicles constituted
21% and pedestrians 13% of the RTA elements. The
most frequent vehicular failures contributing to
RTA causation were, in order of their frequency:
steering defects, tyre bursts or defects, general
unroadworthiness and brake defects. For night RTAs
the dazzling light from on-coming vehicles was a
significant causation factor. From the ages of
vehicles (Fig.4.65) involved in RTAs it was observed
that new vehicles (1-3 years old) featured most
frequently and generally the frequency decreasing
with increasing age.

Road user factors were observed to be causatively
associated to 60% of the RTAs. The causes were
ascribed to the driver's careless behaviour and
inattention. From the data observed RTA involvement
by drivers increased with years of experience (Fig.4.66).
In particular drivers with 9 or more years of
experience were the most involved. Saloon cars were
involved in 36% of the fatal RTAs. Most of them
were single vehicles either overturning or hitting
objects. These were observed to be associated with
speeding. Pedestrians were hit when crossing
the carriageway behind parked vehicles (masked) or
rushing suddenly onto the carriageway in the path of

FIG4.67 KIGANJO-NANYUKI ROAD ACCIDENT INVOLVENT BY AGE GROUP

speeding vehicles. Matatus were involved due to
standing in the carriageway or improperly driven
causing collisions with other vehicles. Pedal
cyclists and motorcyclists were involved in turning
or crossing movements without due care and attention.
Some light goods vehicles were involved in fatal
RTAs due to overloading of miraa (addictive drug
chewing plant) and miraa dealers. Significant
fatal RTAs were also caused by disturbance from
other vehicles such as hanging or projecting objects,
wide loads or falling ones. The use of alcohol played
a significant role in fatal RTAs, particularly in
single vehicle RTAs and pedestrians walking along
the road.

Of all RTAs studied, 26% occurred at night,
the most frequent being head-on collision
turning off the road and those associated with a
single vehicle. A third of the pedestrian RTAs
also occurred at night. One of the most significant
illumination problems on this road was pedestrians
and pedal cyclists not being seen (Appendix A.9).

Traffic signing on the road was observed to be
very unsatisfactory. Many dangerous locations had
no warning signs. The few signs observed on the
road were rusty and a majority of them non-standard.

Other factors associated with RTA causation were
animals and trees on the roadside. RTAs involving
cattle and sheep occurred when they crossed the road
from nearby bushes. RTAs involving wild animals
occurred mainly at night and involved crossing. Along
the Nanyuki end of the road some fatal and serious
RTAs occurred in connection with vehicles hitting
trees along the road.

Age and sex were used as social indicators of
those responsible for RTAs. Of all the drivers,
pedestrians, pedal and motorcyclists involved 87.1%
were found to be male (Appendix A.10). The females
were 4.3%, undetermined road users were 6.7% and
animals had a 1.8% share. The age distribution data
(Fig.4.67) revealed that the most affected ages lie
in the group 17-49. This was true particularly of all
road users and those primary elements in RTAs causation.

The geometric elements on the single carriageway
did not feature directly as causative factors. They
modified RTAs whose main causation lay in other factors.
It was observed however, that the road had many junctions
and accesses which influenced RTA occurrence directly.
Pavement conditions were also observed to have caused
RTAs directly for example potholes, damaged sections
and repair works. The overall assessment of the RTA
situation left as independent variables to be studied

RTAs AND JUNCTIONS

$a_{jp_1} = 0.567 + 1.108j - 0.129j^2$

o smoothed data

COMPARISON BY AGOKI FORMULA

$a_{j_0} = 1.004a_{jp_1} - 0.009$

COMPARISON BY JACOBS FORMULA

$a_{j_0} = 2.670 - 0.050a_{jp_2}$

FIG 4.58 RELATION BETWEEN RTAs/10^{6}VEH-KMs AND JUNCTIONS/KM: KIGANJO-NANYUKI ROAD

as junctions, pavement condition, vehicle flow and
time of day. As earlier noted, the smoothing of
RTA data in relation to the low vehicle flow on this
road yielded only a single data point for modelling.
Therefore it was necessary to combine these data
with the data of the dual carriageway. In making
the combination it was assumed that traffic behaviour
exhibited on the single carriageway trunk road would
be similar to that exhibited on the dual carriageway
trunk road. In any case from the traffic studies
on the single carriageway considerable traffic was
found to originate in Nairobi destined for Nanyuki
and beyond. Further, it was found that when vehicle
flow data and RTAs for the single carriageway were
plotted on the graph of the dual carriageway data,
the observations of the single carriageway clustered
near the origin confirming the assumption. This way
it was possible to use the model developed from data
from both carriageways for predicting the effect of
vehicle flow on RTAs for low trafficked rural roads
as well as for high trafficked urban and semi-urban
roads (Appendix A.12).

The remaining independent variables and the
dependent variable RTAs per 10^6 vehicle-kilometres
were subjected to the same smoothing process as
outlined earlier and plots made. The suggested

curve shapes were then fitted by the modelling
techniques outlined in Chapter 3.

Junctions

For junctions per kilometre and their effect
on RTAs (Fig.4.68) the model developed was a polyno-
mial of second degree given by the equation

$$(a_j)_p = 0.6668 + 1.1082j - 0.1288j^2 \qquad (4.75)$$

$$1 < j < 8$$

where, $(a_j)_p$ is the predicted number of RTAs per
10^6 vehicle-kilometres for j junctions per kilo-
metre. Observed data $(a_j)_o$ compared with predicted
data yielded the equation

$$(a_j)_o = 1.004(a_j)_p - 0.009$$

with $r = 0.98$, $r^2 = 0.96$ and standard error of
0.642. The relationship was found to be signifi-
cant at 5 per cent level.

Pavement Defects

The pavement defects were observed under the
following: rutting; crazing and cracking; potholes,
patches, depressions and upheavals; edge spalling.
After data smoothing each separate plot the data

FIG. 4.69   RELATION BETWEEN RTAs/10^6 VEH-KMs AND RUTTING

FIG. 4.70 RELATION BETWEEN RTAs/10^6 VEH-KMs AND CRAZING,
CRACKING : KIGANJO - NANYUKI ROAD

fitted a quadratic function as before. For rutting
the model developed was represented by the equation
(Fig.4.69)

$$(a_{r'})_p = 4.1084 - 0.17319r' + 0.0033678r'^2 \qquad (4.76)$$

$$0 < r' < 50$$

where, $(a_{r'})_p$ is the predicted value of RTAs per
10^6 vehicle-kilometres for a given value of rutting
measured in millimetres. Observed data $(a_{r'})_o$ was
compared with predicted data and yielded the
equation

$$(a_{r'})_o = 0.998 \ (a_{r'})_p + 0.006$$

with $r = 0.94$, $r^2 = 0.88$ and standard error of
0.785, the model equation (4.76) being significant
at 5 per cent level. The model for cracking and
crazing was developed as (Fig.4.70)

$$(a_c)_p = 2.7363 - 0.1565C + 0.0069C^2 \qquad (4.77)$$

$$0 < C < 25$$

where, $(a_c)_p$ is the predicted value for equation
4.77 and C is the amount of cracking and crazing
(%) in the section of road. Comparing observed data
$(a_c)_o$ and predicted data gave

RTAs AND POTHOLES, PATCHES, DEPRESSIONS, UPHEAVALS

$(a_p)_P = 2.7988 - 0.0627P + 0.0039P^2$

o smoothed data

FIG 4.71 RELATION BETWEEN RTAs/10⁶ VEH-KMs AND POTHOLES, PATCHES, DEPRESSIONS, UPHEAVALS KIGANJO-NANYUKI ROAD

COMPARISON OF OBSERVED AND PREDICTED DATA

$(a_p)_o = 0.9998(a_p)_p - 0.0002$

RTAs AND EDGE SPALLING

$(a_E)_P = 1.4311 + 0.1146E - 0.0011818E^2$

FIG. 4.72 RELATION BETWEEN RTAs/10⁶ VEH-KMs AND EDGE SPALLING KIGANJO-NANYUKI ROAD

COMPARISON OF OBSERVED AND PREDICTED DATA

$(a_E)_o = 1.001(a_E)_p - 0.002$

255

$$(a_c)_o = 0.998 (a_c)_p + 0.005$$

with $r = 0.50$, $r^2 = 0.25$ and standard error of 0.428.
This model related to cracking and crazing equation
4.77 was found to be statistically significant at
the 10 per cent level. The model for relating
potholes, patches, depressions and upheavals to
RTAs was developed as (Fig.4.71)

$$(a_p)_p = 2.7988 - 0.0627P + 0.0039P^2 \qquad (4.78)$$

$$0 < P < 25$$

where, $(a_p)_p$ is the predicted value using equation
4.78 and P is amount of potholes, patches,
depressions and upheavals in % for the road section
where RTAs are observed. Comparing observed data
$(a_p)_o$ and predicted data yielded the equation

$$(a_p)_o = 0.999(a_p)_p - 0.0002$$

with $r = 0.84$, $r^2 = 0.70$ and standard error of 0.318.
The relationship described by equation 4.78 was
found to be statistically significant at the 5 per
cent level. Finally, for pavement defects the model
derived relating edge spalling to RTAs per 10^6
vehicle-kilometres is given by the equation (Fig.4.72)

$$(a_E)_p = 1.4311 + 0.1146E - 0.0011818E^2 \qquad (4.79)$$

$$0 < E < 100$$

where $(A_E)_p$ is the predicted number of RTAs per 10^6 vehicle-kilometres at a given per cent of edge spalling (E) in a road section. Observed and predicted data yielded the equation

$$(a_E)_o = 1.001(a_E)_p - 0.002$$

with $r = 0.92$, $r^2 = 0.84$, standard error of 1.018 where, $(a_E)_o$ is observed data. The relationship described by equation 4.79 was found to be significant at the 5 per cent level.

RTAs per Hour

For the prediction of mean RTAs per 10^6 vehicle-kilometres at a given time of the day, the data for the single carriageway were treated in the same manner as for the dual carriageway. Using harmonic analysis as before models were developed for unsmoothed as well as smoothed data. For the unsmoothed data (Fig.4.73) the model was developed as

$$(a/K)_{p_3} = 0.803 - 0.832 \cos t + 0.357 \cos 2t - 0.327 \cos 3t$$
$$- 0.309 \sin t - 0.103 \sin 2t \qquad \ldots (4.80)$$

PREDICTION OF RTAs WITH TIME OF DAY

$$(a/K)_{P_1} = 0.803 - 0.832\cos t + 0.357\cos 2t - 0.327\cos 3t - 0.309\sin t - 0.103\sin 2t$$

$$(a/K)_{P_2} = 0.704 - 0.456\cos t - 0.02\cos 2t + 0.01\cos 3t - 0.446\sin t + 0.172\sin 2t$$

● UNSMOOTHED DATA   ○ SMOOTHED DATA

COMPARISON OF OBSERVED AND PREDICTED DATA

$$(a/K)_0 = 0.922\,(a/K)_{P_2} + 0.042$$

$$(a/K)_0 = 0.506\,(a/K)_{P_1} + 0.285$$

FIG.4.73 PREDICTION OF MEAN RTAs/10^6 VEH-KMs

KIGANJO-NANYUKI ROAD

where, $(a/K)_{p_3}$ is the predicted value using equation

4.80 at time t. Observed data $(a/K)_o$ was compared

with predicted data and yielded the equation

$$(a/K)_o = 0.506 (a/K)_{p_3} + 0.285$$

with $r = 0.57$, $r^2 = 0.33$ and standard error of 0.734.

For smoothed data the equation was developed as

$$(a/K)_{p_4} = 0.704\ 0.456\ \cos t - 0.02\ \cos 2t + 0.01\ \cos 3t$$

$$-0.446\ \sin t + 0.172\ \sin 2t \qquad .. \quad (4.81)$$

where, $(a/K)_{p_4}$ is predicted data using equation

4.81 . Observed data when compared with the

predicted yielded the equation

$$(a/K)_o = 0.922 (a/K)_{p_4} + 0.042$$

with $r = 0.68$, $r^2 = 0.46$ and standard error of

0.477. Both models (equations 4.80 , 4.81 were

significant at the 5 per cent level.

4.3. Generalised Linear Models

4.3.1 Dual Carriageway

The dependent variable was chosen as the

number of RTAs that had occurred on a given section

of the carriageway during the study period. The
independent variables chosen included the average
daily traffic, vehicles per hour, the percentage of
lorries and buses, junctions per kilometre, horizontal
curve radius, superelevation, longitudinal gradient
and sight distance. These data were observed and
recorded for each section length (in metres) of road
studied (Appendix A.15). The horizontal curve radius
was classified into 3 classes as follows:
0-800m, 800-3200m and >3200m.

Two models were developed. The first model related
to those sections under curves and the second model
related to straight sections. Using the generalised
linear modelling techniques outlined in 3.6 and the
computer program GLIM it was found that the variables
which were reasonably related to RTAs were vehicle-
kilometres (exposure), junctions, lorries and buses,
horizontal curve radius and superelevation. The
estimated values for fitting the model equation 3.74
for the first model are tabulated in Table 4.1. The
estimated values were obtained at cycle 4.

The parameters are interpreted for equation 3.74

$$A = e^k K^b e^{(\Sigma a_1 x_1)}$$

TABLE 4.1   MODEL 1 for DUAL CARRIAGEWAY

Estimate	s.e.	Parameter
-0.09198	0.4942	k
0.7248	0.07462	b
0.0849	0.02295	JPK
0.01904	0.006628	LAB
-0.7595	0.5382	FCR(2)
-0.2860	0.4984	FCR(3)
-0.1692	0.1371	FCR(1).SUE
-0.01629	0.08199	FCR(2).SUE
-0.2639	0.09962	FCR(3).SUE

as follows:

    k   is the constant,

    b   is the power of K,

  JPK   is the coefficient of the junction effect,

  LAB   is the coefficient of the effect of lorries
       and buses,

FCR(2) is the coefficient of the effect of
       horizontal radius (800-3200m),

FCR(3) is the coefficient of the effect of
       horizontal radius (>3200m),

FCR(1).SUE is the coefficient of the effect of
       the interaction of horizontal radius
       (0-800m) and superelevation,

FCR(2).SUE is the coefficient of the effect of horizon-
tal radius (800-3200m) and superelevation and
FCR(3).SUE is the coefficient of the effect of
horizontal radius(>3200m) and superelevation.
For FCR(1), the bottom level of the horizontal radius
(0-800m), the effect is included in the constant term
(e^k). Therefore, the model may be stated as

$$A_1 = e^{-0.09198} \ K^{0.7248} \ e^{(\Sigma a_i x_i)} \qquad (4.82)$$

where the coefficients a_i and the independent
variables x_i are chosen from table 4.1, noting that
A_1 is the predicted number of RTAs for the period
under study and K is the travel in vehicle-kilometres
during the corresponding period. For this model
S.D./d.f. was found to be 1.81.

The parameters of the second model for the dual
carriageway are tabulated in Table 4.2.

TABLE 4.2 MODEL 2 FOR DUAL CARRIAGEWAY

Estimate	s.e.	Parameter
-0.6274	0.2716	k
0.7532	0.07081	b
0.07437	0.02435	JPK
0.02132	0.006333	LAB
67.36	31.50	AR
-0.1213	0.04734	SUE

The parameters are interpreted as before and AR is the reciprocal of the horizontal radius which is equal to zero for a straight road section. Therefore, the second model may be stated as

$$A_2 = e^{-0.6274} \, K^{0.7532} \, e^{(\Sigma a_i x_i)} \qquad (4.83)$$

where A_2 is the predicted number of RTAs for straight sections for the period under study. The remaining terms are as before and the coefficients a_i and independent variables x_i are chosen from Table 4.2. For this second model S.D./d.f was found to be 1.79.

4.3.2 Single Carriageway

The dependent variable was chosen as the number of RTAs that had occurred on a given section of the carriageway during the study period. The independent variables chosen were average daily traffic, vehicles per hour, the percentage of lorries and buses, junctions per kilometre, horizontal curve radius, edge spalling, crazing and cracking, potholes and rutting. These data were observed and recorded for each 500m section (Appendix A.16). The horizontal curve radius was classified as <799m, 800-3099m, >3100m. Edge spalling was classified as <29%, 30%-60%, >60%. Crazing and cracking was classified as <5%, 5%-10%, >10%. Lorries and buses were classified as <28%, >28%. Junctions were classified as <3 junctions,

3-7 junctions, >7 junctions. Using the techniques
outlined in 3.6 the most suitable model was developed
whose parameters are tabulated in Table 4.3.

TABLE 4.3   MODEL FOR SINGLE CARRIAGEWAY

Estimate	s.e	Parameter
-0.1617	0.4770	k
1.138	0.6950	b
0.8154	0.2300	FJ(2)
0.3817	0.4030	FJ(3)
-0.7622	0.2874	FLB(2)
0.6893	0.2863	FCC(2)
0.6791	0.3569	FCC(3)
0.9192	0.3696	FED(2)
0.4871	0.4167	FED(3)
-0.6669	0.3796	FCR(2)
-0.5413	0.3470	FCR(3)

For the single carriageway equation 3.74 therefore
becomes the predictive model

$$A_3 = e^{-0.1617} K^{1.138} e^{(\Sigma a_i x_i)} \qquad (4.84)$$

where A_3 is the predicted RTAs on the single carriage-
way for the period under study, K is the travel in
vehicle-kilometres, the coefficients a_i and the

independent variables x_i are chosen from Table 4.3.
The effects of the bottom classes are included in
the constant term. The independent variables that
were found to be reasonably related to RTAs (Table 4.3)
were:

FJ(2): 3-7 junctions per kilometre,

FJ(3): >7 junctions per kilometre,

FLB(2): <28% lorries and buses,

FLB(3): >28% lorries and buses,

FCC(2): 5%-10% cracking and crazing,

FCC(3): >10% cracking and crazing,

FED(2): 30%-60% edge spalling,

FED(3): >60% edge spalling,

FCR(2): 800-3099m horizontal curve radius,

FCR(3): >3100m horizontal curve radius.

Equation 4.84 was found to have S.D/d.f of 1.42.

4.4   Data Appraisal

At the national level it was found that the
collection of some of the data began only in 1960
and 1973. This had a definite effect on the
calibration of the models developed in this study.

Further there were other extrenous factors that
led to flactuations in RTA trends. These, among others
were the emergency in the early 1950s during the
struggle for independence, independence and the
ensuing socio-economic and political development
in the 1960s and onward, the matatu legislation of
1973, the oil crisis of the early 1970s, the sub-
sequent coffee boom of the middle and late 1970s,
the ban on vehicle importation of the early 1980s
and the subsequent ban on night driving by heavy
vehicles. Modulating the above was the road safety
improvement measures started in the early 1980s to
date. Nonetheless, the data available wereof good
quality and acceptable.

At the micro level on the roads, the location
of RTA spots during coding was affected by missing
files or files pending in court, stored away or
destroyed at the expiry of certain periods of time.
Nonetheless, these difficulties and the inadequacy
of information contained in the Police Form P41 were
solved by the assistance of the local police officers.
It is generally believed that there is under-reporting
of RTAs in Kenya and elsewhere. Further, it is not
mandatory to report to the police RTAs involving
property damage only (the practice of reporting
stopped in 1973 in Kenya). Despite these shortcomings
the data is considered and has been shown to be of good
quality or at least reasonable in deriving RTA patterns in

CHAPTER 5 - DISCUSSION

5.1   Road Traffic Prediction

Firstly, from the models 4.1, 4.2 on human
population growth in Kenya, it was found that human
population is increasing rapidly. The prediction
models obtained in this study were found to be a
good fit for Kenyan data since the slope of observed
versus the predicted regression line was almost
ideal (i.e. unity) and the intercept nearly zero.

Due to rapid population increase, a high temple
of socio-economic development and increased travel,
road traffic is likely to increase rapidly. The
annual rate of population increase was found to be
3.8%- 4.2%. The high variant growth by United Nation
is an over-prediction.

On the assumption that past trends can be used
to predict future motor vehicle levels, assuming the
trends to continue into the future, it was found that
road vehicles are increasing at an annual rate of
6.5%. The predictive models 4.3, 4.4 developed for
the data analysed were found to be a good fit judging
from the slope and the intercept (Fig.4.3). The
prediction was equally consistent as shown by the
correlation coefficient. It is worth noting that
despite using two different limits for the number
of motor vehicles, the results of the two predictions
agree closely.

The level of motorization was found to be in-
creasing at an annual rate of 2.8%. The predictive
model 4.5 was found to be of good fitness, consistency
and calibration (Fig.4.4).

The increase in motor vehicles involved in RTAs
was found to be 3.6% per annum which is growing
approximately as Kenya's human population increase.
The predictive model 4.6
fitted observed data well. The logistic curve was
found to be suited in predicting data related to
population, vehicle fleet and motorization.

The predictive model for the growth in the compo-
sition of cars and utilities was (4.7) found to be the
best compared to the ones for buses/lorries/taxis and
the one for motorcycles. The model for cars and
utilities was nearest to the ideal line (Fig.4.35) and
was very consistent (r = 0.91). The slope for the
model for buses/lorries/taxis was greater than 1
(1.158). Although this may be considered acceptable
the data had very considerable scatter (r = 0.38).
The slope for the model for motorcycles was found
to be acceptable (1.123) and the scatter (r = 0.63)
much better than that for buses/lorries/taxis. The
standard error of the model for cars and utilities
was greater however, than for the other two classes.
The logarithmic curve was found to be the growth curve

suited to describe growth in vehicle composition.

In this study it has been demonstrated that harmonic analysis techniques can be used to predict average vehicle flow per hour at a given time of the day, even with traffic counts taken at 4 hourly intervals (six-ordinate scheme). In fact, for lower accuracy predictions no data smoothing is necessary as the standard errors remain virtually the same. Data smoothing considerably improves the calibration of the model. On the dual carriageway both the smoothed and unsmoothed data models 4.10, 4.11 were found to be of good fitness judging by their slopes and inter- cept (Fig.4.7). They were also consistent, with r=0.93 and 0.91 respectively. The standard errors were large and this may be due to the fact that a lower number of ordinate scheme was used as well as the variations in traffic with time. A lower number ordinate scheme will skip some data points which have high variation leading to over/under estimation. Therefore, using a 12 or 24 ordinate scheme is likely to improve the model. The models 4.12, 4.13 developed for predicting traffic on a low trafficked single carriageway were found to be of very good fit as the slope was very nearly ideal (Fig.4.9). The prediction was very consistent (r = 0.94) for both models and the standard errors were equal. The models for unsmoothed

and smoothed data made almost identical predictions because traffic variation on the single carriageway was low. This implies, therefore, that the interval could be increased from 4 hourly for the traffic counts, as long as some points around the peak flow are observed, and still get good predictions.

All the models discussed above were found to be significant at the 5 per cent level which is the level of statistical significance usually accepted.

5.2   National Road Traffic Accidents Prediction

In this study it was found that injury RTAs are increasing with time at the rate of 3.7% per annum. This is growing at the same rate as that of vehicles involved in RTAs.
The logistic curve

$$A_{P_1} = \frac{8049}{1 + 1.984\ e^{-0.137t}}$$

was found to be a good fit for the data observed. Although the slope of the regression curve was good (0.946) the intercept did not pass through zero (Fig.4.10). This suggests that the model could be improved by calibration. This is likely to have happened because these data, unlike other RTA data that were collected since 1949,were first recorded in 1960. It is likely that earlier data included also non-injury RTAs. However, the prediction was

consistent (r=0.9ł). The standard error was large
(1394). The significance level was found to be 5
per cent. The model may be improved in calibration
by data improvement as well as choice of limit. In
relation to motorization, injury RTAs were found to
increase slowly in the initial stages of motorization,
rising sharply, levelling off and falling. The model
was found to be the cubic polynomial (4.15)

$$A_{p_2} = 6.6981 \times 10^{-8} (V/P) - 0.82263376 (V/P)^2$$

$$+ 270.438684^{V/P} - 15342.4929.$$

The slope for the prediction line for this
model was acceptable (1.135) and the consistency good
(r=0.92) (Fig.4.11). The standard error was large
(1259) and the level of significance was found
to be 10 per cent. This was considered acceptable.
Data smoothing by moving averages as before improved
the shape of the function curve. Although the curve
shows that RTAs will be zero at a level of motoriza-
tion between 250 and 260 vehicles per 10^4 persons
this is not likely to be the case. What is more
likely to be the case is that as the effects of rapid
motorization balance out with those of road safety
improvements the curve will stabilize at some level
in future.

Injury RTAs per vehicle were found to be
decreasing with motorization increase and the relation-
ship described by a cubic polynomial (4.16)

$$(A/V)_p = 0.656 \times 10^{12}(V/P)^3 - 0.00000484(V/P)^2$$

$$+ 0.001059 V/P - 0.0183.$$

The calibration of the predictive model was almost
ideal with slope equal to 1.246 and intercept -0.009
(Fig.4.12). The prediction was consistent although
there was some scatter (r=0.8). The standard error
was small (0.0034) and with the level of significance
of 5 per cent. The data were of good quality and
data smoothing improved the shape of the curve.
However, as earlier mentioned injury RTA data were
limited to the period 1960-83 which was a shortcoming
for long-term prediction purposes.

It was found that injury RTAs per 10^6 vehicle-
kilometres were decreasing at an annual rate of 1.6%
with time. Therefore, as travel increases and there-
fore exposure to risk increases and RTAs per vehicle
decrease this trend is likely. The logistic model
was found to fit these data with slope of prediction
line of 0.89, intercept of 0.168 and some scatter
(r = 0.82). The standard error was small (0.195).
To improve the prediction a model relating this

variable with motorization was developed (4.18)

$$(A/K)_{P_2} = 0.2728 \times 10^{-10} (V/P)^3 - 0.00019613 (V/P)^2$$

$$+ 0.042285 (V/P) - 0.6774$$

which had a better slope (0.956), a smaller intercept
(0.082) and standard error smaller (0.146). The
scatter however, was more considerable (r=0.76)
but both models, were significant at 5 per cent.
Therefore, the cubic polynomial function can predict
injury RTAs per vehicle-kilometre as a function of
motorization. The data smoothing could be improved
by removing the cyclic variation observed through
decomposition of the series.

In this study the analysis showed that casual-
ties from RTAs can be predicted by the logistic
model (4.19)

$$C_{P_1} = \frac{14749}{1 + 3.772\, e^{-0.137t}}$$

Casualties were found to be increasing with time at
the rate of 5.6% per annum. The prediction was
very near the ideal since the slope was 1.014 and
the intercept of 204 was small compared to the
standard error of 3137 which is somewhat large.
However, the consistency of prediction was
good as there was small scatter (r=0.93). The cubic

polynomial prediction model 4.20 improved the
prediction (slope = 1.09) but the intercept dropped)
(-733). Whilst the consistency remained the same
(r=0.93) the standard error dropped (2690). Therefore,
with the significance level for both models of 5 per
cent, the models were considered acceptable. The
logistic model could be improved by improving the
limit and extending the data. The data recording
on casualties like that of injury RTAs begins in
1960. The calibration therefore can be improved.

Casualties per RTA were found to be increasing
at the rate of 1.9% per annum (similar to that of
casualties per 10^4 vehicles).

The prediction was near the ideal
(1.219) but the intercept was not near zero (-0.304)
(Fig.4.15). The calibration could be improved through
revision of limit and data extension and smoothing.
The scatter was acceptable (r=0.87). The standard
error was small (0.159) and the model significant
at 5 per cent level. Considering the limitation
in data the model was acceptable. This result
confirms that found by Jacobs and Hutchinson [1]
for Kenya (an increase of 60% between 1961-1971).

It was found that casualties per RTA decrease
with increase in motorization. The cubic polynomial

model (4.22)

$$(C/A)_{P_2} = 0.11831 \times 10^{-10} (V/P)^3 - 0.00015 (V/P)^2$$

$$+ \; 0.04492 V/P - 1.7667$$

fitted the data with reasonable accuracy. Although the calibration of the model could be improved, the predicted data was scattered around the ideal line (slope = 1.355, intercept = -0.488)(Fig.4.16). The consistency was fair (r=0.84) and the standard error small (0.146). Again, as observed earlier the curve is not likely to decrease to zero but rather it is likely to stabilize at some point in the future. The relationship described by the model was found to be significant at 5 per cent level. Therefore, the pre-dictive model is acceptable.

This study showed that casualties per 10^4 vehicles is increasing at the rate of 1.7% per annum (Fig.4.17) (Similar to casualties per RTA). After data smoothing variation were still afflicting the shape of the curve. The logistic curve (4.23) was much flatter. The predicted data were scattered around the ideal curve with slope = 0.713 and intercept of 145. The calibration could be improved by removal of the variations and revision of limit. The scatter was considerable (r = 0.52) but the standard error small considering the variation in the series

(72.47). The model was found to be significant at the 5 per cent level.

It was further found that casualties per 10^4 vehicles are decreasing with increase in motorization. The cubic polynomial (4.24)

$$(C/V)_{P_2} = 8.8191 \times 10^{-8}(V/P)^3 - 0.04464166(V/P)^2$$

$$+ 12.597057 V/P - 337.6139$$

was a good fit for this data. The prediction was near the ideal (slope = 1.022, intercept =-8.841) (Fig.4.18). Although there was scatter in the predicted data (r = 0.79) it was found that the accuracy of prediction was good. The calibration was therefore acceptable. The standard error (71.44) improved only slightly above the logistic model. The cubic model was found to be significant at 5 per cent level.

During the study it was found that data on RTA deaths can be definitely fitted by a logistic curve (4.25)

$$D_{P_1} = \frac{1720}{1 + 22.889 \ e^{-0.174t}}$$

to show the trend in road deaths over time. It was

further found that RTA deaths in Kenya are increasing
at a high rate of 7.4% per annum nearly twice
the increase in population. The predictive model
was very near the ideal (slope = 0.994 and intercept
= -22)(Fig.4.19). The consistency of prediction was
good (r =0.97). The standard error was found to be
somewhat large (553). The level of significance
was found to be 5 per cent and therefore the prediction
is reasonable.

It was further found that road deaths in Kenya
will first increase, stabilize and then begin to
decrease with motorization. The cubic polynomial
(4.26)

$$D_{P_2} = 229.7235 - 11.960418 V/P + 0.14040617 (V/P)^2$$

$$+ 0.18958 \times 10^{-6} (V/P)^3$$

was found to be a very reasonable fit for this trend.
Predictions for Kenyan data were also made using
Smeed's equation (2.2) [2] and the results compared
in Table 5.1. The Agoki formula came nearest to the
ideal (Fig.4.20) observed = predicted line. The
consistency in prediction is of the same quality.

TABLE 5.1   COMPARISON OF OBSERVED AND PREDICTED RTA DEATHS IN KENYA

Formula	Units	Regression		Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error
		Slope	Intercept			
Agoki	Deaths	1.009	-10	0.97	0.94	545
Smeed	Deaths	1.751	-459	0.96	0.93	310

Although the Smeed formula had a lower standard error,
the slope and intercept suggest that the calibration
could be improved. The Smeed formula over-predicts
for lower rates of death and at higher rates it under-
predicts. Smeed formula correctly predicts at around
500 deaths.

The study showed further, that deaths per 10^4
persons in Kenya increase slowly at first, then sharply,
stabilize, and then may start to fall slowly. The cubic
polynomial (4.27)

$$(D/P)_{P_1} = 1.9788 \times 1^{-10}(V/P)^3 + 0.00001494(V/P)^2$$
$$+ 0.007762 V/P - 0.4127$$

was found to be a good fit for Kenyan data. Predictions
were also made on the same data using Smeed formula
(2.3) [3] and Jacobs and Hutschinson formula (2.4)[1].
The comparison of the three formulae is shown on
Table 5.2.

TABLE 5.2 COMPARISON OF OBSERVED AND PREDICTED RTA DEATHS PER 10^4 PERSONS

Formula	Units	Regression		Correla. Coeffi.,r	Determin. Coeffi.,r^2	Std. Error
		Slope	Inter.			
Agoki	D/P	0.923	0.054	0.93	0.87	0.31
Smeed	D/P	4.809	-2.545	0.91	0.83	0.06
Jacobs & Hutchinson	D/P	3.550	-1.172	0.92	0.85	0.08

Again, the Agoki formula (Fig.4.21) came nearest to the ideal observed = predicted line although it had highest standard error. All formulae have the same consistency in predicting data but the slopes and intercepts of the Smeed, Jacobs and Hutchinson formulae suggest that calibration is needed. Jacobs and Hutchinson formula predict Kenyan data correctly around D/P = 0.4 and the Smeed formula at D/P = 0.6. All models are significant at 5 per cent level.

It was further found that RTA deaths per 10^4 vehicles in Kenya rose slowly at first, rapidly and are likely to stabilize in future and drop as motorization increases. Again, the cubic polynomial function (4.28)

$$(D/V)_{P_1} = 1.4218 \times 10^{-8} (V/P)^3 - 0.00022871 (V/P)^2$$

$$+ 0.577548 V/P - 5.4171$$

was found to be a good fit for the data for Kenya.

Again, the model derived in this study (Fig.4.22)
was compared with those developed by Smeed (2.1)
[1] and Jacobs and Hutchinson (2.5) [1]. The
comparison of the three formulae is shown on Table
5.3.

TABLE 5.3 COMPARISON OF OBSERVED AND PREDICTED RTA DEATHS PER
10^4 VEHICLES

| Formula | Units | Regression | | Correla. | Determ. | Std |
		Slope	Intercept	Coeffi.,r	Coeffi.,r^2	Error
Agoki	D/V	0.769	13	0.77	0.59	15
Smeed	D/V	-0.876	111	-0.70	0.49	12
Jacobs & Hutchinson	D/V	-2.013	151	-0.72	0.51	5

Again, the Agoki formula came nearest to the ideal
prediction line. The Agoki and Smeed formulae had
the same order of standard error and Jacobs and
Hutchinson had the smallest standard error. The
slopes and intercepts of the Smeed, Jacobs and
Hutchinson formulae suggest strongly a need for their
calibration. The Agoki formula could be improved in
calibration particularly with improved data reporting
and keeping for both road deaths and vehicle fleet and
of course more accurate population census and prediction
(although this is not critical considering the trend
prediction for population discussed earlier). All
models were statistically significant at the 5 per
cent level.

With respect to deaths per 10^6 vehicle-kilo-
metres the analysis revealed an increasing trend with
time at the rate of 1.8% per annum, a rate similar
to that shown by casualties per RTA. The data was
fitted using the logistic curve model 4.29 and
the prediction proved consistent although there was
some scatter (r=0.84). The prediction was near the
ideal line (slope = 0.864, intercept = 0.032)
(Fig.4.23). The standard error was small (0.060)
and with a significance level of 5 per cent the
prediction is reasonable. Despite data smoothing
the curve was still afflicted with flactuations.
This difficulty may be overcome if data were available
for decomposition of this time series. With further
data smoothing calibration might be improved.

As seen in earlier trends, deaths per 10^6
vehicle-kilometres has risen slowly at the start,
sharply then stabilized before beginning to drop.
On a long term basis this trend is to continue. A
cubic polynomial (4.30) fitted the curve with a
somewhat poorer fit than the logistic curve. Although
the data were predicted around the ideal line the slope
was only 0.763. The intercept was good (0.053)
(Fig.4.24) and comparable to the previous model. The
standard error was equally small (0.059). The calibra-
tion could be improved by more accurate data on vehicle-
kilometres. The amount of travel is somewhat difficult

to observe or predict with great accuracy. The model
was nonetheless, significant at the 5 per cent level.
This model is reasonable for prediction.

RTA injuries were found to be increasing with
time at the rate of 5.8% per annum. This is about
the same rate as for the increase in casualties.
This would
imply that they are of a greater influence in the
growth of casualties than do deaths. The trend was
fitted by the logistic curve (4.31) for predicting
the trend in the growth of injuries. The prediction
was found to be nearly ideal (slope = 1.045, inter-
cept = -171) (Fig.4.25). The prediction was consistent
(r = 0.97) but the standard error large (3430). The
cubic polynomial (4.32)

$$I_{P_2} = 1.2305 \times 10^{-6} (V/P)^3 + 0.25629308 (V/P)^2$$

$$+ 60.580992 V/P - 4849.778$$

predicted the data satisfactorily and indicated that
injuries have tended to increase with motorization.
The prediction was still near the ideal line with
slope = 0.961 but the intercept nearly trebled
shifting to 344. The prediction remained consistent
(r = 0.94) and the standard error dropped somewhat
(3144). Data smoothing is likely to improve the
calibration of the model. Both models were found

to be significant at 5 per cent level and accordingly
were accepted as predicting the number of injuries
from RTAs.

With respect to injuries per 10^4 persons it was
found that the trend was an increasing one with time
at the rate of 5.2% per annum. The prediction model
(4.33) found to predict this trend was the logistic
curve. The prediction was very nearly ideal (slope=1.047
intercept = -0.149) (Fig.4.26). The prediction was
consistent (r = 0.91) and the standard error small.
With a significance level of 5 per cent, this prediction
model was accepted. The data after smoothing was found
to be still with variations which would imply a need
for further smoothing to improve the calibration of
the model. The reporting of injury data is likely
to have affected the shape of the curve. A polynomial
function (4.34) of the third degree was found to fit
the same data in relation to motorization. It was
found that injuries per 10^4 persons increases slowly
initially, rapidly next then stabilizes before starting
to fall as motorization increases. The prediction was
near ideal (slope = 0.885, intercept = 0.764) (Fig.4.27).
The consistency was fair (r = 0.87) and the standard
error small (1.742). The two models were significant
at 5 per cent level. The polynomial model had a better
curve shape meaning that the data smoothing was very
effective.

Injuries per 10^4 vehicles was found to grow at
a rate of 1.8% per annum. It was observed that the
smooth data curve (4.35) was flatter than some of the
earlier curves. This implies that injuries per 10^4
vehicles are stabilizing with time as borne out by
the growth rate and the fact that the predictions
were scattered around the centre of the prediction line.
The prediction was found to be fairly close to the
ideal line (slope = 0.734, intercept = 117) (Fig.4.28).
The prediction was not very consistent as the scatter
was very considerable (r = 0.44). The standard error
was not large considering the scatter. The model
was statistically significant at 5 per cent level.
The slope, the intercept and the scatter suggest
that the model could be improved. This might be done
by further data smoothing and the revision of the
limit. The variation in injuries per 10^4 vehicles
was found to fit the cubic polynomial function
(4.36) with motorization as the independent variable.
It was further found that even for Kenya as motorization
increases injuries per 10^4 vehicles decreases, a finding
which is at variance with the earlier finding by Jacobs
and Hutchinson [1]. Their finding was based on data
limited to only 10 years (1961-71). Therefore, it is
to be noted that for both national and international
comparisons long term trends based on records kept
for a considerable period are more reliable for
studying RTA patterns of different countries . The

predictive model was nearly ideal (slope = 1.103, intercept = -55) (Fig.4.29). The standard error of 56 was comparable to that obtained for the logistic model, for this phenomenon, above. The model was significant at 5 per cent level. This model is quite suited for prediction and therefore acceptable.

In this study it was further found that injuries per 10^6 vehicle-kilometres are increasing with time at the rate of 2.4% per annum. Further, that the trend can be described by the logistic curve (4.37). The prediction was close to ideal (slope = 0.924, intercept = 0.084) (Fig.4.30). However, the scatter was considerable (r = 0.69). The standard error was small (0.239). It was further found that as motorization increases injuries per 10^6 vehicle-kilometres decrease the trend was described by the cubic polynomial function (4.38). The prediction was very nearly ideal (slope = 1.02, intercept = -0.052) (Fig.4.31). The scatter of r = 0.77 showed that the prediction was not very consistent. The standard error of 0.242 was rather large for the phenomenon being predicted but comparable to the previous logistic model. Both models were significant at 5 per cent level. The trend revealed that injuries per 10^4 vehicle-kilometres have stabilized and are decreasing. Both models could be improved in terms of calibration by further data smoothing and observation of more reliable data on vehicle-kilometres.

The severity index was found to be increasing
with time at the rate of 3.0% per annum. The predictive
model 4.39 was found to be the logistic curve but
with a flatter shape. There appeared to be fluctua-
tions even after data smoothing. This led to a
rather considerable scatter in the prediction (r=0.53).
The slope was 0.643, the intercept was 3.9 (Fig.4.32)
and the standard error 1.65. The predictions are
reasonable but the slope and intercept suggested
necessary improvement in the calibration. The model
was found to be significant at 5 per cent level.

It was found that severity index increases slowly
with motorization and then rapidly. The cubic polyno-
mial function (4.40) was found to fit the data. The
prediction was fairly close to the ideal (slope =0.833,
intercept = 1.655) (Fig.4.33). The scatter had lessened
(r = 0.63) compared to that of the logistic prediction.
The standard error was 1.508 rather like that of the
logistic model. Both models were found to be signifi-
cant at 5 per cent level. The severity indices for
various classes of road users were analysed. Due to
lack of sufficient data no meaningful functional
relations either with time or motorization were
suitable. But the analysis revealed that the severity
index of pedestrians was postively correlated with
motorization. Therefore, pedestrians would seem to
be the main category of road users affecting severity
index in RTAs. The finding by Jacobs and Hutchinson

that the lower the vehicle ownership level, the greater
the severity index is contradicted by the finding in
this study.

The percentage distribution of RTAs by day and
night in Kenya was found to be described by a logarith-
mic trend curve (4.41)

$$\left({}^{(\%)}A_d \right)_p = 81.073 - 5.656 \quad \ell nt$$

over time. This curve was significant at 5 per cent
level and for the data observed the prediction was
about the ideal line (slope = 0.754, intercept = 17.004)
(Fig.4.34).

Due to flactuations remaining in the smoothed data
there was much scatter (r=0.45). The standard error was
4.674 which is considered reasonable. The trend revealed
that the day proportion is decreasing whilst the night
proportion is increasing with a tendency to stabilize
at around 50%. Nonetheless about $2/3$ of RTAs in Kenya
are still occurring during the day in line with the
proportion of day traffic. Therefore, equal efforts
in reducing RTAs should be exerted to both day as well
as night RTAs.

The percentage RTA responsibility of various
classes of vehicles was found to be described by the

logarithmic trend (4.42-4.47) curve over time. The
results for the various classes of vehicle type are
summarized in Table 5.4. It was found that the
prediction for pedal cyclists responsibility came
nearest the ideal line. The prediction for pedal
cyclists responsibility was the most consistent with
some scatter however ($r = 0.76$). The others had
reasonable calibration but with considerable scatter.
When responsibility and composition (Fig.4.35-4.38)
distribution were compared it was found that cars
and utilities were responsible for RTAs in about the
same proportion as their composition. Indeed their
responsibility was just below the composition. Not
surprisingly, buses, lorries and taxis were found to
be responsible for nearly twice as much as their
composition with an increasing tendency in their RTA
involvement. Their composition was found to be stable.
Because this class of vehicles carries passengers it
has agreat effect on casualty, injury, death and RTA
rates. Concerted efforts in RTA reduction should
therefore be directed towards this class of vehicles as
there is much potential for safety improvement. Although
motorcycles were involved in RTAs more than their compo-
sition initially the stabilizing tendency was found to
reveal that they are being involved in RTAs less than
their composition. The responsibility by handcarts
and animals as well as by pedal cyclists has remained more
or less constant in recent years.

TABLE 5.4   COMPARISON OF OBSERVED AND PREDICTED PERCENTAGE RTA RESPONSIBILITY

Vehicle Type	Units	Regression		Correlation Coefficient,r	Determination Coefficient,r^2	Standard Error	Significance Level
		Slope	Intercept				
Cars,Utilities	%	0.787	14.661	0.43	0.18	1.590	5
Buses,Lorries,Taxis	%	0.767	5.894	0.53	0.29	2.190	5
Motorcycles	%	0.868	0.727	0.30	0.09	0.453	10
Pedal cyclists	%	0.979	0.040	0.76	0.57	1.655	5
Handcarts & animals	%	0.596	0.983	0.15	0.02	0.331	25
Pedestrians & Pass.	%	0.736	7.496	0.53	0.28	7.498	5

TABLE 5.5   COMPARISON OF OBSERVED AND PREDICTED % DISTRIBUTION OF PERSONS KILLED

Road user	Units	Regression		Correlation Coefficient,r	Determination Coefficient,r^2	Standard Error	Significance Level
		Slope	Intercept				
Drivers	%	3.147	-30.145	0.45	0.45	0.398	10
Motorcyclists	%	0.630	0.884	0.36	0.13	1.303	20
Pedestrians	%	0.925	3.197	0.45	0.20	1.484	10
Passengers	%	1.389	-11.236	0.21	0.04	0.319	30
Pedal cyclists	%	0.420	3.161	0.38	0.15	2.235	20

Of great concern was the finding that pedestrian and passenger responsibility is continuing to increase despite growth in motorization. Measures in safety improvement for these unprotected class of road users has not kept pace with increased travelling. This is yet another class of road users with much potential for RTAs reduction. Efforts should be directed towards this group both in terms of education as well as design for pedestrian facilities on Kenyan roads and changes in transportation of passengers. The models were found to predict responsibility and composition reasonably well. Their slopes and intercept indicate that their calibration could be improved. There was much fluctuation in motorcycle, buses/lorries/taxis and pedestrian/passenger data, which if smoothed out could greatly improve the calibration of predictive models related to these categories of road users.

The growth in percentage distribution of those killed and injured above 16 years of age was found to be logarithmic (4.49). However, the prediction of data for the distribution of those killed was not found to be statistically significant. The distribution predictive model for those injured was however, found to be significant at 20 per cent. These data had only been observed for a decade and it is not valid to extend them beyond a decade for long term extra-

polative purposes. Nonetheless, they are useful in indicating short term trends. The prediction for the distribution of those injured came nearer the ideal (slope = 0.856, intercept = 13.117) (Fig.4.39). There was considerable scatter (r = 0.40). The standard error was small (1.808). If the data is extended for this model it could prove useful for predictive purposes. More useful information on age distribution could be gained if age grouping of those killed and injured was widened much in line with the grouping used in the analysis for the dual as well as the single carriageways data.

The percentage distribution of those persons killed in RTAs in Kenya was found to be described by the logarithmic trend curve (4.50-4.54) over time. The results for the various road user categories are summarized in Table 5.5. Since these predictions were about the ideal line all the results could be accepted as indicative of the distribution characteristics of those killed. Because data was of short duration it was found that it had very considerable scatter even after smoothing. For indicative purposes the levels of significance found were reasonable. Hence, judging the models by slope and intercept criteria revealed that the prediction for pedestrians came nearest the ideal line, followed by that for passengers and then the one for motorcyclists.

In terms of trends it was found that they were all
stable. However, it was observed that nearly 80%
of those killed are pedestrians and passengers
followed by 14% as drivers. Therefore, in order to
reduce RTAs efforts should be directed towards
drivers as they have responsibility for themselves
as well as for passengers (35% of total killed).
Therefore, drivers were found to be potentially
responsible for nearly 49% of those killed whereas
pedestrian have a share of 45%. This result indicates
that concerted efforts in road safety improvements
should be directed towards the drivers and pedestrians.
Again, passenger transportation is another area of
potential improvement if RTA deaths are to be
reduced. This confirms earlier predictions by
responsibility distribution. For long term predictions
these data need to be extended by further observations
and subsequent recalibration. This may be said to be
true of all the predictive models obtained in this
study.

Finally, for the national RTA trends in Kenya,
the percentage distribution of those persons injured
in RTAs was found to be described by the logarithmic
trend curve (4.55-4.59) over time. The results for
the various road user categories are summarized in
Table 5.6. It was found that the prediction for
motorcyclists and pedal cyclists were not statistically

significant. Again, since all the predictions were about the ideal line (i.e. they all intersected the ideal line at least) they could be accepted as indicative of the distribution characteristics of those injured. The prediction that came nearest the ideal line was that for drivers (slope 1.516, intercept = -8.112) (Fig.4.41). It however requires further calibration after data extension. The scatter was found to be considerable (r = 0.59). The standard error lay between those models found to be statistically significant. Again, for purposes of RTAs reductions efforts should be directed towards driver training, retraining and education. Also efforts should be directed towards passenger carrying regulations particularly for public transport vehicles like matatus and the buses as well as open lorries.

5.3 Prediction of Effect of Road Factors on Road Traffic Accidents

In this study it was found that the effect of longitudinal gradient on RTAs can be predicted by a quadratic polynomial function. Predictions were also made on the same data using Silyanov formula (2.14) [4] and compared (Table 5.7). For both upgrade and downgrade data the Agoki formulas (4.60, 4.61)

TABLE 5.6   COMPARISON OF OBSERVED AND PREDICTED % DISTRIBUTION OF PERSONS INJURED

Road User	Units	Regression		Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error	Significance Level
		Slope	Intercept				
Drivers	%	1.516	- 8.112	0.59	0.35	1.650	5
Motorcyclists	%	0.052	4.202	-0.02	0	0.673	NONE
Pedal cyclists	%	0.057	6.175	-0.01	0	0.529	NONE
Pedestrians	%	0.486	10.677	0.30	0.09	3.460	20
Passengers	%	3.448	-127.706	0.22	0.05	0.268	30

TABLE 5.7   COMPARISON OF OBSERVED AND PREDICTED RTAs PER 10^6 VEHICLE KILOMETRES

Formula	Units	Regression		Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error	Significance Level
		Slope	Intercept				
Agoki (Upgrade)	RTAs/10^6V–K	0.811	0.455	0.68	0.46	0.651	20
Silyanov (Upgrade)	RTAs/10^6V–K	0.025	1.956	0.02	0	0.565	NONE
Agoki (Downgrade)	RTAs/10^6V–K	1.018	-0.095	0.97	0.94	0.853	5
Silyanov (Downgrade)	RTAs/10^6V–K	-1.180	3.584	-0.97	0.97	0.738	5

$$(a_g)_{P_1} = 0.9866 + 1.10666g - 0.18401g^2$$

$$(a_g)_{P_2} = 2.993 + 0.11g - 0.05165g^2$$

came nearer the ideal observed = predicted line. The
slopes and intercepts for the Silyanov formula
suggest that they need recalibration if they are to
interpret Kenyan or similar data. The Agoki formula
for downgrade gradients came nearest the ideal line
and therefore can be used in predicting the effect of
gradient on RTA occurrence. The Agoki formula for
upgrade gradient is a good fit also (slope = 0.811,
intercept = 0.455) (Fig.4.47, 4.48) but the prediction
is not as consistent as that for downgrade. It has
scatter (r = 0.68) and the level of significance of
20 per cent. This could be acceptable since RTAs
are dependent on many causative factors. Silyanov
[4] found out that the number of RTAs increased conti-
nuously with increase in grade particularly being
sharp at 3 per cent. The majority of the vehicles
from Silyanov's result were moving downwards. In
this study it was found that for upgrade gradients
3 per cent had the worst RTA incidence. For the
downgrade the worst gradients are ones less than 3
per cent. The general finding was that flatter
grades were more RTA prone than steeper ones. The
reason for this may lie in the fact that the steeper

the gradient the more careful and alert the drivers become. On the upgrade vehicular speeds reduce with increase in gradient thus better holding possibilities and better control of vehicles is enhanced. Design criteria [14] recommend maximum ranges of gradient for flat country as 3-5 per cent, rolling country as 4-7 per cent and mountainous as 7-10 per cent. These ranges fall largely in the decreasing section of the models developed in this study. This implies that the design of flatter gradients than 3 per cent is crucial to safety particularly when combined with other factors of sight distance and horizontal align-ment. Further, attempts to achieve grades less than 3 per cent imply increase in construction costs which must be justified by increased traffic safety due to reduced RTAs.

The effect of sight distance on RTAs was also found to be predictable by a quadratic polynomial function. The comparison of prediction performance qualities of the models developed in this study was made with that of Silyanov (Table 5.8). The Agoki formulae (4.62, 4.63)

$$(a_s)_{P_1} = -8.745 + 0.07981S - 0.000136\ S^2$$

$$(a_s)_{P_2} = -1.8551 + 0.03467S - 0.00006\ S^2$$

TABLE 5.8 COMPARISON OF OBSERVED AND PREDICTED RTAs/10^6 VEHICLE - KILOMETRES

Formula	Units	Regression		Correlation Coefficient,r	Determination Coefficient,r^2	Standard Error	Significance Level
		Slope	Intercept				
Agoki (Nairobi-to-Thika CWY)	RTAs/10^6V-K	0.723	0.826	0.83	0.69	0.996	5
Silyanov (Nairobi-to-Thika CWY)	RTAs/10^6V-K	-1.734	5.609	0.85	0.72	0.424	5
Agoki Thika-to-(Nairobi CWY)	RTAs/10^6V-K	0.550	1.057	0.74	0.55	0.441	5
Silyanov (Thika-to-Nairobi CWY)	RTAs/10^6V-K	-0.473	3.463	0.59	0.35	0.326	5

TABLE 5.9 COMPARISON OF OBSERVED AND PREDICTED RTAs/10^6 VEHICLE KILOMETRES

Formula	Units	Regression		Correlation Coefficient r	Determination Coefficient r^2	Standard Error	Significance Level
		Slope	Intercept				
Agoki	RTA/10^6V-K	0.883	1.603	0.88	0.78	1.317	5
Silyanov	RTAs/10^6V-K	15.92	-11.293	0.88	0.78	0.073	5

came nearer the ideal line but the consistency, though

better than in Silyanov's, showed scatter. Silyanov's

formulae are unsuited for predicting the study data

and judging from the slopes and intercept (4.49, 4.50)

they need recalibration. The Nairobi-to-Thika carriage-

way model had the highest standard error. Silyanov

found that RTAs occurred on road sections where sight

distance was less than 300 metres. In this study it

was found that sections with 300 metres experience

the worst RTA occurrence. RTAs were found to decrease

for shorter or longer distances than 300m. The possible

explanation here is that at distances greater than

300 drivers have a greater chance of avoiding a RTA

and at distances less than 300m drivers are more care-

ful and speeds are more moderate since other limiting

factors such as gradient and curvature come into play.

Design standards [14, 16] recommend minimum sight

distances, on level roads in Kenya for design speeds

60-120 kph, ranging from 80-310 metres. These standards

are comparable to those used in Great Britain. USA and

Australia have lower standards. From this study it can

be seen that where roads have many sections which fall

within this range RTAs increase with sight distance

to reach a maximum at 300m of sight distance for those

sections between 300 and 500m RTAs decrease with increase

in sight distance. The MOTC [14] standard specifying

300m sight distance for 120 k.p.h. for stopping and

60 k.p.h. for passing requires revision since at this
level of sight distance the horizontal alignment as
well as the vertical alignment appear to be contribu-
ting the most in worsening the safety situation.

In this study it was found that RTAs decrease with
increase in carriageway width. A linear function
(4.64)

$$(a_w)_{p_1} = 32.6439 - 4.2348W$$

was found to be describing this relationship. This
finding was found to be consistent with that of
Silyanov [4] where he observed that RTAs per vehicle-
kilometre becomes markedly sharp when the width is less
than 7 metres. The two models were compared (Table 5.9),
Silyanov's equation being equation (2.12) [4].
The Agoki formula came nearer the ideal line than
Silyanov's although Silyanov's had a much lower
standard error. The consistency in prediction is
comparable for both models. Silyanov's formula
grossly under-predicts when used for Kenyan data.
The slope and intercept for Silyanov's formula
prediction suggest improvement in calibration for
Kenyan data. It is to be noted that the comparison
shows that RTA rates on Kenyan roads are as high as
15 times the rates in Europe for which Silyanov
developed the model. Alternatively seen, there is

potential for RTA reduction on Kenyan roads if carriageway widths are increased and present roadways routinely maintained including the shoulders.

On an urban and semi-urban dual carriageway, the effect of junctions per kilometre was found to be linear (4.65, 4.66).

$$(a_j)_{P_1} = 5.068 + 1.672j$$

$$(a_j)_{P_2} = 3.0099 + 2.9239j$$

RTAs per 10^6 vehicle-kilometres were found to increase with increase in number of junctions per kilometre. On the single carriageway rural road with a higher number of junctions and accesses per kilometre (than the dual carriageway urban and semi-urban road with control and restriction of access) it was found that the effect of junctions is non-linear. It was found to be of a quadratic polynomial function (4.75)

$$(a_j)_{P} = 0.6668 + 1.1082j - 0.1288j^2$$

The prediction by the models developed in this study was compared with that of Jacobs formula (equation (2.9) [4]). Table 5.10 shows the comparison.

TABLE 5.10 COMPARISON OF OBSERVED AND PREDICTED RTAS/10^6 VEHICLE-KILOMETRES

Formula	Units	Regression		Correlation Coefficient,r	Determination Coefficient,r^2	Standard Error	Significance Level
		Slope	Intercept				
Agoki (Nairobi-to-Thika CWY)	RTAs/10^6V-K	1.001	-0.004	0.92	0.84	2.649	5
Jacobs (Nairobi-to-Thika CWY)	RTAs/10^6V-K	1.639	2.691	0.92	0.84	2.346	5
Agoki (Thika-to-Nairobi CWY)	RTAs/10^6V-K	0.970	-0.696	0.93	0.86	6.316	5
Jacobs (Thika-to-Nairobi CWY)	RTAs/10^6V-K	2.781	-1.808	0.808	0.93	2.578	5
Agoki (Kiganjo-Nanyuki Rd)	RTAs/10^6V-K	1.004	-0.009	0.98	0.96	0.642	5
Jacobs (Kiganjo-Nanyuki Rd)	RTAs/10^6V-K	-0.050	2.670	-0.19	0.04	2.498	NONE

Jacobs [4] found, for the Nairobi-Mombasa Road, that
junctions per kilometre were the most significant
independent variable. The critical coefficient (r^2)
for the model for his data was found to be only 0.49
meaning that the model explained only 49 per cent
of the variation. This is rather too low. He states
that there were never more than two junctions per
kilometre and hence an addition of one junction per
kilometre led to RTA increase of over one RTA per 10^6
vehicle-kilometre. From Table 5.10 it is confirmed
that junctions per kilometre are a very significant
independent variable in relation to RTA causation.
The models developed in this study were all significant
at 5 per cent level. The prediction came very near
the ideal. For all intents and purposes they could
be said to be perfect. The slopes and intercepts for
the Nairobi-to-Thika carriageway and the Kiganjo-
Nanyuki Road together with their intercepts reveal that
the models are very well calibrated. The worst condi-
tion on the single carriageway is 4 junctions. Thus
RTAs per 10^6 vehicle-kilometres increase with increase
in junctions per kilometre. After 4 junctions per
kilometre they decrease with increase in junctions per
kilometre. This is to be expected because the fewer
the junctions the greater the mobility, free flow
conditions and hence the higher the speed. The more
the junctions the greater the restriction to flow,the
slower the vehicle and the safer. With reference to the

comparison of the observed and predicted data by Jacobs formula, noting that the slope of the prediction line is nearly zero it can be assumed to be parallel to the predicted axis. Therefore the prediction by Jacobs formula on the study data varies in the observed value (the RTAs/10^6 vehicle-kilometres axis) only to a great extent meaning that it is actually tracing out the locus of the effect of junctions on RTAs. Thus it confirms the finding that the relationship is quadratic. At 4 junctions per kilometre Nairobi-to-Thika carriage-way has an RTA rate of nearly 12 RTAs per 10^6 vehicle-kilometres, the Thika-to-Nairobi carriage has 15 and the single carriageway has 3. Thus in relation to junctions the single carriageway low trafficked road is the safest. The two carriageways of the dual carriage-way road can be said to be quite dangerous to the extent of 5 times. The reasons for this high rate on the dual carriageway is likely to be speed, poor visibility, poor illumination at night and the high level of traffic. The developed models may be said with all certainty to be predictive of the phenomenon of RTAs. Jacobs formula failed in predicting the study data.

The effect of horizontal curves was found to be non-linear. The upgrade curves affected the occurrence of RTAs as a cubic polynomial function, while the effect of downgrade curves was of a

quadratic nature. The two models developed in this study were compared with Silyanov's model (equation (2.13) [4]. Table 5.11 depicts the comparison. The Agoki formulae (4.67, 4.68)

$$(a_R)_{P_1} = 3.346 - 0.0009R + 0.77 \times 10^{-7}$$

$$- 1.361 \times 10^{-12} R^3$$

$$(a_R)_{P_2} = 6.8699 - 0.0031R + 0.000\ 000397R^2$$

came nearest to the ideal in terms of prediction. In particular the prediction by the model developed from upgrade data (the cubic function) predicts well but has some scatter ($r = 0.67$) (Fig.4.55). The standard errors are reasonable. The formula by Silyanov could be improved in terms of calibration. Silyanov formula predicts data for downgrade curves in Kenya judging by the slope and intercept. However it tends to under-predict and could be improved by calibrating the intercept in particular. Silyanov found that the most dangerous horizontal curves were those of less than 500 metres. By Silyanov's formula (2.13) this radius gives a rate of 2 RTAs per 10^6 vehicle-kilometres. Using this rate on the models developed here implies that for downgrade gradient curves the most dangerous are those less

TABLE 5.11 COMPARISON OF OBSERVED AND PREDICTED RTAs PER 10^6 VEHICLE-KILOMETRES

Formula	Units	Regression		Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error	Significance Level
		Slope	Intercept				
Agoki (Upgrade)	RTAs/10^6V–K	1.011	–0.055	0.67	0.44	0.509	5
Silyanov (Upgrade)	RTAs/10^6V–K	0.194	2.112	0.37	0.14	1.467	20
Agoki (downgrade)	RTAs/10^6V–K	0.844	0.555	0.88	0.77	1.909	5
Silyonav (downgrade)	RTAs/10^6V–K	0.827	2.270	0.73	0.53	1.625	5

TABLE 5.12 COMPARISON OF OBSERVED AND PREDICTED RTAs PER 10^{-6} VEHICLE-KILOMETRES

Formula	Units	Regression		Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error	Significance Level
		Slope	Intercept				
Agoki-Kiganjo/ Nairobi-to-Thika	RTAs/10^6V–K	0.966	0.065	0.97	0.94	0.937	5
Jacobs-Kig./ Nbi-Tka	RTAs/10^6V–K	0.415	0.126	0.95	0.90	2.132	5
Silyanov-Kig/ Nbi-Tka	RTAs/10^6V–K	6.782	–1.351	0.96	0.93	0.132	5
Agoki-Kig/ Tka-Nbi	RTAs/10^6V–K	0.921	0.110	0.96	0.91	1.679	5
Jacobs Kig/ Tka-Nbi	RTAs/10^6V–K	0.669	–0.635	0.92	0.84	2.220	5
Silyanov-Kig/ Tka-Nbi	RTAs/10^6V–K	10.969	–3.038	0.93	0.87	0.138	5

than 1500 metres and for upgrade those that are
2000 metres. This would seem to imply that drivers
in Kenya take curves at higher speeds for the same
curves taken in Europe. In fact curves less than
200 metres were found to be extremely dangerous
particularly when occuring at crests. Design
criteria [14] recommend for design speeds ranging
between 40-120 kph curve radius ranges 60-1000m.
Speeds of 90-100 k.p.h. require about 500m radius
curves [14]. These speeds are the design range for
the study roads. From this study it is seen that
all these curves, particularly when negotiated at
high speeds, lie in the RTA prone zone. It is not
unthinkable to imagine that drivers in Kenya particu-
larly with high powered engines negotiate curves at
speeds much higher than 120 kph. As a measure for RTA
reduction the models suggest that for downgrade curves
it is RTA saving to design curves greater than 1500
metres. This would be true also for upgrade curves.
For downgrade curves 4000 metre radius appeared to be
the safest. Not surprisingly however after this
radius RTAs begin to rise with increase in radius.
The influence would seem to be that of high speeds·
and steeper gradients.

The effect of superelevation on RTAs was found
to be virtually the same for upgrade as well as
downgrade curves. It was found that RTAs decrease

with increased superelevation. Taken on the face of
it would seem to imply that small radii curves are
safer. This would contradict the finding from
horizontal curve effect. Rather, what is likely
is that smaller radii curves are taken cautiously
by drivers thus reducing RTAs on high superelevation
curves. Further, an attempt to use larger radii
for horizontal curves must take into consideration
the cost implications versus the saving to be
realized from a drop in RTAs particularly on curves.
Quadratic polynomial functions were found to fit this
effect of superelevation (4.69, 4.70),

$$(a_\alpha)_{P_1} = 2.2729 + 0.6\alpha - 0.098\alpha^2$$

$$(a_\alpha)_{P_2} = 1.0693 + 1.3132\alpha - 0.22\alpha^2.$$

The predictions were near ideal and consistent. The
level of significance was 5 per cent and therefore
the models were acceptable. The standard errors
were small.

In this study it was found that whereas RTAs
per 10^6 vehicle-kilometres increase with increase in
vehicle flow (average vehicles per hour) the growth
in RTAs follows the logistic curve (4.71, 4.72)

$$(a_q)_{p_1} = \frac{18.89}{1 + 24.128\ e^{-0.002q}}$$

$$(a_q)_{p_2} = \frac{24.55}{1 + 38.983\ e^{0.0003q}}$$

This was true for data from the single carriageway
combined with each of the carriageways of the
Nairobi-Thika Road. The models were compared with
those of Jacobs (2.8) and Silyanov (2.11) [4] in
Table 5.12. All the models were consistent in their
prediction and significant at 5 per cent level. The
Agoki and Silyanov formulae had lower errors than
Jacobs formula. Of all the models the Agoki models
came closest to the ideal line. Thus Jacobs formula
that is linear over-predicts on the assumption that
the effect of vehicle flow is linear, Silyanov's
under-predicts as the non-linear quadratic function
rises sharply. The Agoki formula predicts right
rising slowly first then sharply to level off as
traffic saturates. This finding fitted the logistic
curve model. This finding was further enhanced by
the fact that from harmonic analysis RTAs per 10^6
vehicle-kilometres and vehicle flow varied similarly
with time of day to reach the peak i.e. at low traffic
flow RTAs are low and reach peaks much the same way.
These models were found to predict the phenomenon

TABLE 5.13   COMPARISON OF OBSERVED AND PREDICTED RTAS/10^6 VEHICLE – KILOMETRE

Formula	Units	Regression Slope	Intercept	Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error	Significance level
Nairobi-Tka Rd Unsmoothed	RTAs/10^6V-K	1.358	-0.026	0.77	0.59	0.201	5
Smoothed	"	0.979	-0.020	0.87	0.75	0.315	5
Kiganjo-Nyki Rd Unsmoothed	"	0.506	0.285	0.57	0.33	0.734	5
Smoothed	"	0.922	0.042	0.68	0.46	0.477	5

TABLE 5.14   COMPARISON OF OBSERVED AND PREDICTED RTAs/10^6 VEHICLE-KILOMETRES

Formula	Units	Regression Slope	Intercept	Correlation Coefficient, r	Determination Coefficient, r^2	Standard Error	Significance Level
Rutting	RTAs/10^6V-K	0.998	0.006	0.94	0.88	0.785	5
Cracking	"	0.998	0.005	0.50	0.25	0.428	10
Potholes	"	0.999	0	0.84	0.70	0.318	5
Edge Spalling	"	1.001	0.002	0.92	0.84	1.018	5

well and therefore were found acceptable.

The variation of mean RTAs per 10^6 vehicle-kilometres, for both carriageways, was found to vary harmonically with time of the day (4.74, 4.81).

$$(a/K)_{P_2} = 0.538-0.267 \cos t-0.085 \cos 2t+0.033 \cos 3t$$

$$- 0.319 \sin t - 0.095 \sin 2t,$$

$$(a/K)_{P_4} = 0.704-0.456 \cos t-0.02 \cos 2t+0.01 \cos 3t$$

$$- 0.446 \sin t + 0.172 \sin 2t.$$

The comparison of the predictions for unsmoothed and smoothed data for the two study roads is shown in Table 5.13.

The formulae obtained after data smoothing by the moving averages technique improved both the calibration and prediction consistency particularly for low volume flow roads. For both study roads models developed using smoothed data came nearer the ideal line. The standard errors also dropped after data smoothing although the level of significance remained unchanged. The models were accepted for prediction.

Using data from the single carriageway the effect of pavement defects was found to be significant. The

The comparison of the prediction models is tabulated
in Table 5.14. The effects were found to be non-linearly
related to RTA occurence and the quadratic function
gave the best fit. The prediction that came nearest
the ideal (4.76 - 4.79, Fig.4.69 - 4.72) observe =
prediction line was found to be that of edge spalling.
Edge spalling, rutting, and potholes (together with
patches, depressions and upheaval) were found to be
more significant (5 per cent level) than cracking
and crazing (significant at 10 per cent). Overally,
rutting, cracking and potholes behave in the same
manner in their effect on RTA occurrence. At the
initial stages they have a reducing effect on RTA
rates. After a minimum is reached they tend to
increase the incidence of RTAs as they rise. Edge
spalling acts in a counter manner to rutting,
cracking and potholes. RTAs increase with increasing
edge spalling to reach a maximum and then decrease
with further increase in edge spalling. It is to be
noted that for newly surfaced roads, when these
independent variables are zero, driving speeds are
considerably higher than after pavement defects have
taken their toll. At this stage any small rutting
causes the greatest rise in RTA occurence (4 RTAs
per 10^6 vehicle-kilometres), followed by cracking
and potholes (3 RTAs/10^6 vehicle-kilometres) then
edge spalling (nearly 1 RTA per 10^6 vehical-kilometres).

The former three then have a decreasing effect and edge
spalling which trigers RTAs predominates up to a level
when it is nearly 40% causing as high as 4 RTAs per
10^6 vehicle-kilometres. At 100% edge spalling the
road becomes virtually unpaved hence speeds reduce and
consequently RTAs reduce. Maintenance criteria for
paved roads [5] considers for all road bases that
for no cracks, rutting of less then 10mm the pavement
is good and acceptable. It should be noted that at this
stage RTAs are at a level of 3 RTAs per 10^6 vehicle-
kilometres. On the basis of the same criteria rutting
of less than 25mm without cracks is critical. From
this study, at this critical value rutting has the
minimum effect and RTAs are more influenced by the
other pavement defects. On the basis of the same
criteria reconstruction is recommended when the rate
of potholes is more than 40 per 100 metres. If this
is considered to be equivalent to 100% potholes then
patching and overlaying is recommended at 15-40 holes
per 100 metres. Taking the lower limit (15) patching
commences at 38% potholes. At this stage, from the
analysis in this study, this pavement defect is already
contributing close to 5 RTAs per 10^6 vehicle-kilometres.
Therefore poor road maintenance has a considerable
effect on RTA rates. A reduction in RTAs could be
obtained by proper road maintance coupled with speed
regulatory and related safety enforcement measures. These
models are reasonable in their predictions and therefore
acceptable.

5.4 Generalised Linear Models

5.4.1 Dual Carriageway

The two models developed for the dual carriageway indicated that on an interactive basis junctions per kilometre, the percentage of heavy vehicles, horizontal radius and superelevation have a far greater influence on RTAs than longitudinal gradient and sight distance. The first model further showed that the effect of superelevation depends on horizontal curve radius as would be expected. For the first model (equation 4.82)

$$A_1 = e^{-0.09198} K^{0.7248} e^{(\Sigma a_i x_i)}$$

S.D./d.f was found to be 1.81. This value is reasonably close to 1 and therefore the model is quite reasonable in predicting RTAs.
The second model developed for straight sections (equation 4.83)

$$A_2 = e^{-0.6274} K^{0.7532} e^{(\Sigma a_i x_i)}$$

was found to have S.D./d.f. = 1.79. This is quite a good model as $S.D./d.f.$ is also tending to 1. Junctions and heavy vehicles increase RTA risk. Horizontal radius has the following effect on risk:

for radius 300m the effect = exp.(67.36x $\frac{1}{300}$ = 1.24,

for radius 3000m the effect =exp.(67.36x $\frac{1}{3000}$= 1.02

for straight section the effect = 1.00, for example.

The risk decreases as the radius increases.

The result is very interesting and is in the direction
one could imagine.

5.4.2 Single Carriageway

The model developed for the single carriageway
(equation 4.84) indicated that on an interactive basis
the best variables that had any meaning when considering
RTAs were junctions per kilometre, heavy vehicles,
crazing and cracking, edge spalling and horizontal
curve radius. Potholes, patches, depressions and
upheavals as well as rutting did not have as signi-
ficant an influence on RTAs.

Considering equation 4.84

$$A_3 = e^{-0.1617} K^{1.138} e^{(\Sigma a_i x_i)}$$

and the parameters in Table 4.3, given that the effect
of 3 or less junctions per kilometre is 1.00 (bottom
value), the effect for 3-7 junctions per kilometre
was found to be 2.26 (exp.(00.8154)) and that for 7
or more junctions 1.45 (exp.(0.3817)). This means
that risk increases 2.26 if junctions are in the
second class and 1.45 if in the third class. For
lorries and buses the risk decreases from 1.00
(for <28%) to 0.47 (for >28%). This would imply

that there is not much variation and there are
correlations. Therefore more data is required to
be able to see the real effect. For crazing and
cracking, if the effect for the first level of
<5% is 1.00, for 5%-10% being exp(0.6893) which is
1.99 and for the third level of >10% being exp(0.6791)
which is 1.92, this would mean that risk increases
about 1.9 if crazing and cracking is above or equal
to 5%. Similarly for edge spalling if the effect for
the class <29% is 1.00 (bottom class) and that of
30%-60% 2.51 (exp(0.9192)) and that of >60% 1.63
(exp(0.4871)) then risk increases for edge spalling
for the two classes in the order of 2.51 and 1.63
respectively. For horizontal curve radius if the
risk is 1.00 for radius <799m then for 800-3099m is
0.51 (exp(0.3796)) and for >3100m is 0.58 (exp(0.3470))
meaning that when the radius is above 799m the risk
is about 0.5 times compared with the radius
below 800m. The model for the single carriageway
(equation 4.84) had S.D/d.f. of 1.42. This model is
quite good since 1.42 is close to 1 and therefore
quite acceptable in finding out the effect of
various independent variables on RTAs.

CHAPTER 6 - CONCLUSIONS

The objectives of this study were: to study
road traffic accidents (RTAs) in Kenya and determine
where possible their fundamental characteristics
and causal factors related to their occurrence, to
develop predictive models for Kenya at the National
(macro) level to be used for the monitoring of RTAs
and the performance of safety improvement programmes
and lastly to develop predictive models for some
selected Kenyan roads at the road (micro) level
to assist in the proper understanding of the behaviour
of RTAs in relation to the design elements. The
results of this study indicated that:

1)   RTA phenomenon lends itself to mathematical
      modelling and in particular the characteristic
      patterns of RTAs in different countries, and
      in particular Kenya, can be predicted provided
      long term accurate data for the particular
      country in question exist. At the macro level
      the logistic model is well suited in predicting
      the growth of RTAs with time, the logarithmic
      trend curve is well suited in predicting the
      growth in the distribution of RTA responsibility
      and involvement while the polynomial function
      is suited in predicting the trend of RTAs in
      relation to motorization. At the macro level

the Smeed relationships and those developed
later by Jacobs and Hutchinson do not satisfacto-
rily predict RTA phenomena in Kenya. The
corresponding models developed in this study
performed best overall. To predict RTA deaths
the formula developed in this study as follows
was (equation (4.26)):-

$$D_{P_2} = 229.7235 - 11.960418 \; {}^V/_P + 0.14040617 ({}^V/_P)^2$$
$$+ 0.18958 \times 10^{-6} ({}^V/_P)^3$$

with   r=0.97, r^2=0.94, slope = 1.009, intercept

   =-10 which is better than Smeed's
formula (2.2)

$$D = 0.0003 \; (VP)^{\frac{1}{3}}$$

with r=0.96, r^2=0.93, slope=1.751 and

       intercept=-459. To predict RTA
deaths per 10^4 persons the formula developed
from the (equation (4.27)

$$(D/P)_{P_1} = 1.9788 \times 10^{-10} (V/P)^3 + 0.00001494 (V/P)^2$$

       $$+0.007762 \; V/P - 0.4127$$

with r=0.93, r^2=0.87, slope=0.923, intercept=0.054,
which is better than both Smeeds formula (2.3)

$$D/P = 0.0003(V/P)^{\frac{1}{3}}$$

with r=0.91, r^2=0.83, slope=4.809, intercept=-2.545

and Jacobs and Hutchinson's formula (2.4)

$$D/P = 0.00077(V/P)^{\frac{3}{5}}$$

with r=0.92, r^2=0.85, slope=3.550 and intercept

=-1.172. To predict RTA deaths per 10^4

vehicles the model developed here as the equation

(4.28)

$$(D/V)_{p_1}=1.4218\times10^{-8}(V/P)-0.00022871(V/P)^2$$

$$+ 0.577548 \ V/P - 5.4171$$

with r=0.77, r^2=0.59, slope=0.769, intercept=13,

is again better than both Smeed's formula (2.1)

$$D/V = 0.0003 \ (V/P)^{\frac{-2}{3}}$$

with r=-0.70, r^2=0.49, slope=-0.876, intercept=111

and Jacobs and Hutchinson's formula (2.5)

$$D/V = 0.00077 \ (V/P)^{\frac{-2}{5}}$$

with r=-0.72, r^2=0.51, slope=-2.013 and intercept=151.

2.   At the micro level polynomial functions of the
first, second and third degree were found to be
suited in predicting the effects of road factors
on RTA rates, the logistic curve is well suited
in predicting the growth of RTAs in relation
to vehicle flow whilst the variations in RTAs
and vehicle flow with time of day can be predicted
by harmonic functions. The models developed

in this study gave highest correlation between
predicted and actual values. The relationships
developed by Silyanov and later by Jacobs do
not satisfactorily predict RTA phenomena on
Kenyan roads. In particular to predict the
effect of upgrade and downgrade gradients
respectively, the models developed in this
study viz,

$(a_g)_{p_1}$ = 0.9866 + 1.10666g - 0.18401g^2 (for

upgrades)

with r=0.68, r^2=0.46, slope=0.811, intercept=0.455
and $(a_g)_{p_2}$ =2.993 + 0.11g - 0.05165g^2(for downgrades)

with r=0.97, r^2=0.94, slope=1.018,intercept=-0.o95,
were better than Silyanov's formula (2.14)

a_g = 0.265 + 0.105g + 0.0229g^2

with r=0.02, r^2=0, slope=0.025, intercept=1.956
for upgrades and r=-0.97,r^2=0.94, slope=-1.180
and intercept=3.584 for downgrades. To predict
the effect of sight distance one of the models
developed in this study viz.

$(a_s)_{p_1}$ =-8.75 + 0.079 815 - 0.0001365^2

with r=0.83, r^2=0.69, slope=0.723, intercept=0.826
performed better than Silyanov's formula (2.15)

$$a_d = 1/(0.200 + 0.00111d + 0.0000009d^2)$$

with r=0.85, r^2=0.72, slope=-1.734 and intercept=5.609. To predict the effect of carriageway width the model developed from this study, viz.

$$(a_w)_{p_1} = 32.6439 - 4.2348W$$

with r=0.88, r^2=0.78, slope=0.883,intercept=1.603 performed better than Silyanov's formula (2.12)

$$a_w = 1/(0.173W - 0.21)$$

with r=0.88, r^2=0.78, slope=15.92 and intercept=-11.293. To predict the effect of junctions using the model developed from the single carriageway data viz,

$$(a_j)_p = 0.6668 + 1.1082j - 0.1288j^2$$

with r=0.98, r^2=0.96, slope=1.004, intercept=-0.009, performed better than Jacob's formula (2.9)

$$a_j = 1.45 + 1.02j$$

with r=-0.19, r^2=0.04, slope=-0.050 and intercept=2.670. To predict the effect of horizontal radius the model developed in this study for upgrade curves viz.

$$(a_R)_{p_1} = 3.346 - 0.0009R + 0.77 \times 10^{-7}R^2 - 1.361 \times 10^{-12}R^3$$

with r=0.67, r^2=0.44,slope=1.011,intercept=-0.055, performed better than Silyanov's formula (2.15)

$$a_R = 0.647 + 723/R - 649.5/R^2$$

with r=0.37,r^2=0.14,slope=0.194 and intercept=2.112. To predict the effect of vehicle flow one of the models developed in this study, viz,

$$(a_q)_{P_1} = \frac{18.89}{1 + 24.128e^{-0.002q}}$$

with r=0.97, r^2=0.94, slope=0.966,intercept=0.065, performed best compared with Jacobs formula (2.8)

$$a_q = 0.116 + 0.009q$$

with r=0.95,r^2=0.90,slope=0.415,intercept=0.126 and Silyanov's formula (2.11)

$$a_q = 0.256 + 0.000408q + 1.36 \times 10^{-7}q^2$$

with r=0.96,r^2=0.93,slope=6.782 and intercept = -1.351.

3. RTAs do not occur by chance but are causally related to some characteristic factors on the road environment, vehicle or road user. Further, that these relationships can be described by mathematical models.

4.   Generalized linear models are very beneficial
     when trying to study the various effects of
     traffic and geometrical design elements on RTAs.
     Further, that the interactions between various
     variables may be determined from these models.
     With increased data availability such models
     can be improved in order to form predictive
     models for a wider range of road and traffic
     conditions.

5.   There is potential for RTA reduction in any
     country, and in particular Kenya, and the
     performance of improvement schemes can be
     monitored by predictive models similar to
     those developed in this study.

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NOTATIONS

A_o observed number of RTAs

A_{P_1} predicted number of RTAs by the logistic curve model

A_{P_2} predicted number of RTAs as a function of motorization

$(A/K)_o$ observed RTAs per 10^6 vehicle - kilometres

$(A/K)_{P_1}$ predicted RTAs per 10^6 vehicle - kilometres by the logistic curve model

$(A/K)_{P_2}$ predicted RTAs per 106^6 vehicle - kilometres as a function of motorization

$(A/V)_o$ observed RTAs per motor vehicle

$(A/V)_P$ predicted RTAs per motor vehicle as a function of motorization

C_o observed number of casualties i.e. sum of injured and killed

C_{P_1} predicted casualties by the logistic curve model

C_{P_2} predicted casualties as a function of motorization

$(C/A)_o$ observed casualties per RTA

$(C/A)_{P_1}$ predicted casualties per RTA by the logistic curve model

$(C/A)_{P_2}$ predicted casualties RTA as a function of motorization

$(C/V)_o$ observed casulties per 10^4 vehicles

$(C/V)_{P_1}$ predicted casualties per 10^4 by the logistic curve model

$(C/V)_{P_2}$ predicted casualties per 10^4 as a function of motorization

D_o observed RTA deaths

D_{P_1} predicted deaths by the logistic curve model

D_{P_2} predicted deaths as a function of motorization

$(D/K)_o$ observed deaths per 10^6 vehicle - kilometres

$(D/K)_{P_1}$ predicted deaths per 10^6 vehicle - kilometres by the logistic curve model

$(D/K)_{P_2}$ predicted deaths per 10^6 vehicle - kilometres as a function of motorization

$(D/P)_o$ observed deaths per 10^4 persons (population)

$(D/P)_{P_1}$ predicted deaths per 10^4 persons as a function of motorization

$(D/P)_{P_2}$ predicted deaths per 10^4 persons by Jacobs & Hutchinson Formula

$(D/P)_{P_3}$ predicted deaths per 10^4 persons by Smeed formula

$(D/V)_o$ observed deaths per 10^4 vehicles

$(D/V)_{P_1}$ predicted deaths per 10^4 vehicles as a function of motorization

$(D/V)_{P_2}$ predicted deaths per 10^4 vehicles by Jacobs & Hutchinson formula

$(D/V)_{P_3}$ predicted deaths per 10^4 vehicles by Smeed formula

H_o observed human population

H_{p_1} predicted human population by high growth by logistic curve model

H_{p_2} predicted human population by low growth by logistic curve model

I_o observed injuries from RTAs

I_{p_1} predicted injuries by the logistic curve model

K vehicle - kilometres (amount of travel as measured by products of vehicle flow and distance travelled)

$(I/K)_o$ observed injuries per 10^6 vehicle - kilometres.

$(I/K)_{p_1}$ predicted injuries per 10^6 vehicle - kilometres by the logistic curve model

$(I/K)_{p_2}$ predicted injuries per 10^6 vehicle - kilometres as a function of motorization

$(I/P)_o$ observed injuries per 10^4 persons

$(I/P)_{p_1}$ predicted injuries per 10^4 persons by the logistic curve model

$(I/P)_{p_2}$ predicted injuries per 10^4 persons as a function of motorization

$(I/V)_o$ observed injuries per 10^4 vehicles

$(I/V)_{p_1}$ predicted injuries per 10^4 vehicles by the logistic curve model

$(I/V)_{p_2}$ predicted injuries per 10^4 vehicles as a function of motorization

$(V_A)_o$ observed RTA vehicles

$(V_A)_p$ predicted RTA vehicles by the logistic curve model

V_o observed registered vehicles

V_{p_1} predicted registered vehicles by high growth by logistic curve model

V_{p_2} predicted registered vehicles by low growth by logistic curve model

$(V/P)_o$ observed vehicles per 10^4 persons (motorization)

$(V/P)_p$ predicted motorization by logistic curve model

ρ_o observed severity index (deaths/casualties as a percentage)

ρ_{p_1} predicted severity index by logistic curve model

ρ_{p_2} predicted severity index as a function of motorization

$\left((\%)_{A_d}\right)_o$ observed percentage of RTAs occurring during daylight hours

$-\left((\%)_{A_d}\right)_o$ predicted percentage of RTAs occurring during daylight hours

$\left((\%)_{BLT}\right)_o$ observed percentage composition of buses, lorries and taxis

$\left((\%)_{BLT}\right)_p$ predicted percentage composition of buses, lorries and taxis

$\left({}^{(\%)}_{b\ell T'}\right)_o$ observed percentage responsibility of buses, lorries and taxis

$\left({}^{(\%)}_{b\ell T'}\right)_p$ predicted percentage responsibility of buses, lorries and taxis

$\left({}^{(\%)}_{CU}\right)_o$ observed percentage composition of cars and utilities

$\left({}^{(\%)}_{CU}\right)_p$ predicted percentage composition of cars and utilities

$\left({}^{(\%)}_{cu}\right)_o$ observed percentage responsibility of cars and utilities

$\left({}^{(\%)}_{cu}\right)_p$ predicted percentage responsibility of cars and utilities

$\left({}^{(\%)}_{b'}\right)_o$ observed percentage responsibility of pedal cyclists

$\left({}^{(\%)}_{b'}\right)_p$ predicted percentage responsibility of pedal cyclists

$\left({}^{(\%)}_{ha}\right)_o$ observed percentage responsibility of handcarts and animals

$\left({}^{(\%)}_{ha}\right)_p$ predicted percentage responsibility of handcarts and animals

$\left({}^{(\%)}_{wp}\right)_o$ observed percentage responsibility of pedestrians and passengers

$\left({}^{(\%)}_{wp}\right)_p$ predicted percentage responsibility of pedestrians and passengers

$\left({}^{(\%)}_{M}\right)_o$ observed percentage composition of motorcycles

$\left({}^{(\%)}_{M}\right)_p$ predicted percentage composition of motorcycles

$\left({}^{(\%)}D_{16+} \right)_o$ observed percentage of those killed above age 16

$\left({}^{(\%)}D_{16+} \right)_p$ peredicted percentage of those killed above age 16

$\left({}^{(\%)}D_{B'} \right)_o$ observed percentage distribution of pedal cyclists killed

$\left({}^{(\%)}D_{B'} \right)_p$ predicted percentage distribution of pedal cyclists killed

$\left({}^{(\%)}D_{D'} \right)_o$ observed percentage distribution of drivers killed

$\left({}^{(\%)}D_{D'} \right)_p$ predicted percentage distribution of drivers killed

$\left({}^{(\%)}D_{M} \right)_o$ observed percentage distribution of motor cyclists killed

$\left({}^{(\%)}D_{M} \right)_p$ predicted percentage distribution of motor cyclists killed

$\left({}^{(\%)}D_{P} \right)_o$ observed percentage distribution of passengers killed

$\left({}^{(\%)}D_{P} \right)_p$ predicted percentage distribution of passengers killed

$\left({}^{(\%)}D_{W} \right)_o$ observed percentage distribution of pedestrians killed

$\left({}^{(\%)}D_{W} \right)_p$ predicted eprcentage distribution of pedestrians killed

$\left({}^{(\%)}I_{16+} \right)_o$ observed percentage of those injured above age 16

$\left({}^{(\%)}I_{16+} \right)_p$ predicted percentage of those injured above age 16

$\left(^{(\%)}I_{B'}\right)_O$	observed percentage distribution of pedal cyclists injured.
$\left(^{(\%)}I_{B'}\right)_P$	predicted percentage distribution of pedal cyclists injured
$\left(^{(\%)}I_{D'}\right)_O$	observed percentage distribution of drivers injured
$\left(^{(\%)}I_{D'}\right)_P$	predicted percentage distribution of drivers injured
$\left(^{(\%)}I_M\right)_P$	observed percentage distribution of motor cyclists injured
$\left(^{(\%)}I_M\right)_P$	predicted percentage distribution of motor cyclists injured
$\left(^{(\%)}I_P\right)_O$	observed percentage distribution of passengers injured
$\left(^{(\%)}I_P\right)_P$	predicted percentage distribution of passengers injured
$\left(^{(\%)}I_W\right)_O$	observed percentage distribution of pedestrians injured
$\left(^{(\%)}I_W\right)_P$	predicted percentage distribution of pedestrians injured
$(a_C)_O$	observed RTAs/10^6 vehicle-km associated with cracking and crazing
$(a_C)_P$	predicted RTAs/10^6 vehicle-km associated with cracking and crazing
$(a_E)_O$	observed RTAs/10^6 vehicle-km associated with edge spalling
$(a_E)_P$	predicted RTAs/10^6 vehicle-km associated with edge spalling

$(a_g)_o$ observed RTAs/10^6 vehicle-km associated with longitudinal gradient

$(a_g)_p$ predicted RTAs/10^6 vehicle-km associated with longitudinal gradient

$(a_j)_o$ observed RTAs/10^6 vehicle-km associated with junctions per km

$(a_j)_p$ predicted RTAs/10^6 vehicle-km associated with junctions per km

$(a/K)_o$ observed RTAs/10^6 vehicle-km on a road section

$(a/K)_p$ predicted RTAs/10^6 vehicle-km on a road section

$(a_p)_o$ observed RTAs/10^6 vehicle-km associated with potholes, upheavals and depressions

$(a_p)_p$ predicted RTAs/10^6 vehicle-km associated with potholes, upheavals and depressions

$(a_q)_o$ observed RTAs/10^6 vehicle-km associated with vehicle flow per hour

$(a_q)_p$ predicted RTAs/10^6 vehicle-km associated with vehicle flow per hour

$(a_R)_o$ observed RTAs/10^6 vehicle-km associated with horizontal curve radius

$(a_R)_p$ predicted RTAs/10^6 vehicle-km associated with horizontal curve radius

$(a_{r'})_o$ observed RTAs/10^6 vehicle-km associated with rutting

$(a_{r'})_p$ predicted RTAs/10^6 vehicle-km associated with rutting

$(a_s)_o$	observed RTAs/10^6 vehicle-km associated with sight distance
$(a_s)_p$	predicted RTAs/10^6 vehicle-km associated with sight distance
$(a_W)_o$	observed RTAs/10^6 vehicle-km associated with road width
$(a_W)_p$	predicted RTAs/10^6 vehicle-km associated with road width
$(a_\alpha)_o$	observed RTAs/10^6 vehicle-km associated with superelevation
$(a_\alpha)_p$	predicted RTAs/10^6 vehicle-km associated with superelevation
C	pavement cracking and crazing (%)
E	pavement edge spalling (%)
g	Longitudinal gradient (%)
j	junctions per km
P	potholes in pavement (%)
q	vehicles/hour
r'	pavement rutting (mm)
R	horizontal curve radius (m)
S	sight distance (m)
t	time (hours, years etc)
α	superelevation (%)
r	correlation coefficient
r^2	coefficient of determination (critical coefficient)
$S_{y.x}$	standard error of estimate
e	base for natural logarithms

APPENDIX A.1

MANUAL TRAFFIC COUNT AT MUTHAIGA 30TH-31ST OCTOBER

VEHICLE COMPOSITION			(%AGE)	
P			L	
12542/17782 : 70.5%			1986/17782 : 11.2%	
B			M	
925/17782 : 5.2%			2330/17782 : 13.1%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
1881	185	111	151	2328
%AGE 80.8%	7.9%	4.8%	6.5%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
1881	185	111	151	2328
%AGE 10.6%	1.0%	0.6%	0.8%	13.0%

MANUAL TRAFFIC COUNT AT MUTHAIGA 31ST OCT-1ST NOV

VEHICLE COMPOSITION			(%AGE)	
P			L	
12234/16707 : 73.2%			1687/16707 : 10.1%	
B			M	
635/16707 : 3.8			2151/16707 : 12.9%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
2117	173	126	140	2556
%AGE 82.8%	6.8%	4.9%	5.5%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
2117	173	126	140	2556
%AGE 12.7%	1.0%	0.75%	0.84%	15.3%

MANUAL TRAFFIC COUNT AT MUTHAIGA 1ST-2ND NOVEMBER

VEHICLE COMPOSITION			(%AGE)	
P			L	
11090/14900 : 74.4%			1141/14900 : 7.7%	
B			M	
768/14900			1901/14900 : 12.7%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
2875	134	182	193	3384
%AGE 85%	4%	5.3%	5.7%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
2875	134	182	193	3384
%AGE 25.9%	11.6%	23.7%	10.1%	22.6%

MANUAL TRAFFIC COUNT AT MUTHAIGA 2ND-3RD NOVEMBER

VEHICLE COMPOSITION			(%AGE)	
P			L	
8234/10939 : 75.3%			470/10939 : 4.3%	
B			M	
786/10939 : 4.3%			1449/10939 : 13.2%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
2185	144	242	240	2811
%AGE 77.8%	5.1%	8.6%	8.5%	100%
NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
2185	144	242	240	2811
%AGE 26.5%	30.6%	30.3%	16.3%	25.6%

MANUAL TRAFFIC COUNT AT RUIRU 31ST OCT-1ST NOV

VEHICLE COMPOSITION			(%AGE)	
P			L	
5200/9145 : 57.8			1878/9145 : 20.6	
B			M	
534/9145 : 5.8			1445/9145 : 15.8	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
850	246	120	104	1320
%AGE 64.4%	18.6%	9.1%	7.9%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
850	246	120	104	1320
%AGE 16.2%	13.1%	23.4%	7.3%	14.6%

MANUAL TRAFFIC COUNT AT RUIRU 1ST-2ND NOVEMBER

VEHICLE COMPOSITION			(%AGE)	
P			L	
5812/10099 : 57.6%			1462/10099 : 14.5%	
B			M	
632/10099 : 6.2%			2493/10099 : 21.7%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
1132	60	124	51	1467
%AGE 77.2%	10.9%	8.5%	3.4%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
1132	160	124	51	1467
%AGE 18.9%	10.8%	19.1%	2.3%	14.4%

MANUAL TRAFFIC COUNT AT RUIRU 2ND-3RD NOVEMBER

VEHICLE COMPOSITION			(%AGE)	
P			L	
5546/8649 : 64.1%			640/8649 : 7.4%	
B			M	
717/8649 : 8.3%			1746/8649 : 20.2%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
1157	166	137	84	1544
%AGE 74.9%	10.8%	8.9%	5.4%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
1157	166	137	84	1564
%AGE 21.1%	26.6%	19.2%	4.9%	18.3%

MANUAL TRAFFIC COUNT AT RUIRU 3RD-4TH NOVEMBER

VEHICLE COMPOSITION			(%AGE)	
P			L	
5929/11407 : 52.0%			2015/11407 : 17.7%	
B			M	
874/11407 : 7.6%			2589/11407 : 22.7%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
919	530	146	75	1670
%AGE 55.0%	31.8%	8.7%	4.5%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
919	530	146	75	1670
%AGE 15.4%	26.3%	16.6%	2.9%	14.6%

A.1/3

MANUAL TRAFFIC COUNT AT RUIRU 4TH-5TH NOVEMBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
6048/11580 : 52.3%			2004/11589 : 17.3%	
B			M	
966/11580 : 8.3%			2562/11580 : 22.1%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
688	240	115	68	1111
%AGE 6.9%	21.6%	10.4%	6.1%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
688	240	115	68	1111
%AGE 11.4%	12.1%	12.0%	2.7%	9.6%

MANUAL TRAFFIC COUNT AT THIKA 1ST-2ND SEPTEMBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
%AGE 2631/5483 : 48.0%			1175/5483 : 21.4%	
B			M	
%AGE 731/5483 : 13.3%			946/5483 : 17.3%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
583	138	88	88	897
%AGE 65%	15%	10%	10%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
583	138	88	88	897
%AGE 22.1%	13.0%	14.0%	10.0%	162%

MANUAL TRAFFIC COUNT AT THIKA 29TH-30TH OCTOBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
2390/4601 : 52%			1078/4601 : 23.4%	
B			M	
508/4601 : 11.0%			625/4601 : 13.6%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
405	146	62	40	653
%AGE 62.0%	22.3%	9.5%	6.2%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
405	146	62	40	653
17.0%	14%	12.2%	6.2%	14.3%

MANUAL TRAFFIC COUNT AT THIKA 30TH-31ST OCTOBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
2383/4694 : 51%			1199/4694 : 25%	
B			M	
499/4694 : 11%			613/4694 : 13%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
381	162	69	58	670
%AGE 57%	24%	10%	9%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
381	162	69	58	670
16%	13%	14%	10%	14%

MANUAL TRAFFIC COUNT AT THIKA 31ST OCTOBER - 1ST NOVEMBER

VEHICLE COMPOSITION	HOUR (%AGE)		
P		L	
2386/4973 : 48.0%		1251/4973 : 25.2%	
B		M	
605/4973 : 12.1%		731/4973 : 14.7%	

NIGHT TRAFFIC (7.00 pm - 6.30 am)

P	L	B	M	VEHICLES
455	156	88	64	763
%AGE 60%	20%	12%	8%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
455	156	88	64	763
19%	12.4%		9%	15.3%

MANUAL TRAFFIC COUNT AT THIKA 2ND-3RD OCTOBER

VEHICLE COMPOSITION	HOUR (%AGE)		
P		L	
1938/4057 : 47.8%		640/4057 : 15.8%	
B		M	
702/4057 : 17.3%		777/4057 : 19.1%	

NIGHT TRAFFIC (7.00 pm - 6.30 am)

P	L	B	M	VEHICLES
492	87	104	122	805
%AGE 61%	11%	13%	15%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
492	87	104	122	805
%AGE 26.0%	16.0%	17.2%	19.4%	20.2%

MANUAL TRAFFIC COUNT AT THIKA 3RD-4TH NOVEMBER

VEHICLE COMPOSITION	HOUR (%AGE)		
P		L	
2154/4855 : 44.4%		1198%4855 : 24.7%	
B		M	
661/4855 : 4855		842/4855 : 17.3%	

NIGHT TRAFFIC (7.00 pm - 6.30 am)

P	L	B	M	VEHICLES
315	132	61	44	552
%AGE 57%	24%	11%	8%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICELS
315	132	61	44	552
%AGE 14.4%	11%	9.2%	4%	11.3%

MANUAL TRAFFIC COUNT AT THIKA 4TH-5TH NOVEMBER

VEHICLE COMPOSITION	HOUR (%AGE)		
P		L	
2160/5109 : 42.3%		1232/5109 : 24.1%	
B		M	
759/5109 : 14.9%		958/5109 : 18.7%	

NIGHT TRAFFIC (7.00 pm - 6.30 am)

P	L	B	M	VEHICLES
382	154	90	100	726
%AGE 53%	21%	12%	14%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
382	154	90	100	726
%AGE 18%	13%	12%	10%	14.2%

A.1/5

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 29TH-30TH OCTOBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
2543/4301 : 59.1%			812/4301 : 18.9%	
B			M	
278/4301 : 6.5%			668/4301 : 15.5%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
369	132	31	64	596
%AGE 61.9%	22.2%	5.2%	10.7%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
369	132	31	64	596
14.5%	16.3%	11.3%	9.6%	13.9%

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 30TH-31ST OCTOBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
2593/4657 : 55.7%			1082/4657 : 23.2%	
B			M	
224/4657 : 4.8%			758/4657 : 16.3%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
430	221	38	89	778
%AGE 55.3%	28.4%	4.9%	11.4%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
430	221	38	89	778
16.5%	21.01%	17.1%	11.8%	13.9%

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 31ST OCTOBER – 1ST NOVEMBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
3011/4814 : 62.5%			804/4814 : 16.7%	
B			M	
249/4814 : 5.2%			750/4814 : 15.6%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
585	182	42	68	877
%AGE 66.7%	20.8%	4.8%	7.7%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
585	182	42	68	877
19.4%	22.2%	20.2%	8.9%	18.1%

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 1ST-2ND NOVEMBER

VEHICLE COMPOSITION		(%AGE)		
P			L	
3927/6099 : 64.4%			618/6099 : 10.1%	
B			M	
338/6099 : 5.6%			1216/6099 : 19.9%	
NIGHT TRAFFIC (7.00 pm – 6.30 am)				
P	L	B	M	VEHICLES
673	124	65	108	970
%AGE 69.4%	12.8%	6.7%	11.1%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
673	124	65	108	970
%AGE 17.1%	19.8%	19.1%	8.7%	15.8%

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 2ND-3RD NOVEMBER

VEHICLE COMPOSITION	(%AGE)			
P			L	
3302/5199 : 64.4%			352/5199 : 6.8%	
B			M	
315/5199 : 6.1%			1140/5199 : 21.9%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
701	114	41	156	1012
%AGE 69.3%	11.3%	4.0%	15.4%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
701	114	41	156	1012
%AGE 21%	34.8%	13.1%	14.1%	21.9%

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 3RD-4TH NOVEMBER

VEHICLE COMPOSITION	(%AGE)			
P			L	
2543/4301 : 59.1%			686/4301 : 15.9%	
B			M	
240/4301 : 5.5%			832/4301 : 19.3%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
309	129	28	41	507
%AGE 60.9%	25.4%	5.5%	8.2%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
309	129	28	41	507
%AGE 12%	18.5%	11.6%	4.8%	11.6%

A.1/7

APPENDIX A.2

MANUAL TRAFFIC COUNT AT BLUE POST HOTEL 4TH-5TH NOVEMBER

VEHICLE COMPOSITION	(%AGE)			
P			L	
2604/4346 : 59.9%			766/4346 : 17.6%	
B 5/4346 : 5.4%			M	
235/4346 : 5.4%			741/4346 : 17.1%	
NIGHT TRAFFIC (7.00 pm - 6.30 am)				
P	L	B	M	VEHICLES
385	148	39	72	664
%AGE 59.8%	22.9%	6.1%	11.2%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
385	148	39	72	644
%AGE 14.8%	19.6%	16.5%	9.8%	14.8%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE COLLEGE 6TH-7TH FEBRUARY

VEHICLE COMPOSITION	(%AGE)			
P	L	B	M	VEHICLES
652	153	15	101	921
70.8%	16.6%	1.6%	11.0%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
131	28	4	29	192
68.2%	14.6%	2.1%	15.1%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
20.1%	18.3%	26.7%	28.7%	20.8%

A.2/1

MANUAL TRAFFIC COUNT AT KIGANJO POLICE COLLEGE 1ST-2ND FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
968	409	52	202	1631
59.4%	25.0%	3.2%	12.4%	100%
NIGHT TRAFFIC (6.00 pm − 6.00 am)				
P	L	B	M	VEHICLES
201	55	7	16	279
72.1%	19.7%	2.5%	5.7%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
20.8%	13.4%	13.5%	7.9%	17.1%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE COLLEGE 22ND-23RD JANUA.

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
893	479	58	223	1653
54.0%	29.0%	3.5%	13.5%	100%
NIGHT TRAFFIC (6.00 pm − 6.00 am)				
P	L	B	M	VEHICLES
189	40	9	26	264
71.6%	15.2%	3.4%	9.8%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
21.2%	8.4%	15.5%	11.7%	16.0%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE COLLEGE 2ND-3RD FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
733	330	14	70	1147
63.9%	28.8%	1.2%	6.1%	100%
NIGHT TRAFFIC (6.00 pm − 6.00 am)				
P	L	B	M	VEHICLES
172	36	1	13	222
77.5%	16.2%	0.4%	5.9%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
23.5%	10.9%	7.1%	18.6%	19.4%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE COLLEGE 3RD-4TH FEBRUA.

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
662	332	6	64	1064
62.2%	31.2%	0.6%	6.0%	100%
NIGHT TRAFFIC (6.00 pm − 6 am)				
P	L	B	M	VEHICLES
131	32	4	12	179
73.2%	17.9%	2.2%	6.7%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
19.8%	9.6%	66.7%	18.8%	16.8%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE KOLLEGE

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
968	486	52	202	1708
56.7%	28.5%	3.0%	11.8%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
228	44	13	26	311
73.3	14.1%	4.2%	8.4%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
23.6%	9.1%	25.0%	12.9%	18.2%

MANUAL TRAFFIC COUNT AT KIGANJO KENO 2ND-3RD FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
761	302	48	118	1229
61.9%	24.6%	3.9%	9.6%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
161	41	11	16	229
70.3%	17.9%	4.8%	7.0%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
21.2.%	13.6%	22.9%	13.6%	18.6%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE COLLEGE 4TH-5TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
639	296	19	78	1032
61.9%	28.7%	1.8%	7.6%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
201	39	3	11	254
79.1%	15.4%	1.2%	4.3%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
31.5%	13.2%	15.8%	14.1%	24.6%

MANUAL TRAFFIC COUNT AT KIGANJO POLICE KOLLEGE 4TH-5TH FEBR

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
1026	453	64	250	1793
57.2%	25.3%	3.6%	13.9%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
249	54	12	29	344
72.4%	15.7%	3.5%	8.4%	100
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
24.3%	11.9%	18.8%	11.6%	19.2%

MANUAL TRAFFIC COUNT AT KIGANJO KCC 3RD-4TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
811	306	43	126	1286
63.1%	23.8%	3.3%	9.8%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
175	37	11	9	232
75.4%	16.0%	4.7%	3.9%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
21.6%	12.1%	25.6%	7.1%	18.0%

MANUAL TRAFFIC COUNT AT KIGANJO KCC 3RD-4TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
901	434	49	130	1514
59.5%	28.7%	3.2%	8.6%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
192	47	13	12	264
72.7%	17.8%	4.9%	4.6%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
21.3%	10.8%	26.5%	9.2%	17.4%

MANUAL TRAFFIC COUNT AT CHAKA 5TH-6TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	N	VEHICLES
917	322	51	169	1459
62.8%	22.1%	3.5%	11.6%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
213	38	7	38	296
72.0%	12.8%	2.4%	12.8%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
23.3%	11.8%	13.7%	22.5%	20.3%

MANUAL TRAFFIC COUNT AT CHAKA 5TH-6TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
1026	375	49	187	1637
62.7%	22.9%	30%	11.4%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
250	40	7	38	335
74.6%	11.9%	2.2%	11.3%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
24.4%	10.7%	14.3%	20.3%	20.5%

MANUAL TRAFFIC COUNT AT CHAKA 7TH-8TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
751	336	55	147	1289
58.2%	26.1%	4.3%	11.4%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
140	40	11	21	212
66.0%	18.9%	5.2%	9.9%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
18.6%	11.9%	20.0%	14.3%	16.4%

MANUAL TRAFFIC COUNT AT CHAKA 7TH-8TH FEBRUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
805	358	56	153	1372
58.7%	26.1%	4.1%	11.1%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
163	42	12	23	240
67.9%	17.5%	5.0%	9.6%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
20.2%	11.7%	21.4%	15.0%	17.5%

MANUAL TRAFFIC COUNT AT NYANGE 24TH-25TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
565	261	40	121	987
57.2%	26.4%	4.1%	12.3%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
128	35	7	21	191
67.0%	18.3%	3.7%	11.0%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
22.7%	13.4%	17.5%	17.4%	19.4%

MANUAL TRAFFIC COUNT AT NYANGE 24TH-25TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
602	279	40	144	1065
56.5%	26.2%	3.8%	13.5%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
152	37	5	21	215
70.7%	17.2%	2.3%	9.8%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
25.2%	13.3%	12.5%	14.6%	20.2%

A.2/5

MANUAL TRAFFIC COUNT AT NARO MORU 21ST-22ND JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
743	321	41	134	1239
60.0%	25.9%	3.3%	10.8%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
155	51	6	14	226
68.5%	22.6%	2.7%	6.2%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
20.9%	15.9%	14.6%	10.4%	18.2%

MANUAL TRAFFIC COUNT AT NARO MORU 21ST-22ND JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
833	317	45	137	1332
62.5%	23.8%	3.4%	10.3%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
183	53	6	16	258
70.9%	20.5%	2.3%	6.2%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
22.0%	16.7%	13.3%	11.7%	19.4%

MANUAL TRAFFIC COUNT AT NARO MORU 23RD-24TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
555	128	41	141	865
64.2%	14.8%	4.7%	16.3%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
145	19	8	20	192
75.5%	9.9%	4.2%	10.4%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
26.1%	14.8%	19.5%	14.2%	22.2%

MANUAL TRAFFIC COUNT AT NARO MORU 23RD-24TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
647	124	41	171	983
65.8%	12.6%	4.2%	17.4%	100%
NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
165	20	8	23	216
76.4%	9.3%	3.7%	10.6%	100%
%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
25.5%	16.1%	19.5%	13.5%	22.0%

A.2/6

MANUAL TRAFFIC COUNT AT NARO MORU 25TH-26TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
779	283	57	130	1249
62.4%	22.6%	4.6%	10.4%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
220	52	9	13	294
74.8%	17.7%	3.1%	4.4%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
28.2%	18.4%	15.8%	10.0%	23.5%

MANUAL TRAFFIC COUNT AT NARO MORU 25TH-26TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
654	289	38	128	1109
59.0%	26.1%	3.4%	11.5%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
146	37	7	23	213
68.5%	17.4%	3.3%	10.8%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
22.3%	12.8%	18.4%	18.0%	19.2%

MANUAL TRAFFIC COUNT AT MOI EQT. GIRLS S. SCHOOL 27TH-28TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
738	262	45	88	1133
65.1%	23.1%	4.0%	7.8%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
198	47	8	7	260
76.1%	18.1%	3.1%	2.7%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
26.8%	17.9%	17.8%	8.0%	22.9%

MANUAL TRAFFIC COUNT AT SILVERBECK HOTEL 27TH-28TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
1087	350	40	96	1573
69.1%	22.3%	2.5%	6.1%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)				
P	L	B	M	VEHICLES
282	51	7	7	347
81.3%	14.7%	2.0%	2.0%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC				
P	L	B	M	VEHICLES
25.9%	14.6%	17.5%	7.3%	22.1%

A.2/7

MANUAL TRAFFIC COUNT AT MOI EQT. GIRLS S. SCHOOL 27TH-28TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
738	262	45	88	1133
65.1%	23.1%	4.0%	7.8%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)

P	L	B	M	VEHICLES
198	47	8	7	260
76.1%	18.1%	3.1%	2.7%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
26.8%	17.9%	17.8%	8.0%	22.9%

MANUAL TRAFFIC COUNT AT SILVERBECK HOTEL 27TH-28TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
1087	350	40	96	1573
69.1%	22.3%	2.5%	6.1%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)

P	L	B	M	VEHICLES
282	51	7	7	347
81.3%	14.7%	2.0%	2.0%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
25.9%	14.6%	17.5%	7.3%	22.1%

MANUAL TRAFFIC COUNT AT SILVERBECK HOTEL 27TH-28TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
971	311	43	109	1434
67.7%	21.7%	3.0%	7.6%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)

P	L	B	M	VEHICLES
263	65	8	5	351
74.9%	18.5%	2.3%	4.3%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
27.1%	20.9%	18.6%	13.7%	24.5%

MANUAL TRAFFIC COUNT AT NANYUKI TOWN HALL 26TH-27TH JANUARY

VEHICLE COMPOSITION		(%AGE)		
P	L	B	M	VEHICLES
1881	405	69	176	2531
74.3%	16.0%	2.7%	7.0%	100%

NIGHT TRAFFIC (6.00 pm - 6.00 am)

P	L	B	M	VEHICLES
406	60	16	17	499
81.4%	12.0%	3.2%	3.4%	100%

%AGE NIGHT TRAFFIC / 24 HOURS TRAFFIC

P	L	B	M	VEHICLES
21.6%	14.8%	23.2%	9.7%	1.9%

A.2/8

APPENDIX A.3

Public Transportation Usage
Registration Near Safari Park
7:7 - 21.7:80.

Public Transportation Usage
Registration East of Ruiru
30.7 - 13:8:80.

A.3

PEDESTRIANS & BICYCLES CROSSING THIKA ROAD NEAR GITH

TIME	6.30 - 7.30 - 8.30 - 9.30 - 10.30 - 11.30 - 12.30 -						
PEDESTRIANS	337	298	139	124	126	86	13
BICYCLES	6	9	6	4	7	3	8

APPENDIX   A.5

THE KENYA POLICE—TRAFFIC DEPARTMENT

POLICE 41

Charge Reg. No. _____

ACCIDENT REPORT

Acc. Reg. No. _____

O.B. No. _____

Reporting Officer _____ Division _____ Station _____

Location of Accident _____

Highway Authority _____

Date _____ Day _____ Time _____

	Types of Vehicles concerned	Registration Numbers	Names and Addresses of Owners and/or Drivers (state which)	Brief details of damage
1				
2				
3				

	If in order state "Yes" If endorsed or not in order give details overleaf		
	1.	2.	3.
Certificate of Competency			
Driving Licence			
Road Tax			
Insurance			
T.L.B. Licence			
P.S.V. or Taxi Licence			

Class of Persons: "P" Pedestrian, "D" Driver, "MC" Motor Cyclist, "PS" Passenger, "C" Cyclist, "PP" Pillion Passenger, "O.P." Other Person.

In case of Fatal Accident give Date of Death.

Name and Address of injured persons	Class of person	Age	Sex	Fatal Serious Slight	From Vehicle No.	Tribe or Nationality

Names and Addresses of witnesses.—

_____

_____

_____

_____

Nationality or Tribe of vehicle driver primarily responsible _____

Road Surface: Tarmac, Murram/Unmade,   Speed limit of _____ m.p.h.
Wet, Dry.   Street lighting — Yes/No

Apparent Cause Code No _____   Special Features if any _____

Are Court proceedings Contemplated—Yes/No

When a Police vehicle is involved, this form must be despatched to the Commissioner of Police and the Divisional Transport Officer within 24 hours.

P.T.O.

A.5/1

Form of Accident (Prepare sketch plan on separate sheet of paper).

Remarks of Investigating Officer.

Has a notice of intended prosecution been served against either driver and, if so, on whom: _____

_____

_____

_____

Date of serving: _____

Action taken

(Police accidents only)   Date

Despatched to Commissioner of Police   _____

"   " P.P.O.   ⎫
"   " Divisional Officer   ⎬ Through D.T.O. _____

Date _____

_____
Officer in charge Police Station

GPK (i)

A.5/2

A. NODE

①

8
7
6
5
4
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0
　0 1 2 3 4 5 6 7 8

③

B ROAD SECTION

⑤

8
7
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1
0
　1 2 3 4 5 6 7 8

⑥

RECONSTRUCTED SCENE	
ELEMENT	MAKE
FIRST	
SECOND	

1. GENERAL

1.1.　　Type of form
1.2.　　Police Station
1.2.3.　Accident report reg. number
1.2.4.　Classification of the road

2. TIME, PAVEMENT, LIGHT & WEATHER CONDITION

2.1.1.　Day of week
2.1.2.　Year
2.1.3.　Month
2.1.4.　Day of month
2.1.5.　Time of day
2.2.　　Pavement
2.3.　　Road surface condition
2.4.　　Weather condition
2.5.　　Light condition

3. LOCATION

3.1.　　Accident location
3.2.1.　Node I
3.2.2.　Node II/other locations
3.3.1.　Alignment
3.3.2.　Sight obstruction
3.3.3.　Pedestrian/bicycle crossing
3.3.4.　Bus-stop/railway crossing
3.3.5.　Access

4. PRIMARY ACCIDENT SPOT

5. THE FIRST ELEMENT INVOLVED IN THE ACCIDENT

5.1.　　Type of 1st element in the
　　　　accident involvement
5.2.　　Maneuvre
5.3.　　Direction of travel
5.4.1.　Changes in cross-section
5.4.2.　Changes caused by road inventory
5.4.3.　Changes in traffic environment

5.4.4.1.　Type of disturbing traffic element
5.4.4.2.　The traffic element's maneuver
5.4.4.3.　Direction of travel
5.5.1.　　Traffic violation
5.5.2.　　Medical/Psyco-condition
5.5.3.　　Socio. aspects of the traffic violation
5.5.4.　　Vehicle failure, load, passenger
5.6.1.　　Secondary accident element
5.6.2.　　Type of collision obstruction
5.6.3.　　Secondary spot of collision

6. THE SECOND ELEMENT INVOLVED IN THE ACCIDENT

6.1.　　Type of 2nd element in the accident
　　　　involvement
6.2.　　Maneuvre
6.3.　　Direction of travel
6.4.1.　Changes in cross-section
6.4.2.　Changes in road inventory
6.4.3.　Changes in traffic environment
6.4.4.1.　Type of disturbing traffic element
6.4.4.2.　The traffic element's maneuvre
6.4.4.3.　Direction of travel
6.5.1.　　Traffic Violation
6.5.2.　　Medical/Psycho.-condition
6.5.3.　　Socio. Aspects of traffic violation
6.5.4.　　Vehicle failure, load, passenger
6.6.1.　　Secondary accident element
6.6.2.　　Type of collision obstruction
6.6.3.　　Secondary spot of collision

7. INJURIES, PROPERTY DAMAGES

7.1.　　The degree of injury
7.2.　　Number of injured persons
7.3.　　Number of vehicles damaged.

8. TYPE OF ACCIDENT

UNIVERSITY OF NAIROBI	KENYA POLICE	ACCIDENT DATA	G.S.G. AGOMI	NO.
DEPARTMENT OF CIVIL ENGINEERING	TRAFFIC DEPARTMENT	FORM A.		
		AR NAME:	AT NUMBER:	
		DATE:		

1. GENERAL

    1.1.   Type of form
    1.2.1. Police Station
    1.2.2. Accident report reg. number

2. TIME

    Year   Year when accident occurred

3. LOCATION

    3.1.   Junction detail
    3.2.   Road category

4. PRIMARY ACCIDENT ELEMENT 1 (DRIVER1)
    4.1.1. Sex
    4.1.2. Marital status
    4.1.3. Nationality/race/tribe
    4.1.4. Age Group
    4.1.5. Occupation
    4.2.   Degree of injury
    4.3.1. Driving experience
    4.3.2. Frequency of accident involvement
    4.3.3. Previous court conviction
    4.3.4. Driving speed prior to accident
    4.3.5. Speed limit
    4.4.1. Vehicle registration number
    4.4.2. Age of vehicle,
    4.4.3. Pre accident defect
    4.4.4. Degree of damage
    4.4.5. Type of damage
    4.5.   Purpose of trip

5. PRIMARY ACCIDENT ELEMENT 2 (DRIVER2)
    5.1.1. Sex
    5.1.2. Marital Status
    5.1.3. Nationality/race/tribe
    5.1.4. Age group
    5.1.5. Occupation
    5.2.   Degree of injury
    5.3.1. Driving experience
    5.3.2. frequency of accident involvement

    5.3.3. Previous court conviction
    5.3.4. Driving speed prior to accident
    5.3.5. Speed limit
    5.4.1. Vehicle registration number
    5.4.2. Age of vehicle
    5.4.3. Pre-accident defect
    5.4.4. Degree of damage
    5.4.5. Type of damage
    5.5.   Purpose of trip

6. OTHER INJURED PERSONS

    6.1.   Accident element type
    6.2.   Driver/Passenger/pedestrians Sex
    6.3.   Driver/Passenger/Pedestrians age group
    6.4.   Marital status
    6.5.   Nationality/race/tribe
    6.6.   Occupation
    6.7.   Degree of injury
    6.8.   Purpose of trip

    7.1.   Accident element type
    7.2.   Drive/Passenger/Pedestrian, Sex
    7.3.   Driver/Passenger/Pedestrian, age group
    7.4.   Marital Status
    7.5.   Nationality/race/tribe
    7.6.   Occupation
    7.7.   Degree of injury
    7.8.   Purpose of trip

    8.1.   Accident element type
    8.2.   Driver/Passenger/Pedestrian Sex
    8.3.   Driver/Passenger/Pedestrian Age group
    8.4.   Marital status
    8.5.   Nationality/race/tribe
    8.6.   Occupation
    8.7.   Degree of inury
    8.8.   Purpose of trip

UNIVERSITY OF NAIROBI KENYA POLICE DEPARTMENT OF CIVIL TRAFFIC DEPT. ENGINEERING	ACCIDENT DATA FORM B	G.S.O. AGOKI		NO.
	ROAD NAME:		ROAD NUMBER:	
	DATE:			

8	Roadside	
6	Footpath	
6	Parking place, shoulder	
5	Carriageway/traffic lane	
3	Carriageway	

CODING FOR RTA LOCATION

2	Parking place, shoulder
1	Footpath
0	Roadslide
99	Unknown

A.6/3

NAIROBI - THIKA ROAD RTAS DISTRIBUTION BY LO

LOCATION	SEVERITY OF INJURY				
	FATAL	SERIOUS	SLIGHT	NONE	
Junctions	29	27	39	44	
Sections	101	106	145	211	
Total	130	133	184	255	
%	18.6	18.9	26.2	36.3	

NAIROBI - THIKA ROAD DISTRIBUTION OF RTAS IN

TYPE	SEVERITY OF INJURY		
	FATAL	SERIOUS	SL
1. Vehicle-vehicle in same direction	2	7	
2. Head-on collision	1	1	
3. Turning from same direction	0	0	
4. Turning from opposite direction	0	0	
5. Crossing without turning	1	0	
6. Crossing with turning	0	0	
7. Pedestrian crossing carriageway	25	7	
8. Pedestrian walking along	7	0	
9. Vehicle turns off the road	5	9	
0. Other types	10	11	
TOTAL	51	35	
%	25.5	17.5	

TYPE	DAY	NIGHT	TOTAL	%
1: Vehicle-vehicle in same direction	12	1	13	13.8
2: Head-on collision	12	8	20	21.3
3: Turning from same direction	2	0	2	2.1
4: Turnign from opposite direction	1	0	1	1.1
5: Crossing without turning	1	0	1	1.1
6: Crossing with turning	2	0	2	2.1
7: Pedestrian crossing carriageway	9	3	12	12.7
8: Pedestrian walking along the road	3	0	3	3.2
9: Vehicle turns off road	14	6	20	21.3
0: Other types	14	6	20	21.3
TOTAL	70	24	94	100
	74.5	25.5	100	

KIGANJO—NANYUKI ROAD: SEX DISTRIBUTION OF DRIVERS, PEDESTRIANS AND CYCLISTS INVOLVED IN RTAs

SEX	ELEMENTS		TOTAL	%
	FIRST	SECOND		
Irrelevant (animals)	0	3	3	1.8
Unknown	6	5	11	6.7
Male	94	48	142	87.1
Female	1	6	7	4.3
Total	101	62	163	100
%	62.0	38.0	100	

DISTRIBUTION OF RTAS BY TYPE : NAIROBI — THIKA ROAD

TYPE	SEVERITY OF INJURY				TOTAL	%
	FATAL	SERIOUS	SLIGHT	NONE		
1. Vehicle-vehicle in same direction	14	23	53	176	266	37.9
2. Head-on collision	3	3	1	2	8	1.1
3. Turning from same direction	1	3	4	10	18	2.6
4. Turning from opposite direction	0	0	0	2	2	0.3
5. Crossing without turning	2	3	7	9	21	3.0
6. Crossing with turning	1	3	5	3	12	1.7
7. Pedestrian crossing carriageway	57	33	19	1	110	15.7
8. Pedestrian walking along	11	7	6	0	24	3.4
9. Vehicle turns off the road	16	33	52	31	132	18.8
0. Other types	25	26	37	21	109	15.5
TOTAL	130	133	184	255	702	100

KIGANJO—NAYUKI ROAD: DISTRIBUTION OF RTAs BY TYPE   APPENDIX A.12

TYPE	SEVERITY OF INJURY			TOTAL	%
	FATAL	SERIOUS	SLIGHT		
1. Vehicle-vehicle in same direction	4	3	6	13	13.7
2. Head-on collison	5	10	5	20	21.3
3. Turning from same direction	1	1	0	2	2.1
4. Turning from opposite direction	0	1	0	1	1.1
5. Crossing without turning	1	0	0	1	1.1
6. Crossing with turning	0	2	0	2	2.1
7. Pedestrian crossing carriageway	5	7	0	12	12.8
8. Pedestrian waling along the road	0	1	2	3	3.2
9. Vehicle turns off the road	4	8	8	20	21.3
0. Other types	5	12	3	20	21.3
TOTAL	25	45	24	94	100

YEAR	TIME				
1948	0	5.251120	.	5.199167	5.240178
1949	1	.	5.489	5.400955	5.243358
1950	2	.	5.579	5.609661	5.633720
1951	3	.	5.669	5.825452	5.840789
1952	4	.	5.760	6.048491	6.055000
1953	5	.	5.851	6.278941	6.276563
1954	6	.	5.948	6.516960	6.505693
1955	7	.	6.048	6.762704	6.742609
1956	8	.	6.993	7.016326	6.987533
1957	9	.	7.209	7.277972	7.240690
1958	10	.	7.432	7.547785	7.502307
1959	11	.	7.880	7.825900	7.772616
1960	12	.	8.115	8.112446	8.051849
1961	13	.	8.352	8.407545	8.340241
1962	14	8.636263	.	8.711310	8.638030
1963	15	.	8.847	9.023844	8.945455
1964	16	.	9.104	9.345242	9.262756
1965	17	.	9.385	9.675586	9.590173
1966	18	.	9.643	10.014948	9.927946
1967	19	.	9.948	10.363386	10.276319
1968	20	.	10.209	10.720704	10.635530
1969	21	10.742706	.	11.087655	11.005817
1970	22	.	11.247	11.463533	11.387421
1971	23	.	11.694	11.849578	11.780594
1972	24	.	12.091	12.242773	12.185511
1973	25	.	12.504	12.646083	12.602459
1974	26	.	12.935	13.058455	13.031642
1975	27	.	13.413	13.479816	13.473281
1976	28	.	13.853	13.910076	13.927589
1977	29	.	14.348	14.349121	14.394773
1978	30	.	14.875	14.796821	14.875033
1979	31	15.487000	.	15.252021	15.368559
1980	32	.	15.953	15.717547	15.875524	16.440 16.491
1981	33	.	15.500	15.190202	15.396150
1982	34	.	17.100	16.670769	16.936509
1983	35	.	18.784	17.159005	17.478616
1984	36	.	.	17.654652	18.041194
1985	37	.	.	18.157426	18.617759	20.154 20.207
1986	38	.	.	18.567083	19.208920
1987	39	.	.	19.183119	19.813833
1988	40	.	.	19.705366	20.433577
1989	41	.	.	20.232404	21.096782
1990	42	.	.	20.766648	21.716581	23.811 25.032
1991	43	.	.	21.305299	22.379931
1992	44	.	.	21.848239	23.057847
1993	45	.	.	22.395535	23.750303
1994	46	.	.	22.946442	24.457250
1995	47	.	.	23.500600	25.178618
1996	48	.	.	24.057539	25.914305
1997	49	.	.	24.616779	26.664193
1998	50	.	.	25.177831	27.428134
1999	51	.	.	25.740201	28.205951
2000	52	.	.	26.303389	28.997445	31.125 38.550
2001	53	.	.	26.866899	29.802365
2002	54	.	.	27.430209	30.620515
2003	55	.	.	27.992836	31.451551
2004	56	.	.	28.554271	32.295179
2005	57	.	.	29.114020	30.151058
2006	58	.	.	29.671599	34.008813
2007	59	.	.	30.228604	34.870043
2008	60	.	.	30.778285	30.763368
2009	61	.	.	31.328471	36.439281

YEAR	TIME	VEHICLES
1949	0	.030014
1950	1	.035408
1951	2	.039900
1952	3	.044170
1953	4	.043939
1954	5	.049778
1955	6	.056861
1956	7	.063686
1957	8	.067670
1958	9	.073358
1959	10	.077669
1960	11	.089505
1961	12	.084540
1962	13	.087130
1963	14	.087073
1964	15	.092581
1965	16	.099258
1966	17	.103175
1967	18	.109439
1968	19	.113329
1969	20	.124348
1970	21	.137271
1971	22	.149750
1972	23	.159969
1973	24	.164222
1974	25	.184086
1975	26	.199715
1976	27	.203446
1977	28	.213241
1978	29	.225447
1979	30	.232069
1980	31	.240435
1981	32	.246132
1982	33	.247162
1983	34	.250919
1984	35	.
1985	36	.
1986	37	.
1987	38	.
1988	39	.
1989	40	.
1990	41	.
1991	42	.
1992	43	.
1993	44	.
1994	45	.
1995	46	.
1996	47	.
1997	48	.
1998	49	.
1999	50	.
2000	51	.
2001	52	.
2002	53	.
2003	54	.
2004	55	.
2005	56	.
2006	57	.
2007	58	.

AGOKI1	AGOKI2	YEAR	TIME	MOTORIZA	MOV.AVER	AGOK.TFRE
.036600	.036680	1949	0	.00547	.	.00785
.039026	.039097	1950	1	.00635	.	.00744
.041612	.041671	1951	2	.00688	.00678	.00764
.044367	.044415	1952	3	.00767	.00736	.00784
.047304	.047339	1953	4	.00751	.00797	.00804
.050433	.050454	1954	5	.00837	.00841	.00823
.053766	.053773	1955	6	.00940	.00875	.00847
.057318	.057309	1956	7	.00910	.00923	.00869
.061102	.061076	1957	8	.00939	.00952	.00891
.065132	.065089	1958	9	.00987	.00985	.00914
.069425	.069364	1959	10	.00986	.01005	.00938
.073997	.073918	1960	11	.01103	.01019	.00962
.078865	.078769	1961	12	.01012	.01019	.00986
.084048	.083935	1962	13	.01009	.01025	.01011
.089566	.089438	1963	14	.00984	.01014	.01037
.095440	.095297	1964	15	.01017	.01025	.01063
.101692	.101537	1965	16	.01047	.01044	.01090
.108345	.108181	1966	17	.01070	.01069	.01112
.115423	.115255	1967	18	.01100	.01093	.01146
.122953	.122786	1968	19	.01110	.01127	.01175
.130962	.130803	1969	20	.01136	.01170	.01204
.139478	.139336	1970	21	.01221	.01214	.01234
.148532	.148418	1971	22	.01291	.01255	.01265
.159157	.158082	1972	23	.01323	.01312	.01296
.168384	.168366	1973	24	.01313	.01366	.01328
.179251	.179307	1974	25	.01423	.01403	.01360
.190792	.190945	1975	26	.01489	.01438	.01394
.203047	.203324	1976	27	.01469	.01478	.01427
.216057	.216488	1977	28	.01494	.01474	.01462
.229962	.230485	1978	29	.01516	.01496	.01497
.244508	.245365	1979	30	.01504	.01501	.01532
.260039	.261132	1980	31	.01493	.	.01570
.275503	.277989	1981	32	.01492	.	.01607
.295847	.295847	1982	33	.01450	.	.01646
.312428	.314816	1983	34	.01336	.	.01684
.331992	.334961	1984	35	.	.	.01724
.352695	.356349	1985	36	.	.	.01764
.374593	.379052	1986	37	.	.	.01805
.397740	.403142	1987	38	.	.	.01847
.422176	.428698	1988	39	.	.	.01890
.448020	.455900	1989	40	.	.	.01933
.475271	.484532	1990	41	.	.	.01977
.504008	.514981	1991	42	.	.	.02022
.534271	.547239	1992	43	.	.	.02067
.566180	.581395	1993	44	.	.	.02114
.599735	.617552	1994	45	.	.	.02161
.635013	.655805	1995	46	.	.	.02209
.672072	.696260	1996	47	.	.	.02257
.710966	.739020	1997	48	.	.	.02306
.751748	.784193	1998	49	.	.	.02357
.794468	.831989	1999	50	.	.	.02407
.839169	.882220	2000	51	.	.	.02459
.885895	.935296	2001	52	.	.	.02511
.934631	.991233	2002	53	.	.	.02564
.985556	1.050143	2003	54	.	.	.02618
.033551	1.112139	2004	55	.	.	.02673
.093675	1.177390	2005	56	.	.	.02730
.150942	1.245953	2006	57	.	.	.02780
.210050	1.317969	2007	58	.	.	.02841
.271890	1.393761	2008	59	.	.	.02900

APPENDIX A.

YEAR	TIME	RTAVEHIC	MOVINGAV	AGOKI
1960	0	2788	.	2106
1961	1	2377	.	2217
1962	2	2469	2494	2335
1963	3	2385	2445	2458
1964	4	2450	2458	2588
1965	5	2542	2554	2725
1966	6	2445	2703	2868
1967	7	2947	2786	3020
1968	8	3129	2987	3179
1969	9	2869	3342	3346
1970	10	3546	3686	3522
1971	11	4217	3999	3706
1972	12	4668	4301	3901
1973	13	4695	4582	4105
1974	14	4379	4660	4319
1975	15	4751	4633	4545
1976	16	4605	4819	4782
1977	17	4536	5187	5031
1978	18	5625	5073	5292
1979	19	6216	5302	5567
1980	20	4485	5591	5855
1981	21	5648	5738	6158
1982	22	5981	.	6476
1983	23	6262	.	6809
1984	24	.	.	7159
1985	25	.	.	7526
1986	26	.	.	7910
1987	27	.	.	8314
1988	28	.	.	8736
1989	29	.	.	9190
1990	30	.	.	9644
1991	31	.	.	10130
1992	32	.	.	10639
1993	33	.	.	11171
1994	34	.	.	11727
1995	35	.	.	12313
1996	36	.	.	12921
1997	37	.	.	13559
1998	38	.	.	14224
1999	39	.	.	14919
2000	40	.	.	15644
2001	41	.	.	16401
2002	42	.	.	17190
2003	43	.	.	18013
2004	44	.	.	18870
2005	45	.	.	19762
2006	46	.	.	20671
2007	47	.	.	21657
2008	48	.	.	22661
2009	49	.	.	23704
2010	50	.	.	24736

YEAR	TIME	RTAS	MOVAVERA	AGOKI
1960	0	4101	.	2698
1961	1	3573	.	2948
1962	2	3595	3708	3207
1963	3	3578	3600	3474
1964	4	3693	3655	3746
1965	5	3562	3813	4020
1966	6	3847	4000	4295
1967	7	4387	4101	4566
1968	8	4511	4421	4833
1969	9	4196	4860	5093
1970	10	5163	5305	5344
1971	11	6042	5761	5583
1972	12	6613	6171	5810
1973	13	6799	6446	6024
1974	14	6250	6547	6224
1975	15	6534	6414	6409
1976	16	6548	6447	6581
1977	17	5949	6807	6737
1978	18	6956	6732	6880
1979	19	8049	6873	7010
1980	20	6162	.	7129
1981	21	7250	.	7233
1982	22	7524	.	7328
1983	23	8023	.	7413
1984	24	.	.	7488
1985	25	.	.	7556
1986	26	.	.	7615
1987	27	.	.	7668
1988	28	.	.	7715
1989	29	.	.	7756
1990	30	.	.	7792
1991	31	.	.	7824
1992	32	.	.	7852
1993	33	.	.	7876
1994	34	.	.	7895
1995	35	.	.	7917
1996	36	.	.	7934
1997	37	.	.	7948
1998	38	.	.	7961
1999	39	.	.	7978
2000	40	.	.	7982
2001	41	.	.	7990
2002	42	.	.	7998
2003	43	.	.	8004
2004	44	.	.	8010
2005	45	.	.	8015
2006	46	.	.	8019
2007	47	.	.	8023
2008	48	.	.	8026
2009	49	.	.	8029
2010	50	.	.	8032

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YEAR	MOTORIZA	OBSERVED	MODEL	AGOKI
1960	110.29	.04582	.	.03962
1961	101.22	.04226	.	.03930
1962	100.88	.04126	.04206	.03928
1963	98.42	.04109	.04015	.03904
1964	101.69	.03916	.03916	.03934
1965	104.69	.03625	.03892	.03952
1966	106.99	.03729	.03866	.03960
1967	110.01	.04009	.03743	.03963
1968	111.00	.03980	.03771	.03962
1969	113.63	.03374	.03832	.03954
1970	122.05	.03761	.03857	.03885
1971	128.05	.04035	.03888	.03795
1972	132.30	.04134	.03892	.03709
1973	131.33	.04134	.03794	.03730
1974	142.31	.03395	.03624	.03439
1975	148.59	.03272	.03353	.03208
1976	146.30	.03187	.03143	.03231
1977	149.39	.02775	.03158	.03189
1978	151.54	.03086	.03016	.03104
1979	150.40	.03469	.02968	.03149
1980	149.77	.02563	.	.03174
1981	149.17	.02946	.	.03198
1982	145.70	.03020	.	.03325
1983	133.58	.03797	.	.03680
1984	172.40	.	.	.02041
1985	176.40	.	.	.01739
1986	180.50	.	.	.01515
1987	184.70	.	.	.01218
1988	189.00	.	.	.00895
1989	193.30	.	.	.00555
1990	197.70	.	.	.00188
1991	202.20	.	.	.
1992	206.70	.	.	.
1993	211.40	.	.	.
1994	215.10	.	.	.
1995	220.70	.	.	.
1996	225.50	.	.	.
1997	230.60	.	.	.
1998	235.70	.	.	.
1999	240.70	.	.	.
2000	245.90	.	.	.
2001	251.10	.	.	.
2002	256.40	.	.	.
2003	261.80	.	.	.
2004	267.20	.	.	.
2005	272.80	.	.	.
2006	278.30	.	.	.
2007	284.00	.	.	.
2008	289.70	.	.	.
2009	295.50	.	.	.
2010	301.40	.	.	.

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YEAR	TIME	OBSERVED	MODEL	AGOKI1	AGOKI2
1960	0	1.833	.	1.689	1.061
1961	1	1.691	.	1.677	1.593
1962	2	1.650	1.702	1.663	1.592
1963	3	1.644	1.654	1.648	1.585
1964	4	1.596	1.579	1.632	1.594
1965	5	1.450	1.552	1.615	1.600
1966	6	1.491	1.551	1.596	1.602
1967	7	1.603	1.499	1.576	1.601
1968	8	1.592	1.560	1.555	1.600
1969	9	1.350	1.563	1.532	1.595
1970	10	1.504	1.546	1.508	1.562
1971	11	1.614	1.539	1.482	1.521
1972	12	1.654	1.543	1.455	1.484
1973	13	1.654	1.570	1.427	1.493
1974	14	1.358	1.423	1.397	1.368
1975	15	1.309	1.285	1.365	1.271
1976	16	1.287	1.268	1.332	1.280
1977	17	1.110	1.270	1.297	1.263
1978	18	1.234	1.194	1.261	1.227
1979	19	1.388	1.135	1.224	1.246
1980	20	1.025	.	1.186	1.256
1981	21	1.178	.	1.146	1.266
1982	22	1.210	.	1.106	1.320
1983	23	1.279	.	1.065	1.471
1984	24	.	.	1.023	.783
1985	25	.	.	.791	.679
1986	26	.	.	.935	.505
1987	27	.	.	.900	.442
1988	28	.	.	.853	.309
1989	29	.	.	.811	.168
1990	30	.	.	.769	.017
1991	31	.	.	.728	.
1992	32	.	.	.687	.
1993	33	.	.	.648	.
1994	34	.	.	.609	.
1995	35	.	.	.572	.
1996	36	.	.	.536	.
1997	37	.	.	.502	.
1998	38	.	.	.469	.
1999	39	.	.	.437	.
2000	40	.	.	.407	.
2001	41	.	.	.378	.
2002	42	.	.	.351	.
2003	43	.	.	.325	.
2004	44	.	.	.301	.
2005	45	.	.	.279	.
2006	46	.	.	.257	.
2007	47	.	.	.237	.
2008	48	.	.	.219	.
2009	49	.	.	.201	.
2010	50	.	.	.193	.

YEAR	TIME				
1949	0	713.67	.	363.38	217.73
1950	1	341.16	.	371.93	282.04
1951	2	357.87	405.57	380.47	317.59
1952	3	325.56	335.36	388.98	365.88
1953	4	290.18	360.37	397.46	356.62
1954	5	362.61	383.22	405.88	403.96
1955	6	466.22	399.85	414.26	452.17
1956	7	471.51	427.29	422.57	439.51
1957	8	408.75	436.99	430.80	451.59
1958	9	427.36	345.68	438.96	470.92
1959	10	411.10	441.02	447.02	470.38
1960	11	481.20	457.06	454.99	508.82
1961	12	476.70	481.47	462.85	480.18
1962	13	488.92	505.14	470.60	478.96
1963	14	549.42	504.53	478.23	469.85
1964	15	529.48	506.69	485.74	481.84
1965	16	478.13	517.55	493.13	492.00
1966	17	487.52	509.11	500.37	499.25
1967	18	543.22	502.00	507.48	509.04
1968	19	507.20	519.38	514.45	510.75
1969	20	493.64	536.13	521.27	517.52
1970	21	505.01	559.11	527.94	535.03
1971	22	571.29	584.46	543.04	
1972	23	655.13	607.33	540.62	547.81
1973	24	655.64	590.67	547.03	547.00
1974	25	572.56	593.49	553.08	551.24
1975	26	461.74	560.55	559.98	546.63
1976	27	525.36	514.21	564.71	549.02
1977	28	493.44	526.62	570.29	548.27
1978	29	437.97	517.02	573.71	546.48
1979	30	535.55	504.36	590.98	547.48
1980	31	438.67	.	586.03	547.95
1981	32	523.55	.	591.04	548.43
1982	33	552.45	.	595.23	550.38
1983	34	549.44	.	600.48	548.74
1984	35	.	.	604.98	507.74
1985	36	.	.	608.33	495.83
1986	37	.	.	613.53	462.24
1987	38	.	.	617.59	466.71
1988	39	.	.	621.31	449.13
1989	40	.	.	625.39	430.00
1990	41	.	.	625.95	409.67
1991	42	.	.	632.47	385.07
1992	43	.	.	635.65	359.57
1993	44	.	.	639.12	331.20
1994	45	.	.	642.27	300.77
1995	46	.	.	645.29	257.66
1996	47	.	.	648.20	232.49
1997	48	.	.	651.00	174.47
1998	49	.	.	653.69	152.62
1999	50	.	.	656.28	109.25
2000	51	.	.	658.76	61.97
2001	52	.	.	661.14	12.19
2002	53	.	.	663.43	.
2003	54	.	.	665.63	.
2004	55	.	.	667.73	.
2005	56	.	.	669.75	.
2006	57	.	.	671.49	.
2007	58	.	.	673.75	.
2008	59	.	.	675.02	.
2009	60	.	.	677.13	.
2010	61	.	.	679.27	.
.	.	.	.	.	.

YEAR	TIME	DEATHS
1949	0	148
1950	1	159
1951	2	162
1952	3	157
1953	4	150
1954	5	186
1955	6	288
1956	7	302
1957	8	268
1958	9	282
1959	10	303
1960	11	332
1961	12	329
1962	13	394
1963	14	548
1964	15	521
1965	16	552
1966	17	559
1967	18	596
1968	19	670
1969	20	750
1970	21	944
1971	22	1046
1972	23	1331
1973	24	1402
1974	25	1351
1975	26	1338
1976	27	1640
1977	28	1560
1978	29	1588
1979	30	1662
1980	31	1413
1981	32	1720
1982	33	1462
1983	34	1515
1984	35	.
1985	36	.
1986	37	.
1987	38	.
1988	39	.
1989	40	.
1990	41	.
1991	42	.
1992	43	.
1993	44	.
1994	45	.
1995	46	.
1996	47	.
1997	48	.
1998	49	.
1999	50	.
2000	51	.
2001	52	.
2002	53	.
2003	54	.
2004	55	.
2005	56	.
2006	57	.
2007	58	.
2008	59	.
2009	60	.
2010	61	.
.	.	.

MOVINGAV	RMK1	SMOOTHED	Real ...			MOVINGAV	RMK1	SMOOTHED
.	72	290	1949	54.61	148	.	-	290
.	85	310	1950	63.46	159	.	34	310
155	100	326	1951	68.77	148	155	71	326
163	118	341	1952	76.68	157	163	138	341
189	139	344	1953	75.09	150	189	123	344
217	163	362	1954	83.68	186	217	212	362
239	190	383	1955	94.01	288	239	346	383
265	222	438	1956	90.98	302	265	303	438
289	258	456	1957	93.86	268	289	344	456
297	298	478	1958	98.70	282	297	417	478
303	343	507	1959	98.56	303	303	415	507
328	393	542	1960	110.29	332	328	619	542
381	449	542	1961	101.22	329	381	458	542
425	509	560	1962	100.88	394	425	452	560
469	574	569	1963	98.42	548	469	413	569
515	642	592	1964	101.69	521	515	466	592
555	714	616	1965	104.69	552	555	517	616
580	798	637	1966	106.99	559	580	559	637
625	862	664	1967	110.01	596	625	613	664
704	937	683	1968	111.00	670	704	632	683
801	1011	738	1969	113.63	750	801	684	739
948	1081	777	1970	122.05	944	948	862	777
1095	1150	821	1971	128.05	1046	1095	1001	821
1215	1215	859	1972	132.30	1331	1215	1105	858
1294	1274	885	1973	131.33	1402	1294	1081	895
1412	1329	940	1974	142.31	1351	1412	1372	940
1458	1380	990	1975	148.89	1338	1458	1562	990
1495	1424	1018	1976	148.30	1640	1495	1545	1018
1558	1465	1059	1977	149.39	1560	1558	1577	1059
1573	1500	1104	1978	151.54	1598	1573	1642	1104
1589	1532	1142	1979	150.40	1562	1589	1607	1142
.	1559	1187	1980	149.77	1413	.	1539	1187
.	1583	1219	1981	149.17	1720	.	1570	1219
.	1603	1250	1982	145.70	1462	.	1468	1250
.	1620	1337	1983	133.58	1515	.	1136	1337
.	1625	1429	1984	172.44	.	.	2342	1429
.	1646	1489	1985	175.47	.	.	2490	1489
.	1657	1557	1986	180.50	.	.	2546	1557
.	1669	1615	1987	184.70	.	.	2312	1615
.	1677	1662	1988	189.00	.	.	2996	1682
.	1683	1751	1989	193.30	.	.	3165	1751
.	1689	1822	1990	197.70	.	.	3354	1822
.	1694	1896	1991	202.20	.	.	3553	1896
.	1698	1972	1992	206.70	.	.	3758	1972
.	1702	2051	1993	211.40	.	.	3978	2051
.	1704	2132	1994	216.10	.	.	4204	2132
.	1707	2215	1995	220.90	.	.	4441	2215
.	1709	2301	1996	225.70	.	.	4685	2301
.	1711	2390	1997	230.60	.	.	4940	2390
.	1712	2481	1998	235.70	.	.	5213	2481
.	1713	2575	1999	240.70	.	.	5488	2575
.	1715	2671	2000	245.90	.	.	5781	2671
.	1715	2770	2001	251.10	.	.	6082	2770
.	1716	2871	2002	256.40	.	.	6397	2871
.	1717	2975	2003	261.30	.	.	6725	2975
.	1717	3081	2004	267.20	.	.	7062	3081
.	1718	3190	2005	272.50	.	.	7420	3190
.	1718	3301	2006	278.30	.	.	7760	3301
.	1718	3414	2007	284.0.	.	.	9162	3414
.	1719	3529	2008	289.70	.	.	8553	3529
.	1719	3648	2009	295.5.	.	.	8951	3648
.	1719	3768	2010	301.40	.	.	9365	3768

YEAR	MOTORIZA	DPERFERS	MOVINGAV	AGDK1	JACOBS	SPEED
1949	54.68	.270		.056	.338	.589
1950	63.46	.285		.140	.370	.555
1951	68.77	.286		.192	.388	.570
1952	76.68	.273		.274	.414	.572
1953	75.09	.256		.283	.409	.567
1954	83.68	.313		.254	.437	.609
1955	94.01	.350		.342	.459	.633
1956	90.98	.370		.350	.468	.633
1957	93.86	.394		.394	.417	.626
1958	98.70	.379		.448	.488	.644
1959	98.56	.365		.499	.488	.643
1960	110.29	.409		.498	.515	.649
1961	101.22	.394		.405	.489	.668
1962	100.88	.453		.490	.488	.648
1963	98.42	.456		.523	.481	.643
1964	101.69	.619		.526	.491	.650
1965	104.69	.572		.563	.499	.656
1966	106.99	.580		.599	.506	.661
1967	110.01	.589		.622	.514	.667
1968	111.00	.599		.633	.517	.674
1969	113.63	.685		.735	.525	.633
1970	122.05	.835		.758	.548	.691
1971	128.05	.694		.827	.564	.702
1972	132.30	.928		.876	.575	.710
1973	131.33	1.032		.995	.572	.708
1974	145.17	1.121		.965	.499	.727
1975	148.31	1.044		.995	.656	.727
1976	148.89	1.087		1.075	.617	.738
1977	148.30	1.097		1.068	.615	.737
1978	145.30	1.184		1.076	.737	.738
1979	151.63	1.033		1.081	.739	.739
1980	151.54	1.087		1.107	.623	.741
1981	150.40	1.063		.883	.623	.742
1982	149.77	1.077		.928	.739	.739
1983	135.58	1.031		1.036	.736	.733
1984					.609	.733
1985					.617	.712
1986					.578	.775
2000					.893	
2001					.855	
2002					.867	
2003					.897	
2004					.891	
2005					.805	
2006					.893	
2007					.915	
2008					.924	
2009					.521	
2010					.524	

YEAR	MOTORIZA	DPERVEH1	MOVINGAV	AGDK1	JACOBS	SPEED
1949	54.68	49.30		25.48	76.65	
1950	63.46	44.50		30.32	87.52	
1951	68.77	41.50		33.22	82.96	
1952	76.68	37.53		37.15	82.17	
1953	75.09	34.10		36.67	78.23	
1954	83.68	37.40		41.32	72.78	
1955	94.01	42.30		41.84	68.84	
1956	90.98	47.50		45.25	68.84	
1957	93.86	46.79		49.83	67.42	
1958	98.70	48.84		50.46	65.20	
1959	98.56	49.37		49.30	65.26	
1960	110.29	48.87		46.72	64.26	
1961	101.22	48.41		64.11	64.26	
1962	100.88	48.26		65.32	63.91	
1963	98.42	50.96		48.89	63.91	
1964	101.69	52.55		42.90	62.67	
1965	104.69	53.77		46.76	61.79	
1966	106.99	54.50		49.60	60.65	
1967	110.01	55.37		52.17	60.29	
1968	111.00	55.89		54.59	59.36	
1969	113.63	57.28		56.59	58.59	
1970	122.05	61.02		54.81	54.81	
1971	128.05	64.82		53.63	53.63	
1972	132.30	83.20		53.89	53.89	
1973	131.33	66.52		51.09	51.09	
1974	145.17	72.18		49.56	49.56	
1975	148.31	75.55		41.43	41.70	
1976	148.89	75.81		41.38	41.70	
1977	148.39	78.90		48.99	49.46	
1978	151.54	76.02		48.99	54.81	
1979	150.40	76.32		41.27	51.63	
1980	149.77	79.06		41.34	49.37	
1981	149.17	75.69		47.51	47.51	
1982	145.70	73.92		41.79	50.29	
1983	133.58	67.68		43.27	53.29	
1984	172.42	80.42		39.02	44.95	
1985	176.40	80.72		38.72	44.27	
1986	180.50	91.46		43.00	43.00	
1987	184.70	93.54		42.93	42.93	
1988	189.00	95.67		37.66	42.23	
1989	193.30	77.78		37.32	41.65	
1990	197.70	69.83		36.89	41.03	
1991	202.20	102.50		40.42	40.42	
1992	202.20	105.30		39.83	39.83	
1993	206.70	106.60		39.24	39.24	
1994	211.40	108.9		38.67	38.67	
1995	216.10	111.2		38.11	38.11	
1996	220.90	113.4		37.55	37.55	
1997	225.70	115.8		37.03	37.03	
1998	230.60	118.2		36.49	36.49	
1999	235.70	120.6		35.99	35.99	
2000	240.70	122.9		35.48	35.48	
2001	245.90	125.4		34.99	34.99	
2002	251.10	127.9		34.50	34.50	
2003	256.40	130.4		34.03	34.03	
2004	261.80	132.7		33.57	33.57	
2005	267.20	135.4		33.10	33.10	
2006	272.30			32.52		
2007	278.20			32.00		
2008	283.20			31.75		
2009	289.50			31.30		
2010	301.40			30.78		

YEAR	MOTORIZA	DPERVEHK	MOVINGAV	AGDK1
1949	54.68	1772		1008
1950	63.46	1776		1204
1951	68.77	1662		1283
1952	76.68	1428		1359
1953	75.09	1366		1436
1954	83.68	1495		1514
1955	94.01	2026		1593
1956	90.98	1899		1672
1957	93.86	1584		1752
1958	98.70	1538		1831
1959	98.56	1560		1909
1960	110.29	1484		1987
1961	101.22	1557		2064
1962	100.88	1809		2139
1963	98.42	1924		2212
1964	101.69	2078		2284
1965	104.69	2154		2353
1966	106.99	2220		2420
1967	110.01	2242		2484
1968	111.00	2375		2544
1969	113.63	2413		2604
1970	122.05	2730		2730
1971	128.05	2751		2814
1972	132.30	3328		3045
1973	131.33	3031		3031
1974	145.17	3117		3117
1975	148.31	3033		3033
1976	148.89	2936		3035
1977	148.39	3415		3045
1978	151.54	3351		3045
1979	150.40	2617		2633
1980	149.77	2415		2740

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YEAR	MOTORIZA	OBSDRTAS	MOVAVERA	AGOKI
1960	110.29	4101	.	4478
1961	101.22	3573	.	3603
1962	100.88	3595	3708	3568
1963	98.42	3578	3600	3305
1964	101.90	3693	3655	3652
1965	104.69	3562	3813	3954
1966	106.99	3847	4000	4175
1967	110.01	4387	4101	4453
1968	111.00	4511	4421	4541
1969	113.63	4196	4860	4766
1970	122.05	5163	5305	5411
1971	126.05	6042	5761	5799
1972	132.30	6613	6171	6038
1973	131.33	6737	6446	5996
1974	142.31	6250	6484	6484
1975	148.89	6534	6414	6687
1976	148.30	6543	6467	6671
1977	149.39	5949	6807	6700
1978	151.54	6756	6752	6749
1979	150.40	6049	6873	6734
1980	149.77	6153	.	6709
1981	149.17	7250	.	6694
1982	145.70	7554	.	6567
1983	133.58	6063	.	6104
1984	178.40	.	.	5831
1985	176.40	.	.	5745
1986	180.50	.	.	5670
1987	184.70	.	.	6545
1988	189.00	.	.	6336
1989	193.30	.	.	6196
1990	197.70	.	.	5971
1991	202.20	.	.	5706
1992	206.70	.	.	5411
1993	211.40	.	.	5065
1994	216.10	.	.	4684
1995	220.90	.	.	4255
1996	225.70	.	.	3791
1997	230.60	.	.	3277
1998	225.70	.	.	2700
1999	240.70	.	.	2093
2000	245.90	.	.	1417
2001	251.10	.	.	698
2002	256.40	.	.	.
2003	261.80	.	.	.
2004	267.20	.	.	.
2005	272.60	.	.	.
2006	278.30	.	.	.
2007	284.00	.	.	.
2008	289.70	.	.	.
2009	295.50	.	.	.
2010	301.40	.	.	.

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YEAR	TIME	CASUALTI	MOVAVERA	AGOKI1	AGOKI2
1960	0	4307	.	3123	6172
1961	1	4030	.	3474	4532
1962	2	4260	4457	3851	4467
1963	3	4784	4535	4253	3989
1964	4	4902	4754	4680	4621
1965	5	4698	5091	5128	5182
1966	6	5125	5254	5595	5596
1967	7	5945	5530	6078	6124
1968	8	5599	6141	6573	6293
1969	9	6282	6827	7075	6730
1970	10	7756	7744	7580	8027
1971	11	8555	8824	8083	8856
1972	12	10528	9675	8579	9395
1973	13	10997	10048	9065	9276
1974	14	10540	10719	9535	10506
1975	15	9621	10729	9987	11115
1976	16	11909	10504	10418	11065
1977	17	10577	11346	10825	11158
1978	18	9874	11502	11207	11334
1979	19	14749	11294	11562	11242
1980	20	10403	.	11891	11190
1981	21	10867	.	12194	11139
1982	22	13840	.	12470	10832
1983	23	13526	.	12723	9549
1984	24	.	.	12950	12510
1985	25	.	.	13135	13305
1986	26	.	.	13340	12717
1987	27	.	.	13505	12752
1988	28	.	.	13652	12757
1989	29	.	.	13784	12722
1990	30	.	.	13900	12543
1991	31	.	.	14002	12518
1992	32	.	.	14075	12349
1993	33	.	.	14175	12124
1994	34	.	.	14246	11350
1995	35	.	.	14309	11520
1996	36	.	.	14363	11137
1997	37	.	.	14412	13697
1998	38	.	.	14454	10191
1999	39	.	.	14491	9619
2000	40	.	.	14524	8976
2001	41	.	.	14552	8273
2002	42	.	.	14577	7494
2003	43	.	.	14599	6638
2004	44	.	.	14618	5716
2005	45	.	.	14635	4692
2006	46	.	.	14649	3619
2007	47	.	.	14662	2437
2008	48	.	.	14673	1188
2009	49	.	.	14683	.
2010	50	.	.	14691	.

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YEAR	TIME	CASPERTA	MOVAVERA	AGOKI
1960	0	1.10	.	1.14
1961	1	1.10	.	1.18
1962	2	1.20	1.20	1.21
1963	3	1.30	1.24	1.25
1964	4	1.30	1.28	1.29
1965	5	1.30	1.32	1.32
1966	6	1.30	1.30	1.35
1967	7	1.40	1.34	1.38
1968	8	1.20	1.38	1.41
1969	9	1.50	1.40	1.44
1970	10	1.50	1.44	1.47
1971	11	1.40	1.52	1.49
1972	12	1.60	1.56	1.52
1973	13	1.60	1.56	1.54
1974	14	1.70	1.64	1.56
1975	15	1.50	1.68	1.58
1976	16	1.80	1.64	1.60
1977	17	1.80	1.66	1.61
1978	18	1.40	.	1.63
1979	19	1.80	.	1.64
1980	20	1.70	.	1.66
1981	21	1.80	.	1.67
1982	22	1.30	.	1.68
1983	23	1.70	.	1.69
1984	24	.	.	1.70
1985	25	.	.	1.71
1986	26	.	.	1.72
1987	27	.	.	1.72
1988	28	.	.	1.73
1989	29	.	.	1.74
1990	30	.	.	1.74
1991	31	.	.	1.75
1992	32	.	.	1.75
1993	33	.	.	1.76
1994	34	.	.	1.76
1995	35	.	.	1.76
1996	36	.	.	1.77
1997	37	.	.	1.77
1998	38	.	.	1.77
1999	39	.	.	1.77
2000	40	.	.	1.78
2001	41	.	.	1.78
2002	42	.	.	1.78
2003	43	.	.	1.78
2004	44	.	.	1.78
2005	45	.	.	1.79
2006	46	.	.	1.79
2007	47	.	.	1.79
2008	48	.	.	1.79
2009	49	.	.	1.79
2010	50	.	.	1.79

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YEAR	MOTORIZA	CASPERTA	MOVAVERA	AGOKI
1960	110.29	1.10	.	1.36
1961	101.22	1.10	.	1.24
1962	100.88	1.20	1.20	1.24
1963	98.42	1.30	1.24	1.20
1964	101.69	1.30	1.28	1.25
1965	104.69	1.30	1.32	1.29
1966	106.99	1.30	1.30	1.32
1967	110.01	1.40	1.34	1.36
1968	111.00	1.20	1.38	1.37
1969	113.63	1.50	1.40	1.40
1970	122.05	1.50	1.44	1.48
1971	129.05	1.40	1.52	1.53
1972	132.30	1.60	1.56	1.55
1973	131.33	1.60	1.56	1.55
1974	142.31	1.70	1.64	1.59
1975	148.89	1.50	1.68	1.60
1976	146.30	1.30	1.64	1.60
1977	149.39	1.80	1.66	1.60
1978	151.54	1.40	1.50	1.60
1979	150.40	1.80	1.50	1.60
1980	149.77	1.70	.	1.60
1981	149.17	1.80	.	1.60
1982	145.70	1.80	.	1.59
1983	133.58	1.70	.	1.56
1984	172.40	.	.	1.52
1985	176.40	.	.	1.49
1986	180.50	.	.	1.45
1987	184.70	.	.	1.41
1988	189.00	.	.	1.37
1989	193.30	.	.	1.31
1990	197.70	.	.	1.18
1991	202.20	.	.	1.11
1992	206.70	.	.	1.11
1993	211.40	.	.	1.03
1994	216.10	.	.	.94
1995	220.90	.	.	.84
1996	225.70	.	.	.73
1997	230.80	.	.	.62
1998	235.70	.	.	.49
1999	240.70	.	.	.36
2000	245.90	.	.	.21
2001	251.10	.	.	.06
2002	256.40	.	.	.
2003	261.80	.	.	.
2004	267.20	.	.	.
2005	272.80	.	.	.
2006	278.30	.	.	.
2007	284.00	.	.	.
2008	289.70	.	.	.
2009	295.50	.	.	.
2010	301.40	.	.	.

YEAR	TIME	INJURIES	MOVINGAV	ABOK11	ABOK12
1949	0	1994	.	.	.
1950	1	1049	1155	1033	.
1951	2	1231	1336	1291	27
1952	3	1281	1261	1441	529
1953	4	1125	1524	1606	290
1954	5	1619	1786	1145	256
1955	6	2363	2061	1784	321
1956	7	2498	2406	2015	285
1957	8	2498	2200	2112	375
1958	9	2953	2660	2784	381
1959	10	2890	2983	3095	387
1960	11	3975	3183	3627	398
1961	12	3701	3457	3612	372
1962	13	3846	3734	3909	436
1963	14	4236	4032	3971	444
1964	15	4381	4238	3596	487
1965	16	4471	4220	3962	453
1966	17	4146	4600	4303	452
1967	18	5089	4978	4567	456
1968	19	5078	4665	4918	433
1969	20	5399	4633	5034	448
1970	21	6812	5370	5345	455
1971	22	7509	5977	5977	473
1972	23	9197	6799	6585	451
1973	24	9595	7702	6997	487
1974	25	9187	7814	7408	420
1975	26	8283	8754	7654	425
1976	27	10269	8600	7530	431
1977	28	9017	8975	8966	436
1978	29	8286	9270	8975	441
1979	30	13087	9008	9856	446
1980	31	8980	9336	9775	454
1981	32	11167	9788	9924	421
1982	33	12378	9680	10221	368
1983	34	13526	10007	10063	373
1984	35	.	10315	9976	454
1985	36	.	10107	9894	499
1986	37	.	.	.	539

YEAR					
1949		362 0.0	136 540		
1950		1 1195	130 1056		
1951	2 1253 1423	1391 1173			
1952	3 1160 1187	1419 1334			
1953	4 1024 1282	1447 1303			
1954	5 1301 1369	1475 1462			
1955	6 1662 1432	1502 1623			
1956	7 1496 1533	1530 1579			
1957	8 1477 1576	1557 1621			
1958	9 1556 1599	1584 1686			
1959	10 1488 1610	1610 1685			
1960	11 1776 1659	1637 1815			
1961	12 1751 1747	1663 1718			
1962	13 1775 1828	1689 1718			
1963	14 1946 1811	1715 1683			
1964	15 1893 1807	1740 1723			
1965	16 1698 1824	1745 1758			
1966	17 1733 1792	1790 1792			
1967	18 1840 1782	1814 1818			
1968	19 1772 1664	1838 1821			
1969	20 1787 1670	1862 1844			
1970	21 1925 1894	1885 1904			
1971	22 2006 2003	1926 1935			
1972	23 2209 2125	1951 1950			
1973	24 2337 2060	1952 1947			
1974	25 1996 2052	1974 1964			
1975	26 1659 1940	1995 1957			
1976	27 2019 1767	2016 1953			
1977	28 1669 2255	2036 1956			
1978	29 1470 1766	2056 1951			
1979	30 2256 1745	2075 1954			
1980	31 1494 .	2094 1955			
1981	32 1915 .	2112 1907			
1982	33 1987 .	2130 1958			
1983	34 2156 .	2142 1953			
1984	35 . .	2164 1830			
1985	36 . .	2181 1792			
1986	37 . .	2197 1748			
1987	38 . .	2213 1699			
1988	39 . .	2229 1642			
1989	40 . .	2243 1580			
1990	41 . .	2258 1512			
1991	42 . .	2272 1435			
1992	43 0.0 .	2285 1353			
1993	44 . .	2299 1261			
1994	45 . .	2311 1162			
1995	46 . .	2324 1054			
1996	47 . .	2336 940			
1997	48 . .	2348 817			
1998	49 . .	2359 680			
1999	50 . .	2370 540			
2000	51 . .	2380 385			
2001	52 . .	2391 222			
2002	53 . .	2401 49.0			
2003	54 . .	2410 .			
2004	55 . .	2419 .			
2005	56 . .	2428 .			
2006	57 . .	2437 .			
2007	58 . .	2445 .			
2008	59 . .	2450 .			
2009	60 . .	2461 .			
2010	61 . .	2469 .			

YEAR				
1949	5456 5633 .	100 2299		
1950	5346 1880 .	1098 2084		
1951	6877 2171 2356 1663 2534			
1952	7668 2224 2184 2451 2690			
1953	7509 1923 8589 2298 2850			
1954	8368 2722 2927 3094 3015			
1955	9401 3907 3175 3951 3184			
1956	9098 3858 3558 3711 3357			
1957	9386 3465 3747 3940 3533			
1958	9870 3838 3945 4304 3711			
1959	9856 4060 4294 3892			
1960	1102 4898 4262 5079 4073			
1961	1012 4431 4452 4495 4256			
1962	1008 4476 4681 4461 4438			
1963	9842 4788 4587 4284 4619			
1964	1019 4812 4628 4517 4800			
1965	1049 4427 4756 4722 4979			
1966	1069 4637 4793 4873 5152			
1967	1101 5115 4813 5052 5925			
1968	1110 4974 5143 5122 5494			
1969	1133 4934 5500 5277 5658			
1970	1235 5127 5990 5704 5818			
1971	1325 6431 6508 5795 5978			
1972	1329 7606 6837 6189 6122			
1973	1313 7674 6861 6132 6265			
1974	1421 6429 7059 6590 6403			
1975	1489 6175 6795 6661 6535			
1976	1420 7413 6374 6648 6661			
1977	1499 6294 6785 6671 6781			
1978	1514 5570 5550 6715 6395			
1979	1500 8433 6540 6692 7003			
1980	1497 5594 .	6677 7106		
1981	1497 6758 .	6667 7202		
1982	1450 7239 .	6586 7293		
1983	1333 7613 .	6213 7378		
1984	1720 . .	6373 7458		
1985	1760 . .	5852 7533		
1986	1800 . .	6814 7503		
1987	1840 . .	6757 7669		
1988	1890 . .	6630 7730		
1989	1930 . .	6583 7787		
1990	1970 . .	6465 7840		
1991	2020 . .	6323 7989		
1992	2060 0.0 .	6161 7935		
1993	2110 . .	5969 7977		
1994	2160 . .	5754 8017		
1995	2200 . .	5511 8053		
1996	2250 . .	5245 8087		
1997	2300 . .	4948 5118		
1998	2350 . .	4613 8147		
1999	2400 . .	4259 8174		
2000	2450 . .	3863 8198		
2001	2510 . .	3439 8221		
2002	2560 . .	2978 8242		
2003	2610 . .	2479 8261		
2004	2670 . .	1951 8279		
2005	2720 . .	1370 8296		
2006	2780 . .	769 8311		
2007	2840 . .	113 8325		
2008	2890 . .	. 8338		
2009	2950 . .	. 8349		
2010	3010 . .	. 8359		

YEAR	TIME	VEHICLES	MCYCLES	PERCENC	MOVINGAC
1960	0	89505	4769	5.33	.
1961	1	84540	4600	5.44	. 5.3
1962	2	87130	4713	5.41	5.43 5.5
1963	3	87073	4736	5.44	5.54 5.6
1964	4	92581	5100	5.51	5.65 5.7
1965	5	93258	5793	5.90	5.77 5.8
1966	6	103175	6164	5.97	5.92 5.8
1967	7	109439	6593	6.02	6.10 5.9
1968	8	113329	7004	6.13	6.20 5.9
1969	9	124348	7970	6.41	6.31 6.0
1970	10	137271	8823	6.43	6.45 6.0
1971	11	149750	9774	6.53	6.30 6.0
1972	12	159969	10681	6.68	6.14 6.0
1973	13	164222	8966	5.46	5.99 6.1
1974	14	184086	10332	5.61	5.85 6.1
1975	15	199715	11312	5.66	5.70 6.1
1976	16	203446	11870	5.83	5.83 6.1
1977	17	214351	12763	5.95	5.96 6.2
1978	18	225447	13746	6.10	6.11 6.2
1979	19	232029	14573	6.28	6.27 6.2
1980	20	240435	15343	6.38	6.43 6.2
1981	21	246132	16345	6.64	6.55 6.2
1982	22	249162	16870	6.77	. 6.2
1983	23	250919	16823	6.70	. 6.2
1984	24	.	.	.	. 6.3
1985	25	.	.	.	. 6.3
1986	26	.	.	.	. 6.33
1987	27	.	.	.	. 6.3
1988	28	.	.	.	. 6.3
1989	29	.	.	.	. 6.3
1990	30	.	.	.	. 6.3
1991	31	.	.	.	. 6.3
1992	32	.	.	.	. 6.4
1993	33	.	.	.	. 6.40
1994	34	.	.	.	. 6.4
1995	35	.	.	.	. 6.4
1996	36	.	.	.	. 6.4
1997	37	.	.	.	. 6.4
1998	38	.	.	.	. 6.4
1999	39	.	.	.	. 6.4
2000	40	.	.	.	. 6.4
2001	41	.	.	.	. 6.4
2002	42	.	.	.	. 6.4
2003	43	.	.	.	. 6.4
2004	44	.	.	.	. 6.4
2005	45	.	.	.	. 6.5
2006	46	.	.	.	. 6.5
2007	47	.	.	.	. 6.5
2008	48	.	.	.	. 6.5
2009	49	.	.	.	. 6.5
2010	50	.	.	.	. 6.5

TIME	Y	A	B	C	D	E	F	G	H	I	J
1960	0	99505	64594	7217	.	.	2788	1885	6761	.	.
1961	1	84540	64163	7590	.	7643	2377	1664	7000	.	7416
1962	2	87130	65673	7560	7476	7493	2469	1745	7068	7019	7283
1963	3	87073	65511	7524	7493	7405	2385	1680	7044	7107	7205
1964	4	92581	69357	7491	7423	7343	2450	1769	7220	7256	7150
1965	5	98259	71750	7302	7347	7295	2542	1831	7203	7292	7107
1966	6	103175	74697	7240	7264	7256	2445	1894	7746	7242	7072
1967	7	109439	73554	7178	7179	7222	2947	2135	7245	7178	7042
1968	8	113329	80556	7108	7116	7194	3129	2127	6798	7054	7016
1969	9	124349	87897	7069	7066	7168	2969	1979	6898	6922	6994
1970	10	137371	95573	6984	7036	7145	3546	2334	6582	6858	6973
1971	11	144758	104953	6793	7089	7125	4217	2988	7035	6834	6955
1972	12	184448	111695	6790	7037	7106	4548	3239	6925	6968	6933
1973	13	194225	118945	7151	7063	7099	4695	3252	6927	7032	6923
1974	14	194466	132535	7115	7085	7073	4979	3205	7317	7037	6909
1975	15	199915	142025	7111	7113	7053	4951	3540	7150	6951	6895
1976	16	202446	144490	7102	7102	7044	4605	2185	6373	6978	6883
1977	17	214251	152536	7116	7081	7031	4536	3011	6628	6861	6871
1978	18	225147	159324	7087	7066	7013	5625	3888	6912	6714	6360
1979	19	234129	168580	7011	7045	7007	6216	4246	5821	6718	6850
1980	20	240425	169153	7036	7024	6996	4485	2832	6314	6736	6840
1981	21	246132	172166	6995	7017	6985	5648	3889	6386	6690	6831
1982	22	242148	174374	7010	.	6975	5381	4029	6736	.	6822
1983	23	.	176470	7035	.	6965	6362	4253	6585	.	6813
1984	24	.	.	.	6750	.	.	.	.	.	6805
1985	25	.	.	.	6947	.	.	.	.	.	6797
1986	26	.	.	.	6939	.	.	.	.	.	6790
1987	27	.	.	.	6931	.	.	.	.	.	6782
1988	28	.	.	.	6923	.	.	.	.	.	6775
1989	29	.	.	.	6915	.	.	.	.	.	6769
1990	30	.	.	.	6905	.	.	.	.	.	6762
1991	31	.	.	.	6901	0	.	.	.	.	6756
1992	32	.	.	.	6894	.	.	.	.	.	6750
1993	33	.	.	.	6887	.	.	.	.	.	6744
1994	34	.	.	.	6881	.	.	.	.	.	6738
1995	35	.	.	.	6875	.	.	.	.	.	6733
1996	36	.	.	.	6869	.	.	.	.	.	6727
1997	37	.	.	.	6863	.	.	.	.	.	6722
1998	38	.	.	.	6857	.	.	.	.	.	6717
1999	39	.	.	.	6851	0	.	.	.	.	6712
2000	40	.	.	.	6846	.	.	.	.	.	6707
2001	41	.	.	.	6841	.	.	.	.	.	6702
2002	42	.	.	.	6835	.	.	.	.	.	6697
2003	43	.	.	.	6830	.	.	.	.	.	6693
2004	44	.	.	.	6825	.	.	.	.	.	6689
2005	45	.	.	.	6820	.	.	.	.	.	6684
2006	46	.	.	.	6816	.	.	.	.	.	6680
2007	47	.	.	.	6811	.	.	.	.	.	6676
2008	48	.	.	.	6807	.	.	.	.	.	6672
2009	49	.	.	.	6802	.	.	.	.	.	6668
2010	50	.	.	.	6798	.	.	.	.	.	6664

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YEAR	TIME	RTAS	DAYRTAS	PERCENT	MOVINGAV	AGOKI
1960	0	4102	3125	76.18	.	.
1961	1	3573	2730	76.41	.	81.07
1962	2	3595	2647	73.63	74.67	77.15
1963	3	3578	2616	73.11	73.94	74.86
1964	4	3693	2734	74.03	73.20	73.23
1965	5	3562	2583	72.52	73.11	71.97
1966	6	3847	2798	72.73	73.33	70.94
1967	7	4387	3209	73.15	72.82	70.07
1968	8	4511	3349	74.24	72.46	69.31
1969	9	4196	3000	71.50	71.89	68.65
1970	10	5163	3648	70.66	70.61	68.05
1971	11	6042	4222	69.88	69.64	67.51
1972	12	6613	4415	66.76	68.57	67.02
1973	13	6789	4710	69.38	61.84	66.57
1974	14	6250	2261	36.18	61.20	66.15
1975	15	6534	4379	67.02	61.33	65.76
1976	16	6534	4357	66.68	60.11	65.39
1977	17	5949	4010	67.41	65.55	65.05
1978	18	6956	4402	63.26	66.36	64.72
1979	19	8049	5099	63.35	66.13	64.42
1980	20	6162	4381	71.10	66.43	64.13
1981	21	7250	4748	65.49	67.59	63.85
1982	22	7524	5186	68.93	.	63.59
1983	23	8023	5544	69.10	.	63.34
1984	24	.	.	.	.	63.10
1985	25	.	.	.	.	62.97
1986	26	.	.	.	.	62.84
1987	27	.	.	.	.	62.42
1988	28	.	.	.	.	62.23
1989	29	.	.	.	.	62.03
1990	30	.	.	.	.	61.95
1991	31	.	.	.	.	61.65
1992	32	.	.	.	.	61.47
1993	33	.	.	.	.	61.30
1994	34	.	.	.	.	61.13
1995	35	.	.	.	.	60.96
1996	36	.	.	.	.	60.80
1997	37	.	.	.	.	60.65
1998	38	.	.	.	.	60.50
1999	39	.	.	.	.	60.35
2000	40	.	.	.	.	60.21
2001	41	.	.	.	.	60.07
2002	42	.	.	.	.	59.93
2003	43	.	.	.	.	59.80
2004	44	.	.	.	.	59.67
2005	45	.	.	.	.	59.54
2006	46	.	.	.	.	59.42
2007	47	.	.	.	.	59.30
2008	48	.	.	.	.	59.18
2009	49	.	.	.	.	59.06
2010	50	.	.	.	.	58.95

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YEAR	TIME	TOTELEME	PEDALCYC	PERCENT	MOVINGAV	AGOKI
1960	0	3515	429	12.20	.	.
1961	1	3473	398	11.46	.	11.75
1962	2	3595	340	9.46	10.59	10.36
1963	3	3578	374	10.45	9.84	9.54
1964	4	3693	346	9.37	9.02	8.97
1965	5	3564	302	8.47	8.63	8.53
1966	6	3991	293	7.34	7.87	8.16
1967	7	4321	324	7.50	7.55	7.85
1968	8	4511	302	6.69	7.30	7.58
1969	9	4103	319	7.77	7.24	7.35
1970	10	5163	372	7.21	6.92	7.14
1971	11	6042	425	7.03	7.01	6.95
1972	12	6622	392	5.92	6.55	6.77
1973	13	6799	484	7.13	6.07	6.61
1974	14	6250	342	5.47	5.65	6.46
1975	15	6534	312	4.78	6.45	6.33
1976	16	6548	325	4.96	6.94	6.20
1977	17	5703	565	9.91	5.73	6.07
1978	18	6956	665	9.56	6.57	5.96
1979	19	8049	356	4.42	6.65	5.85
1980	20	6162	246	3.99	5.76	5.75
1981	21	7122	383	5.38	4.75	5.65
1982	22	7524	410	5.45	.	5.56
1983	23	8023	363	4.52	.	5.47
1984	24	.	.	.	.	5.38
1985	25	.	.	.	.	5.30
1986	26	.	.	.	.	5.22
1987	27	.	.	.	.	5.15
1988	28	.	.	.	.	5.08
1989	29	.	.	.	.	5.01
1990	30	.	.	.	.	4.94
1991	31	.	.	.	.	4.87
1992	32	.	.	.	.	4.81
1993	33	.	.	.	.	4.75
1994	34	.	.	.	.	4.69
1995	35	.	.	.	.	4.63
1996	36	.	.	.	.	4.57
1997	37	.	.	.	.	4.52
1998	38	.	.	.	.	4.46
1999	39	.	.	.	.	4.41
2000	40	.	.	.	.	4.36
2001	41	.	.	.	.	4.31
2002	42	.	.	.	.	4.26
2003	43	.	.	.	.	4.22
2004	44	.	.	.	.	4.17
2005	45	.	.	.	.	4.13
2006	46	.	.	.	.	4.08
2007	47	.	.	.	.	4.04
2008	48	.	.	.	.	4.00
2009	49	.	.	.	.	3.95
2010	50	.	.	.	.	3.91

	TIME	Y	A	B	C	D	E	F	G	H
1960	0	3515	129	367	.	.		169	481	.	.
1961	1	3473	84	242	.	353	614	1768	.	7
1962	2	3595	90	250	271	325	696	1936	1717	13
1963	3	3578	93	260	242	309	791	2211	1981	16
1964	4	3693	88	238	357	297	809	2191	2092	19
1965	5	3564	79	222	349	288	641	1799	2146	21
1966	6	3991	325	814	345	281	928	2325	2135	23
1967	7	4321	92	213	339	275	953	2206	2102	24
1968	8	4511	108	239	330	270	972	2155	2189	25
1969	9	4103	84	205	200	265	831	2025	2254	26
1970	10	5163	92	176	211	261	1153	2293	2531	27
1971	11	6042	101	167	562	257	1677	3653	3718	28
1972	12	6622	177	267	243	253	1560	2937	2718	29
1973	13	6799	193	292	241	250	2114	3114	2813	30
1974	14	6250	193	309	258	247	1559	3036	2957	31
1975	15	6534	112	171	231	244	1797	2750	2575	31
1976	16	6548	99	151	214	242	2166	3303	4051	32
1977	17	5703	132	231	234	229	2233	5569	4023	32
1978	18	6956	143	206	245	237	2320	5498	3771	33
1979	19	8049	329	409	250	225	2253	2923	3791	33
1980	20	6162	139	226	225	233	901	1462	3763	34
1981	21	7122	127	176	242	231	2267	3211	3835	34
1982	22	7524	116	154	.	229	3429	3228	.	350
1983	23	8023	196	244	.	227	2565	3350	.	354
1984	24	.	.	.	.	226	.	.	.	352
1985	25	.	.	.	.	224	.	.	.	358
1986	26	.	.	.	.	224	.	.	.	365
1987	27	.	.	.	.	221	.	.	.	369
1988	28	.	.	.	.	219	.	.	.	372
1989	29	.	.	.	.	218	.	.	.	375
1990	30	.	.	.	.	217	.	.	.	378
1991	31	.	.	.	.	215	.	.	.	381
1992	32	.	.	.	.	214	.	.	.	384
1993	33	.	.	.	.	213	.	.	.	387
1994	34	.	.	.	.	212	.	.	.	390
1995	35	.	.	.	.	211	.	.	.	392
1996	36	.	.	.	.	209	.	.	.	395
1997	37	.	.	.	.	208	.	.	.	397
1998	38	.	.	.	.	207	.	.	.	400
1999	39	.	.	.	.	206	.	.	.	402
2000	40	.	.	.	.	205	.	.	.	404
2001	41	.	.	.	.	204	.	.	.	407
2002	42	.	.	.	.	203	.	.	.	409
2003	43	.	.	.	.	202	.	.	.	411
2004	44	.	.	.	.	201	.	.	.	413
2005	45	.	.	.	.	200	.	.	.	417
2006	46	.	.	.	.	200	.	.	.	419
2007	47	.	.	.	.	199	.	.	.	421
2008	48	.	.	.	.	198	.	.	.	423
2009	49	.	.	.	.	197	.	.	.	423
2010	50	.	.	.	.	196	.	.	.	425

TIME	Y	A	B	C	D	E	F	G	H	IU	J
1973	0	1402	1222	8716	.	.	9595	8871	9245	.	.
1974	1	1352	1183	8750	.	8605	9187	8439	9186	.	9101
1975	2	1338	1145	8558	8424	8405	8283	7553	9119	8937	8930
1976	3	1640	1388	8463	8370	8288	10269	9268	9025	8905	8830
1977	4	1560	1191	7635	8253	8205	9017	7313	8110	8754	8759
1978	5	1588	1341	8445	7707	814?	10856	9865	9087	8572	8704
1979	6	1556	1357	8165	7?0?	8?66	13087	11029	8427	8556	8059
1980	7	?13	9?5	862?	8?85	8044	9990	7362	9211	8797	8621
1981	8	1720	3491	9?1?	815?	80?5	11157	7787	8943	3?57	8630
1982	9	1462	1244	8509	.	7971	12378	10773	8867	.	8559
1983	10.5	10.5	1300	8561	.	7?41	10525	11951	8843	.	8533
1984	11	.	.	.	.	7913	.	.	.	.	8509
1985	12	.	.	.	.	7898	.	.	.	.	8488
1986	13	.	.	.	.	7865	.	.	.	.	8468
1987	14	.	.	.	.	7844	.	.	.	.	8450
1988	15	.	.	.	.	7824	.	.	.	.	8433
1989	16	.	.	.	.	7805	.	.	.	.	8417
1990	17	.	.	.	.	7786	.	.	.	.	8402
1991	18	0	.	.	.	7771	.	.	.	.	8388
1992	19	.	.	.	.	7756	.	.	.	.	8374
1993	20	.	.	.	.	7741	.	.	.	.	8362
1994	21	.	.	.	.	7727	.	.	.	.	8350
1995	22	.	.	.	.	7714	.	.	.	.	8338
1996	23	.	.	.	.	7701	.	.	.	.	8327
1997	24	.	.	.	.	7689	.	.	.	.	8317
1998	25	.	.	.	.	7677	.	.	.	.	8307
1999	26	.	.	.	.	7663	.	.	.	.	8297
2000	27	.	.	.	.	7655	.	.	.	.	8288
2001	28	.	.	.	.	7644	.	.	.	.	8279
2002	29	.	.	.	.	7634	.	.	.	.	8270
2003	30	.	.	.	.	7624	.	.	.	.	8262
2004	31	.	.	.	.	7615	.	.	.	.	8254
2005	32	.	.	.	.	7606	.	.	.	.	8246
2006	33	.	.	.	.	7597	.	.	.	.	8238
2007	34	.	.	.	.	7589	.	.	.	.	8231
2008	35	.	.	.	.	7580	.	.	.	.	8224
2009	36	.	.	.	.	7572	.	.	.	.	8217
2010	37	.	.	.	.	7564	.	.	.	.	8210

YEAR	TIME	DEATHS	MCYCLIST
1973	0	1402	24
1974	1	1351	17
1975	2	1338	15
1976	3	1640	28
1977	4	1560	21
1978	5	1588	21
1979	6	1662	24
1980	7	1413	123
1981	8	1720	45
1982	9	1462	36
1983	10	1515	31
1984	11	.	.
1985	12	.	.
1986	13	.	.
1987	14	.	.
1988	15	.	.
1989	16	.	.
1990	17	.	.
1991	18	.	.
1992	19	.	.
1993	20	.	.
1994	21	.	.
1995	22	.	.
1996	23	.	.
1997	24	.	.
1998	25	.	.
1999	26	.	.
2000	27	.	.
2001	28	.	.
2002	29	.	.
2003	30	.	.
2004	31	.	.
2005	32	.	.
2006	33	.	.
2007	34	.	.
2008	35	.	.
2009	36	.	.
2010	37	.	.

PCENTMC	MOVINGAM	AGOKIM	PEDALCYC	PCENTPCY	MOVINGAP	AGOKIP
1.71	.	.	129	9.20	.	.
1.26	.	-.80	72	5.33	.	10.40
1.12	1.43	.96	103	7.70	9.02	8.28
1.71	1.35	1.68	130	7.93	6.96	7.05
1.35	1.39	2.19	155	9.94	6.87	6.17
1.32	2.90	2.59	64	4.03	5.56	5.49
1.44	3.09	2.91	79	4.75	4.83	4.93
8.70	3.31	3.18	15	1.13	3.98	4.46
2.62	3.45	3.42	74	4.30	4.20	4.06
2.46	.	3.63	83	5.68	.	3.70
2.05	.	3.82	78	5.15	.	3.38
.	.	3.99	.	.	.	3.09
.	.	4.14	.	.	.	2.82
.	.	4.23	.	.	.	2.59
.	.	4.42	.	.	.	2.35
.	.	4.54	.	.	.	2.14
.	.	4.65	.	.	.	1.94
.	.	4.76	.	.	.	1.76
.	.	4.86	.	.	.	1.59
.	.	4.96	.	.	.	1.42
.	.	5.05	.	.	.	1.26
.	.	5.14	.	.	.	1.12
.	.	5.22	.	.	.	.97
.	.	5.30	.	.	.	.64
.	.	5.37	.	.	.	.71
.	.	5.45	.	.	.	.58
.	.	5.52	.	.	.	.46
.	.	5.58	.	.	.	.35
.	.	5.65	.	.	.	.24
.	.	5.71	.	.	.	.13
.	.	5.77	.	.	.	.03
.	.	5.83	.	.	.	.
.	.	5.89	.	.	.	.
.	.	5.94	.	.	.	.
.	.	5.99	.	.	.	.
.	.	6.05	.	.	.	.
.	.	6.10	.	.	.	.
.	.	6.14	.	.	.	.

YEAR	TIME	DEATHS	PEDESTRI	PERCENT	MOVINGAV	AGOKI	PASSENGE	PERCENTP	MOVINGAG	AGOKI2
1973	0	1402	597	42.58	.	.	445	31.74	.	.
1974	1	1351	528	39.08	.	39.74	487	36.05	.	34.24
1975	2	1338	591	44.17	41.12	41.14	454	33.93	34.74	34.54
1976	3	1640	650	39.63	42.17	41.96	601	36.65	34.84	34.72
1977	4	1560	626	40.13	42.56	42.54	551	35.32	34.67	34.84
1978	5	1588	760	47.86	42.17	43.00	512	32.24	34.83	34.94
1979	6	1662	682	41.03	43.68	43.36	585	35.20	34.48	35.02
1980	7	1413	596	42.18	44.52	43.68	491	34.75	34.57	35.09
1981	8	1720	812	47.21	43.41	43.95	600	34.88	36.16	35.14
1982	9	1462	648	44.32	.	44.18	523	35.77	.	35.20
1983	10	1515	641	42.31	.	44.40	609	40.20	.	35.24
1984	11	.	.	.	.	44.59	.	.	.	35.26
1985	12	.	.	.	.	44.77	.	.	.	35.32
1986	13	.	.	.	.	44.93	.	.	.	35.35
1987	14	.	.	.	.	45.08	.	.	.	35.39
1988	15	.	.	.	.	45.22	.	.	.	35.44
1989	16	.	.	.	.	45.35	.	.	.	35.44
1990	17	.	.	.	.	45.47	.	.	.	35.47
1991	18	.	.	.	.	45.59	.	.	.	35.50
1992	19	.	.	.	.	45.70	.	.	.	35.52
1993	20	.	.	.	.	45.80	.	.	.	35.54
1994	21	.	.	.	.	45.90	.	.	.	35.56
1995	22	.	.	.	.	45.99	.	.	.	35.58
1996	23	.	.	.	.	46.08	.	.	.	35.60
1997	24	.	.	.	.	46.17	.	.	.	35.62
1998	25	.	.	.	.	46.25	.	.	.	35.64
1999	26	.	.	.	.	46.33	.	.	.	35.66
2000	27	.	.	.	.	46.41	.	.	.	35.67
2001	28	.	.	.	.	46.48	.	.	.	35.69
2002	29	.	.	.	.	46.55	.	.	.	35.70
2003	30	.	.	.	.	46.62	.	.	.	35.72
2004	31	.	.	.	.	46.69	.	.	.	35.73
2005	32	.	.	.	.	46.75	.	.	.	35.75
2006	33	.	.	.	.	46.81	.	.	.	35.76
2007	34	.	.	.	.	46.87	.	.	.	35.77
2008	35	.	.	.	.	46.93	.	.	.	35.79
2009	36	.	.	.	.	46.99	.	.	.	35.80
2010	37	.	.	.	.	47.04	.	.	.	35.81

TIMT	Y	A	B	C	D	E	F	G	H	I
1973	0	9595	2082	2170	.	.	4758	4959	.	.
1974	1	9187	1975	2150	.	2964	4997	5439	.	5209
1975	2	8283	1695	2046	2504	2637	4266	5150	5242	5234
1976	3	10269	2635	2566	2457	2446	5625	5487	5242	5249
1977	4	9017	3228	2530	2443	2310	4866	5175	5241	5260
1978	5	10856	2112	1945	2346	2205	5924	5457	5146	5269
1979	6	13087	2720	2078	2243	2118	6459	4935	5150	5275
1980	7	8990	1404	1562	1896	2046	4203	4675	5213	5280
1981	8	11157	2287	2048	1865	1983	6151	5508	5417	5285
1982	9	12378	2283	1796	.	1927	7412	5586	.	5294
1983	10	13526	2487	1839	.	1927	9158	5901	.	5293
1984	11	.	.	.	.	1832	.	.	.	5297
1985	12	.	.	.	.	1791	.	.	.	5300
1986	13	.	.	.	.	1754	.	.	.	5303
1987	14	.	.	.	.	1719	.	.	.	5306
1988	15	.	.	.	.	1686	.	.	.	5308
1989	16	.	.	.	.	1656	.	.	.	5311
1990	17	.	.	.	.	1627	.	.	.	5313
1991	18	0	.	.	.	1600	.	.	.	5315
1992	19	.	.	.	.	1574	.	.	.	5317
1993	20	.	.	.	.	1550	.	.	.	5319
1994	21	.	.	.	.	1527	.	.	.	5321
1995	22	.	.	.	.	1504	.	.	.	5322
1996	23	.	.	.	.	1483	.	.	.	5324
1997	24	.	.	.	.	1464	.	.	.	5326
1998	25	.	.	.	.	1445	.	.	.	5327
1999	26	0	.	.	.	1426	.	.	.	5328
2000	27	.	.	.	.	1409	.	.	.	5330
2001	28	.	.	.	.	1391	.	.	.	5331
2002	29	.	.	.	.	1375	.	.	.	5332
2003	30	.	.	.	.	1359	.	.	.	5334
2004	31	.	.	.	.	1343	.	.	.	5335
2005	32	.	.	.	.	1329	.	.	.	5336
2006	33	.	.	.	.	1314	.	.	.	5337
2007	34	.	.	.	.	1300	.	.	.	5338
2008	35	.	.	.	.	1286	.	.	.	5339
2009	36	.	.	.	.	1273	.	.	.	5340
2010	37	.	.	.	.	1260	.	.	.	5341

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YEAR	TIME	INJURIES	DRIVERS	PERCENT	MOVINGAV	AGOKI
1973	0	9595	2607	27.17	.	.
1974	1	9187	2464	26.82	.	19.19
1975	2	8283	1251	15.10	18.66	17.63
1976	3	10269	1124	10.95	16.69	16.72
1977	4	9017	1196	13.26	14.43	16.07
1978	5	10856	1879	17.31	14.54	15.57
1979	6	13087	2035	15.55	15.53	15.16
1980	7	8990	1403	15.61	15.64	14.81
1981	8	11167	1776	15.90	14.97	14.51
1982	9	12378	1712	13.83	.	14.24
1983	10	13526	1890	13.97	.	14.01
1984	11	.	.	.	.	13.79
1985	12	.	.	.	.	13.60
1986	13	.	.	.	.	13.42
1987	14	.	.	.	.	13.23
1988	15	.	.	.	.	13.09
1989	16	.	.	.	.	12.95
1990	17	.	.	.	.	12.81
1991	18	.	.	.	.	12.68
1992	19	.	.	.	.	12.56
1993	20	.	.	.	.	12.45
1994	21	.	.	.	.	12.34
1995	22	.	.	.	.	12.23
1996	23	.	.	.	.	12.13
1997	24	.	.	.	.	12.04
1998	25	.	.	.	.	11.94
1999	26	.	.	.	.	11.86
2000	27	.	.	.	.	11.77
2001	28	.	.	.	.	11.69
2002	29	.	.	.	.	11.61
2003	30	.	.	.	.	11.53
2004	31	.	.	.	.	11.46
2005	32	.	.	.	.	11.39
2006	33	.	.	.	.	11.32
2007	34	.	.	.	.	11.25
2008	35	.	.	.	.	11.19
2009	36	.	.	.	.	11.12
2010	37	.	.	.	.	11.06

TIMT	Y	A	B	C	D	E	F	G	H	I
1973	0	9595	278	289	.	.	704	734	.	.
1974	1	9187	193	613	.	222	563	613	.	484
1975	2	8283	169	204	334	300	539	651	514	534
1976	3	10269	340	331	339	346	535	521	538	563
1977	4	9017	211	284	298	379	316	350	509	584
1978	5	10826	240	313	423	404	501	554	482	600
1979	6	13087	537	410	423	423	615	471	684	613
1980	7	5990	752	636	460	442	1136	1264	647	624
1981	8	11167	414	371	461	457	539	483	629	634
1982	9	12378	455	368	.	470	576	465	.	643
1983	10	13526	433	320	.	482	628	464	.	650
1984	11	.	.	.	.	493	.	.	.	657
1985	12	.	.	.	.	503	.	.	.	663
1986	13	.	.	.	.	512	.	.	.	669
1987	14	.	.	.	.	520	.	.	.	675
1988	15	.	.	.	.	528	.	.	.	680
1989	16	.	.	.	.	535	.	.	.	684
1990	17	.	.	.	.	542	.	.	.	689
1991	18	.	.	.	.	548	.	.	.	693
1992	19	.	.	.	.	554	.	.	.	697
1993	20	.	.	.	.	560	.	.	.	700
1994	21	.	.	.	.	566	.	.	.	704
1995	22	.	.	.	.	571	.	.	.	707
1996	23	.	.	.	.	575	.	.	.	.
1997	24	.	.	.	.	581	.	.	.	.
1998	25	.	.	.	.	584	.	.	.	.
1999	26	.	.	.	.	590	.	.	.	719
2000	27	.	.	.	.	594	.	.	.	722
2001	28	.	.	.	.	598	.	.	.	725
2002	29	.	.	.	.	602	.	.	.	727
2003	30	.	.	.	.	606	.	.	.	730
2004	31	.	.	.	.	610	.	.	.	732
2005	32	.	.	.	.	613	.	.	.	734
2006	33	.	.	.	.	617	.	.	.	737
2007	34	.	.	.	.	620	.	.	.	739
2008	35	.	.	.	.	624	.	.	.	741
2009	36	.	.	.	.	627	.	.	.	743
2010	37	.	.	.	.	630	.	.	.	745

APPENDIX A.14

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HOURS	OBSDFLOW	MODFLOW	AGOKI1	AGOKI2
0	240	177	240	177
1	70	123	114	39
2	60	53	-10	-61
3	30	43	-55	-67
4	40	63	40	63
5	120	.	286	321
6	260	.	627	647
7	1180	880	764	950
8	1200	1147	1200	1147
9	1060	1087	1287	1203
10	1000	1003	1245	1143
11	950	1007	1146	1040
12	1070	973	1070	973
13	900	983	1064	986
14	980	1020	1116	1070
15	1180	1107	1170	1173
16	1160	1227	1160	1227
17	1340	1133	1054	1191
18	900	960	870	1065
19	640	680	665	888
20	500	707	500	707
21	980	580	405	554
22	260	487	362	430
23	220	240	321	310

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HOURS	OBSDFLOW	MODFLOW	AGOKI1	AGOKI2
0	7	10	7	10
1	6	5	4	6
2	2	3	2	2
3	1	4	2	1
4	9	7	9	7
5	11	.	22	21
6	18	.	41	40
7	45	47	60	61
8	78	78	78	78
9	112	106	92	90
10	127	119	102	98
11	117	121	111	105
12	120	115	120	115
13	107	110	130	128
14	104	116	138	140
15	136	124	140	145
16	132	141	132	141
17	154	142	114	123
18	140	116	89	96
19	54	78	62	66
20	39	41	37	41
21	31	27	23	24
22	12	20	14	16
23	17	12	10	12

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GRADIENT	GRADIEN1	RTA1	RTA2	OBSERVED	PRED1	PRED2
.5	.5	.53	2.73	1.41	1.49	.32
2.1	2.1	2.73	1.11	1.92	2.50	.57
3.9	3.9	1.80	0.0	3.34	2.50	1.02
5.0	5.6	1.67	.	1.67	1.92	1.36
6.0	.	.49	2.65	1.57	1.00	1.72

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DOWNGRAD	MODGRADE	RTA1	RTA2	MODRTA	OBSRTAS	AGOKI	SILYANOV
.4	.4	.52	5.45	2.99	2.99	3.03	.31
.8	.8	.49	6.28	3.39	3.39	3.05	.36
1.7	1.7	.49	5.91	3.20	3.20	3.03	.51
1.8	1.8	.49	5.00	2.75	2.75	3.02	.53
2.0	2.0	2.73	.	2.73	2.73	3.01	.57
2.9	2.9	.49	4.71	2.60	2.60	2.80	.76
3.8	3.8	.	2.73	2.73	2.73	2.67	.99
8.0	8.0	.	.49	.49	.49	.57	2.57

A.14/5

RTA1	RTA2	RTA3	RTA4	RTA5	MODELRTA	MODELSD	OBSEDSD	OBSEDRTA	AGOKI	SILYANOV
.49	.53	.	.	.	.	.	150	.51	.17	2.59
.91	1.06	1.11	1.48	2.61	.	.	175	1.48	1.06	2.37
1.06	2.22	2.94	.	.	1.41	187.5	200	2.07	1.78	2.18
.49	.53	.91	3.33	7.64	.	.	225	2.50	2.43	2.02
2.35	2.50	2.73	2.90	.	2.67	237.5	250	2.62	2.71	1.87
.49	.53	.98	1.06	1.11	.	.	275	3.17	2.92	1.74
2.50	2.73	9.09	10.0	.	.	.	.	.	.	.
1.11	1.18	1.82	1.67	2.12	2.87	287.5	300	2.45	2.96	1.63
2.61	2.65	3.33	3.45	4.50	.	.	.	.	.	.
.49	.53	.59	.83	.91	2.30	362.5	400	2.07	1.42	1.27
1.04	1.18	1.67	1.76	1.82	.	.	.	.	.	.
1.48	2.12	2.35	2.46	1.11	.	.	.	.	.	.
2.50	3.33	3.64	4.12	4.71	.	.	.	.	.	.
4.17	5.29	5.56	8.89	9.17	.	.	.	.	.	.

A.14/6

RTA1	RTA2	RTA3	RTA4	RTA5	RTA6	RTA7	MODELRTA	MODELSD	OBSDSD	OBSDRTAS	AGOKI	SILYANOV
.49	.53	.83	1.11	2.95	3.33	4.24	2.01	150.0	150	2.01	1.99	2.59
3.53	6.28	7.39	8.18	.	.	.	.	.	.	.	.	.
.52	1.04	2.94	5.45	.	.	.	2.94	237.5	250	2.48	3.03	1.87
.91	.98	1.57	1.67	2.50	3.71	8.89	.	.	275	2.89	3.11	1.74
1.18	1.67	1.82	4.17	4.55	4.71	.	.	.	300	3.02	3.10	1.63
.49	.63	.91	1.05	1.57	1.67	1.76	.	.	325	2.48	3.03	1.52
2.09	2.22	3.53	4.71	3.18	8.18	.	3.04	337.5	.	.	.	.
.49	1.76	5.29	.	.	.	.	.	.	375	2.51	2.64	1.35
.49	.52	.53	.59	1.76	2.61	2.95	2.60	400.0	400	2.60	2.34	1.27
1.82	.83	1.67	2.39	2.50	3.53	3.94	.	.	.	.	.	.
3.33	3.64	4.17	4.55	.	.	.	.	.	.	.	.	.

A.14/7

RTA1	RTA2	RTA3	MODELRTA	MODELWID	OBSDWIDT	OBSEDRTA	AGOKI	SILYANOV
7.59	.	.	.	.	5.80	7.59	8.08	1.26
8.47	.	.	8.03	5.82	5.84	8.47	7.91	1.25
8.18	.	.	.	.	5.92	8.12	7.57	1.23
8.32	.	.	.	.	5.99	8.32	7.28	1.21
9.17	.	.	7.19	6.00	6.08	9.17	6.90	1.20
7.67	.	.	.	.	6.28	7.67	6.05	1.14
6.42	.	.	5.84	6.33	6.42	6.42	5.46	1.11
5.76	.	.	.	.	6.48	5.76	5.20	1.10
1.67	7.49	8.62	.	.	6.51	5.93	5.08	1.09
6.34	.	.	5.00	6.53	6.55	6.34	4.91	1.08
5.91	.	.	.	.	6.56	5.91	4.86	1.08
5.83	.	.	.	.	6.58	5.83	4.78	1.08
3.82	5.09	6.13	.	.	6.60	5.01	4.69	1.07
4.73	5.83	.	.	.	6.62	5.28	4.61	1.07
5.82	.	.	.	.	6.65	5.82	4.48	1.06

A.14/8

RTA1	RTA2	RTA3	RTA4	RTA5	MODELRTA	OBSEDJNS	OBSEDRTA	AGOKI	JACOBS
.91	1.67	2.73	9.17	.83	5.88	0	5.88	5.07	1.45
2.50	4.17	10.00	20.90	.	.	.	.	.	.
1.11	7.50	7.27	6.67	10.59	6.80	1	6.80	6.74	2.47
2.62	11.82	.	.	.	.	.	.	.	.
1.11	3.33	3.64	5.00	5.24	7.28	2	7.28	8.41	3.49
6.36	7.78	9.85	13.26	17.27	.	.	.	.	.
3.45	7.06	9.41	15.76	.	8.92	3	8.92	10.08	4.51
5.00	10.00	24.55	.	.	13.18	4	13.18	11.76	5.53

A.14/9

RTA1	RTA2	RTA3	RTA4	RTA5	RTA6	RTA7	MODELRTA	OBSEDJNS	OBSEDRTA	AGOKI	JACOBS
2.09	2.50	3.53	3.64	4.17	5.00	.	3.49	0	3.49	3.01	1.45
1.82	3.64	6.40	6.89	5.88	6.67	4.55	5.79	1	5.79	5.93	2.47
6.36	7.88	7.78	.	.	.	.	.	.	.	.	.
1.11	2.72	5.83	5.45	4.44	8.37	12.94	8.47	2	8.47	8.86	3.49
16.47	18.89	.	.	.	.	.	.	.	.	.	.
5.42	5.76	5.83	.	.	.	.	.	3	5.67	11.78	4.51
14.40	.	.	.	.	.	.	14.40	4	14.40	14.71	5.53
.	.	.	.	.	.	.	.	5	.	17.63	6.55
20.91	.	.	.	.	.	.	20.91	6	20.91	20.55	7.57

A.14/15

RTA1	RTA2	RTA3	RTA04	RTA5	MODELRTA	MODFLOW	OBSDFLOW	OBSDRTAS	AGOKI	JACOBS	SILYANOV
.80	.	.	.	.	.80	80	80	.80	.76	.84	.29
.49	2.00	.	.	.	.	.	302	1.24	1.35	2.83	.39
.29	.86	2.57	3.14	.	.	.	417	1.72	1.81	3.87	.45
.29	.49	.57	1.71	2.57	.	.	.	.	.	.	.
3.14	6.57	.	.	.	1.85	464	475	2.19	2.09	4.39	.48
.86	1.71	2.57	3.14	.	.	.	500	2.07	2.22	4.62	.49
2.29	.	.	.	.	.	.	537	2.29	2.43	4.95	.51
.29	.86	5.43	6.57	.	3.19	627	628	3.29	2.98	5.71	.56
1.14	4.86	.	.	.	.	.	641	3.00	3.12	5.89	.57
.29	3.43	5.14	4.57	.	.	.	667	3.36	3.32	6.12	.59
2.86	.	.	.	.	3.49	723	779	2.86	4.29	7.13	.66
.29	7.14	9.14	.	.	6.17	920	840	5.52	4.91	7.68	.69
5.71	6.86	.	.	.	.	.	1000	6.29	6.83	9.12	.80

A.14/16

HOURS	TOTRTAS	OBSEDRTA	MODELRTA	AGOKI1	AGOKI2
0	9	.164064	.218736	.164064	.218736
1	6	.109368	.115440	.070451	.099420
2	4	.072912	.060744	.010700	.021971
3	0	0.0	.048600	.007880	.004586
4	4	.072912	.054696	.072912	.054696
5	5	.091140	.	.197020	.162901
6	12	.218760	.	.351902	.303610
7	27	.492192	.437520	.498586	.443270
8	33	.601584	.552960	.601584	.552950
9	31	.565128	.656280	.642035	.619436
10	44	.802104	.704680	.623767	.649190
11	41	.510432	.686640	.569667	.663463
12	28	.510432	.686640	.510432	.686640
13	44	.802104	.601560	.471350	.733950
14	27	.492192	.710952	.463273	.804318
15	46	.838560	.692712	.481294	.881430
16	41	.510432	.941856	.510432	.941857
17	68	1.239624	1.014758	.534226	.963380
18	58	1.057320	1.184928	.541346	.942294
19	69	1.257840	.935784	.527358	.874695
20	27	.492192	.772032	.492192	.772032
21	31	.565128	.455736	.436535	.645828
22	17	.309912	.419280	.360633	.505532
23	21	.382824	.285600	.266831	.359536

A.14/17

TOTJUNCT	RTA1	RTA2	RTA3	RTA4	RTA5	RTA6	RTA7	RTA8	MODELRTA	AGOKI	JACOBS
1	1.13	1.21	2.27	2.43	.	.	.	.	1.76	1.65	2.47
2	1.01	.86	1.13	4.29	2.57	2.43	3.64	2.27	2.20	2.37	3.49
3	.86	3.64	2.43	2.27	6.03	1.13	.	.	2.73	2.83	4.51
4	.76	1.13	6.86	3.40	.	.	.	.	3.04	3.04	5.53
5	1.53	2.27	4.93	.	.	.	.	.	2.91	2.98	6.55
6	.86	2.27	5.38	.	.	.	.	.	2.04	2.68	7.57
7	2.27	.	.	.	.	.	.	.	2.27	2.11	8.59
8	1.13	.	.	.	.	.	.	.	1.13	1.29	9.61

.14/18

HOURS	TOTRTAS	OBSEDRTA	MODELRTA	AGOKI1	AGOKI2
0	0	0.0	.237824	0.0	.237824
1	3	.535103	.178368	-.055103	.051794
2	0	0.0	.277279	.016503	-.023871
3	2	.356735	.178368	.123854	.112945
4	1	.178368	.237823	.178367	-.069456
5	1	.178368	.	.159599	.073800
6	1	.178368	.	.136977	.277131
7	1	.178368	.897280	.230995	.500483
8	3	.535103	.416191	.535103	.713471
9	3	.535103	.713470	1.044534	.888976
10	6	1.070205	.654014	1.636599	1.013925
11	2	.356735	1.248573	2.118374	1.090005
12	13	2.318779	1.129661	2.318778	1.129660
13	4	.713470	1.426941	2.175313	1.149367
14	7	1.248573	1.070205	1.767172	1.162095
15	7	1.248573	1.129661	1.275482	1.176799
16	5	.891838	1.486097	.891838	1.199118
17	13	2.318779	1.426941	.724844	1.190936
18	6	1.070205	1.545852	.754861	1.169629
19	7	1.248573	1.070205	.059411	1.112601
20	5	.891838	.713470	.891838	1.010749
21	0	0.0	.416191	.766736	.821414
22	2	.356735	.178368	.503813	.620657
23	1	.178368	.178368	.207798	.454423

A.14/21

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A.14/23

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<tr><td>5</td><td>.86</td><td>1.09</td><td>1.53</td><td>.</td><td>.</td><td>1.16</td><td>1.97</td></tr>
<tr><td>10</td><td>.86</td><td>4.29</td><td>4.93</td><td>.</td><td>.</td><td>3.36</td><td>2.46</td></tr>
<tr><td>20</td><td>3.40</td><td>.</td><td>.</td><td>.</td><td>.</td><td>3.40</td><td>3.25</td></tr>
<tr><td>25</td><td>3.64</td><td>.</td><td>.</td><td>.</td><td>.</td><td>3.64</td><td>3.56</td></tr>
<tr><td>30</td><td>2.27</td><td>2.57</td><td>4.93</td><td>5.38</td><td>.</td><td>3.73</td><td>3.81</td></tr>
<tr><td>60</td><td>1.13</td><td>2.27</td><td>3.64</td><td>6.03</td><td>6.86</td><td>3.99</td><td>4.05</td></tr>
<tr><td>80</td><td>1.13</td><td>1.21</td><td>6.07</td><td>.</td><td>.</td><td>2.80</td><td>3.04</td></tr>
<tr><td>90</td><td>1.01</td><td>1.13</td><td>3.64</td><td>.</td><td>.</td><td>1.92</td><td>2.17</td></tr>
<tr><td>100</td><td>.76</td><td>.86</td><td>1.13</td><td>1.21</td><td>2.27</td><td>1.44</td><td>1.00</td></tr>
<tr><td>.</td><td>2.43</td><td>.</td><td>.</td><td>.</td><td>.</td><td>.</td><td>.</td></tr>
</tbody>
</table>

1	2	MUTHAIGA	400	17782	598	16	3	6		1.5	-0.2	250
2	2	MUTHAIGA	400	17782	598	16	0	6		0.0	+0.4	250
3	2	MUTHAIGA	258	17782	598	16	5	1		1.0	-3.3	250
4	2	MUTHAIGA	231	17782	593	16	0	2		1.5	+1.5	250
5	2	MUTHAIGA	168	17782	598	16	0	0	312	4.0	-6.6	250
6	2	MUTHAIGA	142	17782	598	16	0	0	300	2.0	+3.0	250
7	2	MUTHAIGA	201	17782	578	16	0	2		1.0	-4.2	325
8	2	MUTHAIGA	398	17782	598	16	0	1		2.0	+6.6	325
9	2	UTALII	480	17782	598	16	0	7	600	4.0	+4.6	325
10	2	UTALII	309	17782	598	16	0	6	340	3.0	-5.1	325
11	2	UTALII	617	17782	598	16	2	6	3200	1.0	-0.6	325
12	2	UTALII	431	17782	598	16	5	5		1.0	+0.7	325
13	2	DRIVEIN	628	17782	598	16	3	5		1.0	-0.9	300
14	2	DRIVEIN	96	17782	598	16	0	0	700	2.0	+0.8	325
15	2	DRIVEIN	123	14609	464	15	0	0	2000	0.0	-0.3	300
16	2	DRIVEIN	1235	14609	464	14	3	4		1.5	+1.4	300
17	2	DRIVEIN	55	14609	464	14	0	0		0.5	-0.2	300
18	2	BREWERY	34	14609	464	14	0	2	110	2.0	+0.3	300
19	2	DRIVEIN	90	14609	465	14	0	2	1500	0.5	-1.0	300
20	2	BREWERY	105	14609	434	14	9	1	500	4.0	+1.2	300
21	2	DRIVEIN	214	14609	464	15	0	1		1.5	-1.1	300
22	2	BREWERY	587	14609	455	14	0	5		1.0	+6.0	400
23	2	ISU	129	14609	464	14	5	4	85	2.0	-1.1	300
24	2	BREWERY	145	14609	436	14	0	0	500	4.0	-6.0	400
25	2	BREWERY	84	14609	486	14	0	1	140	2.0	-2.6	300
26	2	BREWERY	138	14609	462	14	0	1		1.0	-6.0	400
27	2	BREWERY	57	14609	465	14	0	1	350	2.0	-1.6	300
28	2	BREWERY	131	14609	436	14	8	1	500	4.0	-6.0	300
29	2	BREWERY	555	14609	486	14	0	5		1.0	-7.0	400
30	2	BREWERY	579	12358	417	14	4	4		1.0	-0.7	300
31	2	BREWERY	398	14609	436	14	3	1	550	1.0	+5.5	300
32	2	BREWERY	541	12358	417	14	4	3		1.5	-2.0	325
33	2	BREWERY	343	12358	417	14	5	2		1.0	+5.0	300
34	2	ROYSAMBU	623	12358	417	14	0	2		1.0	+3.0	300
35	2	BREWERY	364	12358	417	14	0	2		1.0	+0.3	300
36	2	ROYSAMBU	75	10372	350	14	14	1	100	2.5	-1.4	325
37	2	BREWERY	350	12358	417	14	0	2	6000	1.5	-3.0	400
38	2	ROYSAMBU	594	10372	350	14	5	10		1.0	+2.0	325
39	2	SAFARIPK	748	12358	417	14	3	6		1.5	-0.7	400
40	2	ROYSAMBU	149	10372	350	14	0	1	450	4.5	+5.1	325
41	2	ROYSAMBU	286	12358	417	14	0	1	1250	1.5	-0.4	300
42	2	ROYSAMBU	326	10372	350	14	0	1		1.0	+2.2	325
43	2	ROYSAMBU	105	12358	417	14	0	2		1.5	-2.0	325
44	2	ROYSAMBU	1124	10372	350	14	3	7		0.5	+2.5	325
45	2	ROYSAMBU	93	12358	417	14	11	1	110	2.5	-0.4	325
46	2	ROYSAMBU	296	10372	350	14	0	2	620	4.0	+2.6	325
47	2	ROYSAMBU	119	12358	417	14	8	11	130	2.5	-1.2	325
48	2	ROYSAMBU	725	10372	350	14	1	8		0.5	+0.6	300
49	2	RUIRU	241	9592	254	26	4	2	5000	0.0	-2.5	275
50	2	RUIRU	124	9592	254	26	0	1	2500	3.0	+5.0	275
51	2	RUIRU	833	9592	313	26	0	3		0.0	0.4	275
52	2	RUIRU	965	9592	313	26	0	20		0.0	+2.3	275
53	2	RUIRU	507	9592	313	26	2	2	900	5.0	+1.6	275
54	2	RUIRU	189	9592	313	26	0	0	590	6.5	-0.6	275
55	2	RUIRU	148	9592	313	26	0	1		2.5	+6.0	275
56	2	RUIRU	319	9592	313	26	0	3	900	5.0	-2.4	275
57	2	RUIRU		9592	313	26	0	1	900		+1.1	275
58	2	RUIRU		9592	313	26	7	2		2.5	-7.	275

59	2	RUIRU		9592	312	26	1	2		2.0	-1.2	275
60	2	RUIRU	294	9592	313	26	4	2	900	3.0	-3.8	275
61	2	ROYSAMBU	102	10372	350	14	10	2	130	2.5	-1.2	325
62	2	GITHURAI	124	10045	340	14	0	2	2000	0.5	+2.8	300
63	2	ROYSAMBU	491	10372	350	14	0	2		2.0	-2.5	325
64	2	GITHURAI	218	10045	340	14	3	2		0.0	-2.0	300
65	2	ROYSAMBU	366	10372	350	14	0	1	900	4.5	-3.9	325
66	2	KAHAWA	4239	10045	254	14	2	14		0.5	+2.0	400
67	2	ROYSAMBU	1217	10372	350	14	0	8		1.5	+2.0	325
68	2	SUKARI	261	10045	254	14	0	4	900	4.0	-2.0	300
69	2	ROYSAMBU	288	10372	350	14	0	2	600	6.0	-2.0	325
70	2	SUKARI	803	10045	254	14	1	2		0.5	+0.4	300
71	2	ROYSAMBU	610	10372	350	14	2	7		1.5	-0.6	325
72	2	SUKARI	636	10045	254	14	0	1		1.5	-2.0	300
73	2	GITHURAI	170	10045	340	14	6	1	2000	1.5	-1.8	300
75	2	GITHURAI	606	10045	340	14	2	3	2	2.0	-5.0	300
77	2	KAHAWA	2037	10045	297	14	2	2		1.5	+1.0	275
79	2	SUKARI	132	10045	254	14	0	1	560	2.5	-1.0	275
81	2	SUKARI	54	10045	254	14	0	1		0.0	-2.0	275
82	2	SUKARI	161	10045	254	14	0	2	470	4.0	-4.0	300
83	2	SUKARI	298	10045	254	14	0	4		1.5	+0.9	300
84	2	RUIRU	722	9592	311	26	0	1		1.5	+1.2	275
85	2	RUIRU	449	9592	311	26	0	2	2100	2.0	-1.0	300
86	2	RUIRU	662	9592	313	26	1	3	910	1.5	-0.4	275
87	2	RUIRU	233	9592	311	26	0	1		1.5	-0.5	300
88	2	RUIRU		9592	311	20	0	1	1592	1.5	+2.2	300
89	2	RUIRU	326	9592	311	26	3	3	3000	2.5	-2.0	300
90	2	RUIRU	537	9592	311	26	0	2		1.0	-1.4	300
91	2	JUJA	470	7208	270	31	0	2		0.0	-5.0	300
92	2	JUJA	641	7208	270	31	0	4		0.5	+5.0	300
93	2	JUJA	638	7208	270	31	3	3	920	5.0	+1.8	300
94	2	JUJA	1295	7208	270	31	2	5		0.5	+1.7	300
95	2	JUJA	1159	7208	270	31	1	6		0.5	-0.5	300
96	2	JUJA	516	7208	244	31	0	1	929	4.5	-0.8	285
97	2	JUJA	609	7208	244	31	1	3		0.0	+2.5	285
98	2	JUJA	560	7208	244	31	0	4		0.0	-1.5	295
99	2	JUJA	65	7208	244	31	0	1	920	4.5	+0.3	285
100	2	JUJA	191	7208	244	31	0	1	700	2.0	-3.0	285
101	2	JUJA	1322	7208	244	31	2	19		0.5	-0.4	285
102	2	JUJA	1135	7208	244	31	3	22		0.5	+0.9	285
103	2	JUJA	1083	7208	240	31	1	2		1.0	-0.7	285
104	2	JUJA	709	7208	240	31	1	2		1.0	-0.7	285
105	2	JUJA	218	7208	240	31	0	1	1250	0.5	-3.4	225
106	2	JUJA	204	7208	240	31	0	2	1250	0.5	+3.7	225
107	2	JUJA	532	7208	240	31	0	4		1.0	-5.9	225
108	2	JUJA	781	7208	240	31	0	8		2.5	+4.5	225
109	2	JUJA	919	7208	240	31	1	2		1.0	+2.5	225
110	2	THIKA	1448	4824	210	3	2	1		1.5	+2.0	225
111	2	THIKA	1744	4824	210	36	2	2		1.0	-4.0	225
112	2	THIKA	313	4824	210	36	0	1	1500	3.0	+4.6	225
113	2	THIKA	339	4824	210	36	0	5	1625	3.0	+4.6	225
114	2	THIKA	1134	4824	210	36	1	5		1.0	+5.0	225
115	2	THIKA	1429	4824	210	36	1	9		1.5	+1.5	225
116	2	THIKA	355	4824	260	36	3	2	5000	1.5	-0.2	265
117	2	THIKA	1676	4824	260	36	2	10		1.0	0.0	265
118	2	THIKA	1857	4824	260	36	2	5		1.0	+0.2	265
119	2	BLUEPOST	123	4817	162	21	0	1	900	4.5	+1.5	300
120	2	BLUEPOST	247	4817	162	21	4	1		2.0	-5.0	300

A.15 (CONTD)

```
? $CAL %RE=(%YV-%FV)/%SQRT(%U(??
? $LOOK ONO 6 JPK LAB AR SUE %FV %RE P1?4
— list truncated
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	ONO	6	JPK	LAB	AR	SUE	%FV
1	1.000	2.34033	3.000	16.00	0.00000000	1.5000	4.5513
2	2.000	2.34033	0.000	16.00	0.00000000	0.0000	4.3771
3	3.000	1.80226	5.000	16.00	0.00000000	1.0000	3.7828
4	4.000	1.79189	0.000	16.00	0.00000000	1.5000	2.4132
5	5.000	1.47203	0.000	16.00	0.00000000	4.0000	1.7899
6	6.000	1.39459	0.000	16.00	0.00028313	2.0000	1.9767
7	7.000	1.65217	0.000	16.00	0.00000000	1.0000	2.5070
8	8.000	2.33532	0.000	16.00	0.00000000	2.0000	3.4815
9	9.000	2.52265	0.000	16.00	0.00000000	4.0000	5.4557
10	10.000	2.08221	0.000	16.00	0.00000000	3.0000	3.0535
11	11.000	2.77374	5.000	16.00	0.00000000	1.0000	6.3589
12	12.000	2.41498	5.000	16.00	0.00000000	1.0000	5.9109
13	13.000	2.79141	3.000	16.00	0.00000000	1.0000	5.8071
14	14.000	0.91322	0.000	16.00	0.00024378	2.0000	1.2908
15	15.000	0.96450	0.000	16.00	0.00000000	0.0000	1.6062
16	16.000	3.34951	3.000	16.00	0.00000000	1.5000	9.7502
17	17.000	0.15765	0.000	16.00	0.00000000	0.5000	0.7972
18	18.000	0.58314	0.000	16.00	0.00000000	3.0000	1.4911
19	19.000	0.65213	0.000	14.00	0.00000000	0.5000	1.1578
20	20.000	0.80628	9.000	14.00	0.00020000	1.0000	1.8174
21	21.000	1.90171	0.000	16.00	0.00000000	1.5000	2.5825
22	22.000	2.52734	0.000	14.00	0.00000000	1.0000	4.2773
23	23.000	1.44562	5.000	16.00	0.01117000	2.0000	5.6036
24	24.000	1.13593	0.000	14.00	0.00017000	4.0000	1.1663
25	25.000	0.58314	0.000	14.00	0.00000000	2.0000	1.1709
26	26.000	1.07957	0.000	14.00	0.00000000	1.0000	1.4375
27	27.000	0.19537	0.000	14.00	0.00000000	2.0000	0.7951
28	28.000	1.02752	8.000	14.00	0.00000000	1.0000	1.9487
29	29.000	2.68611	0.000	14.00	0.00000000	1.0000	4.3206
30	30.000	2.33587	4.000	14.00	0.00000000	1.0000	4.7859
31	31.000	2.13877	3.000	14.00	0.00000000	1.0000	4.3191
32	22.000	2.71958	4.000	14.00	0.00000000	1.5000	6.2549
33	33.000	1.82853	0.000	14.00	0.00000000	1.0000	3.9562
34	34.000	2.25408	0.000	14.00	0.00000000	1.0000	3.1817
35	35.000	1.67109	0.000	14.00	0.00000000	1.0000	2.6093
36	36.000	0.10025	14.000	14.00	0.00000000	2.5000	2.7260
37	37.000	1.78408	0.000	14.00	0.00031827	1.5000	2.0857
38	38.000	2.19667	5.000	14.00	0.00000000	1.0000	4.0363
39	39.000	2.60239	3.000	14.00	0.00000000	1.5000	5.3266
40	40.000	0.81374	0.000	14.00	0.00000000	4.5000	0.8911
41	41.000	1.64098	0.000	14.00	0.00000000	1.5000	2.1292
42	42.000	1.59669	0.000	14.00	0.00000000	1.0000	2.1281
43	43.000	0.64842	0.000	14.00	0.00000000	1.5000	0.9773
44	44.000	2.83441	3.000	14.00	0.00000000	0.5000	7.1590
45	45.000	0.46232	11.000	14.00	0.00000000	2.5000	3.1475
46	46.000	1.50015	0.000	14.00	0.00000000	4.0000	1.5207
47	47.000	0.76411	8.000	14.00	0.00000000	2.5000	2.8755
48	48.000	2.39596	1.000	14.00	0.00000000	0.5000	4.4844
49	49.000	1.21641	4.000	26.00	0.00000000	0.0000	3.1711
50	50.000	0.55189	0.000	26.00	0.00000000	0.0000	1.6058
51	51.000	2.45664	0.000	26.00	0.00000000	0.0000	5.9139
52	52.000	2.60374	0.000	26.00	0.00000000	0.0000	6.2060
53	53.000	1.96012	2.000	26.00	0.00011131	5.0000	2.7746
54	54.000	0.97336	0.000	26.00	0.00000000	2.5000	0.9861
55	55.000	0.72882	0.000	26.00	0.00010000	2.5000	1.1085
56	56.000	1.11811	0.000	26.00	0.00011111	5.0000	1.2583

57	57.000	1.36410	0.000	26.00	0.00000000	4.5000	1.5463
58	58.000	0.72204	7.000	26.00	0.00000000	2.5000	1.9962
59	59.000	2.38059	1.000	26.00	0.00000000	2.0000	4.7148
60	60.000	1.38058	4.000	26.00	0.00000000	5.0000	2.0808
61	61.000	0.43476	10.000	14.00	0.00000000	2.5000	2.6046
62	62.000	0.59804	0.000	14.00	0.00000000	0.5000	1.0992
63	63.000	2.00624	0.000	14.00	0.00000000	2.0000	2.5569
64	64.000	1.16225	5.000	14.00	0.00000000	0.0000	2.5051
65	65.000	1.71242	0.000	14.00	0.00000000	4.5000	1.5325
66	66.000	4.12960	2.000	14.00	0.00000000	0.5000	17.6224
67	67.000	2.99442	0.000	14.00	0.00000000	1.5000	5.7289
68	68.000	1.34228	0.000	14.00	0.00000000	4.0000	1.3135
69	69.000	1.47275	7.000	14.00	0.00000000	6.0000	1.1794
70	70.000	2.46611	1.000	14.00	0.00000000	0.5000	4.4749
71	71.000	2.22325	2.000	14.00	0.00000000	1.5000	3.7151
72	72.000	2.23298	0.000	14.00	0.00000000	1.5000	3.2251
73	73.000	0.91355	4.000	14.00	0.00000000	1.5000	1.9293
74	75.000	2.18464	3.000	14.00	0.00000000	2.0000	3.8565
75	77.000	3.61649	3.000	14.00	0.00000000	1.5000	11.4265
76	79.000	0.98176	0.000	14.00	0.00000000	2.5000	1.2887
77	81.000	-0.06336	0.000	14.00	0.00000000	0.0000	0.6652
78	82.000	0.85916	0.000	14.00	0.00000000	4.0000	0.9767
79	83.000	1.47485	0.000	14.00	0.00000000	1.5000	1.6221
80	84.000	2.31364	0.000	26.00	0.00000000	1.5000	4.4285
81	85.000	1.83843	0.000	26.00	0.00000000	1.5000	2.9771
82	86.000	2.23590	1.000	26.00	0.00000000	1.5000	4.8425
83	87.000	1.37706	0.000	26.00	0.00000000	1.5000	2.1853
84	88.000	0.00000	0.000	26.00	0.00000000	1.5000	0.7114
85	89.000	1.51851	3.000	26.00	0.00000000	2.5000	2.7540
86	90.000	2.10664	0.000	26.00	0.00000000	1.0000	4.0242
87	91.000	1.59861	0.000	31.00	0.00000000	0.0000	3.4472
88	92.000	1.90890	0.000	31.00	0.00000000	0.5000	4.0936
89	93.000	1.90421	3.000	31.00	0.00000000	5.0000	3.1827
90	94.000	2.61445	2.000	31.00	0.00000000	0.5000	8.0911
91	95.000	2.50119	1.000	31.00	0.00000000	0.5000	6.8920
92	96.000	1.69198	0.000	31.00	0.00000000	4.5000	2.3041
93	97.000	1.85769	1.000	31.00	0.00000000	0.0000	4.5135
94	98.000	1.93311	0.000	31.00	0.00000000	0.0000	4.4516
95	99.000	-0.33462	0.000	31.00	0.00000000	4.5000	0.5011
96	100.000	0.69815	0.000	31.00	0.00000000	2.0000	1.5115
97	101.000	2.63277	2.000	31.00	0.00000000	0.5000	8.2036
98	102.000	2.48026	3.000	31.00	0.00000000	0.5000	7.8760
99	103.000	2.43336	1.000	31.00	0.00000000	1.0000	6.1880
100	104.000	2.00973	1.000	31.00	0.00000000	1.0000	4.4331
101	105.000	0.83037	0.000	31.00	0.00000000	0.5000	1.9198
102	106.000	0.76399	0.000	31.00	0.00000000	0.5000	1.8282
103	107.000	1.72252	0.000	31.00	0.00000000	1.0000	3.3522
104	108.000	2.10645	0.000	31.00	0.00000000	2.5000	3.7818
105	109.000	2.26916	1.000	31.00	0.00000000	1.5000	5.4505
106	110.000	2.32222	2.000	30.00	0.00000000	1.5000	5.6299
107	111.000	2.50822	2.000	36.00	0.00000000	1.0000	7.8204
108	112.000	0.79049	0.000	36.00	0.00000000	3.0000	1.5169
109	113.000	0.87028	0.000	36.00	0.00000000	3.0000	1.5053
110	114.000	2.07779	1.000	36.00	0.00000000	1.0000	5.2477
111	115.000	2.30901	1.000	36.00	0.00000000	1.5000	5.6808
112	116.000	0.91640	3.000	36.00	0.00000000	1.0000	2.4831
113	117.000	2.48149	2.000	36.00	0.00000000	1.0000	7.6645
114	118.000	2.57100	2.000	36.00	0.00000000	1.0000	8.1990
115	119.000	1.09020	0.000	21.00	0.00000000	4.5000	1.1860
116	120.000	0.55222	4.000	21.00	0.00000000	2.0000	1.3961

? %LOOK	ONO	PIA	%FV	%RE	RAD	AF%
	ONO	PIA	%FV	%RE	R&D	AF
1	1.000	8.000	4.5813	1.610095		0.0000000
2	2.000	6.000	4.3771	0.775710		0.0000000
3	3.000	1.000	3.9020	-1.474617		0.0000000
4	4.000	3.000	2.4132	0.377714	112.000	0.0032051
5	5.000	0.000	1.7399	-1.319071	370.000	0.0033333
6	6.000	0.000	1.9787	-1.403818		0.0000000
7	7.000	2.000	2.3090	-0.203371		0.0000000
8	8.000	1.000	3.4215	-1.309108	600.000	0.0016667
9	9.000	7.000	3.4587	1.904148	340.000	0.0029412
10	10.000	6.000	3.0055	1.686172	320.000	0.0003125
11	11.000	6.000	6.3633	-0.145953	280.000	0.0000000
12	12.000	5.000	5.9489	-0.309042		0.0000000
13	13.000	5.000	6.8071	-0.692635		0.0000000
14	14.000	0.000	1.2903	-1.135123	700.000	0.0014286
15	15.000	0.000	1.5042	-1.267949	2000.000	0.0005000
16	16.000	6.000	9.7502	-1.201023		0.0000000
17	17.000	0.000	0.7972	-0.392867		0.0000000
18	18.000	2.000	1.4941	0.413900	110.000	0.0090909
19	19.000	2.000	1.1578	0.752679	1500.000	0.0002667
20	20.000	1.000	1.8174	0.6063002		0.0000000
21	21.000	1.000	2.6265	-1.601916		0.0000000
22	22.000	1.000	4.0773	0.344434		0.0000000
23	23.000	4.000	5.6048	-0.479311		0.0117647
24	24.000	0.000	1.1483	-1.079950	60.000	0.0016667
25	25.000	1.000	1.4175	-0.350607	140.000	0.0071429
26	26.000	2.000	3.4917	-0.934426		0.0000000
27	27.000	1.000	1.4375	-0.364926		0.0000000
28	28.000	1.000	0.7931	-0.577172	350.000	0.0028571
29	29.000	1.000	1.9467	-0.574621	600.000	0.0016667
30	30.000	5.000	4.8204	0.081703		0.0000000
31	31.000	1.000	4.9857	1.347863	850.000	0.0011755
32	32.000	6.000	4.3191	-1.577057		0.0000000
33	33.000	2.000	6.2649	-0.1058442	1250.000	0.0021739
34	34.000	1.000	3.9482	-0.930426		0.0000000
35	35.000	2.000	2.6043	-0.577172		0.0000000
36	36.000	1.000	2.7260	-1.045386	130.000	0.0076923
37	37.000	1.000	2.3259	-0.213717	6500.000	0.0001657
38	38.000	13.000	4.8563	3.712216		0.0000000
39	39.000	6.000	5.0240	0.292711	400.000	0.0021739
40	40.000	1.000	0.8411	1.153647	1250.000	0.0008000
41	41.000	1.000	2.1793	-0.798857		0.0000000
42	42.000	2.000	2.1221	-0.710272		0.0000000
43	43.000	7.000	0.9778	1.033700		0.0000000
44	44.000	0.000	7.1590	-0.059423	110.000	0.0090909
45	45.000	1.000	3.1475	-1.216471	650.000	0.0016129
46	46.000	2.000	1.5289	0.581017	130.000	0.0076923
47	47.000	11.000	2.4765	4.789709	180.000	0.0000000
48	48.000	8.000	4.4344	1.693253		0.0000000
49	49.000	2.000	3.1711	-0.657634	5000.000	0.0002000
50	50.000	1.000	1.0055	-0.798857	2500.000	0.0000000
51	51.000	3.000	5.9133	-1.198024		0.0000000
52	52.000	20.000	6.6060	5.211222	110.000	0.0090909
53	53.000	2.000	2.7746	-0.465012	650.000	0.0016129
54	54.000	0.000	0.9061	-0.943019	300.000	0.0016949
55	55.000	1.000	1.1089	-0.172990		0.0000000
56	56.000	3.000	1.2453	1.507900	400.000	0.0011111
57	57.000	1.000	1.6443	-0.543730	900.000	0.0011111

58	58.000	2.000	1.9902	0.006964	0.00	0.0000000
59	59.000	2.000	4.7198	-1.251766	0.00	0.0000000
60	60.000	2.000	2.5898	-0.056082	900.00	0.0011111
61	61.000	8.000	2.6046	-0.374626	130.00	0.0076923
62	62.000	2.000	1.0272	0.859804	2000.00	0.0005000
63	63.000	2.000	2.5589	0.347376	0.00	0.0000000
64	64.000	2.000	2.5051	-0.319186	0.00	0.0000000
65	65.000	1.000	1.6323	-0.494892	900.00	0.0011111
66	66.000	14.000	17.5274	-6.863970	0.00	0.0000000
67	67.000	8.000	5.7027	0.931872	0.00	0.0000000
68	68.000	4.000	1.3124	2.346075	900.00	0.0011111
69	69.000	2.000	1.1779	0.755640	600.00	0.0016667
70	70.000	2.000	4.6747	-1.237157	0.00	0.0000000
71	71.000	7.000	3.7151	1.704254	0.00	0.0000000
72	72.000	1.000	3.2251	-1.237020	0.00	0.0000000
73	73.000	1.000	1.5273	-0.667071	2000.00	0.0005000
74	75.000	8.000	3.6586	2.267762	0.00	0.0000000
75	77.000	8.000	11.4835	-1.010165	0.00	0.0000000
76	79.000	1.000	1.2397	-0.254482	460.00	0.0021739
77	81.000	1.000	0.6962	0.378037	0.00	0.0000000
78	82.000	2.000	0.9767	1.037002	470.00	0.0021277
79	83.000	4.000	1.8221	1.613410	0.00	0.0000000
80	84.000	1.000	4.4764	-1.628597	0.00	0.0000000
81	85.000	2.000	2.9771	-0.956274	3100.00	0.0003226
82	86.000	3.000	1.5425	-0.837255	910.00	0.0010989
83	87.000	1.000	2.1063	-0.802226	0.00	0.0000000
84	88.000	1.000	0.7114	0.342177	1592.00	0.0006281
85	89.000	3.000	2.7540	0.149256	3000.00	0.0003333
86	90.000	2.000	0.0242	-1.007043	0.00	0.0000000
87	91.000	2.000	2.1472	-0.772970	0.00	0.0000000
88	92.000	4.000	4.0906	-0.049602	0.00	0.0000000
89	93.000	3.000	2.1827	-0.102426	920.00	0.0010870
90	94.000	5.000	3.0911	-1.036711	0.00	0.0000000
91	95.000	6.000	8.8970	-0.341542	0.00	0.0000000
92	96.000	1.000	2.3941	-0.857143	929.00	0.0010764
93	97.000	3.000	4.3135	-0.713401	0.00	0.0000000
94	98.000	4.000	4.4516	-0.214050	0.00	0.0000000
95	99.000	1.000	6.5011	0.704757	920.00	0.0010870
96	100.000	1.000	1.2115	-0.416021	700.00	0.0014286
97	101.000	19.000	11.2036	3.767461	0.00	0.0000000
98	102.000	22.000	7.8780	5.031382	0.00	0.0000000
99	103.000	2.000	6.1680	-1.273246	0.00	0.0000000
100	104.000	2.000	4.4931	-1.172745	0.00	0.0000000
101	105.000	1.000	1.9198	-0.663843	1250.00	0.0003000
102	106.000	2.000	1.3262	0.128621	1250.00	0.0008000
103	107.000	4.000	1.3502	0.353800	0.00	0.0000000
104	108.000	8.000	3.7318	2.209179	0.00	0.0000000
105	109.000	2.000	5.4505	-1.477958	0.00	0.0000000
106	110.000	1.000	5.6297	-1.751283	0.00	0.0000000
107	111.000	2.000	7.7204	-2.081313	0.00	0.0000000
108	112.000	1.000	1.5169	-0.419652	1500.00	0.0006667
109	113.000	5.000	1.6053	2.673374	1625.00	0.0006154
110	114.000	5.000	5.2497	-0.107001	0.00	0.0000000
111	115.000	9.000	5.8808	1.286247	0.00	0.0000000
112	116.000	2.000	2.4731	-0.271824	5000.00	0.0002000
113	117.000	10.000	7.6645	0.943396	0.00	0.0000000
114	118.000	5.000	8.1790	-1.117218	0.00	0.0000000
115	119.000	1.000	1.1860	-0.170804	900.00	0.0011111
116	120.000	1.000	1.3381	-0.272250	0.00	0.0000000

Note for Appendix A.15

1: Constant,
S: power to which travel is raised,
JPK: junctions per kilometre,
LAB: lorries and buses
FCR(2): horizontal radius (800-3200m),
FCR(3): horizontal radius (73200m),
FCR(1): SUE: horizontal radius (0-800m) and superelevation,
FCR(2). SUE: horizontal radius (800-3200m) and superelevation,
FCR(3). SUE: horizontal radius (>3200m) and superelevation,
AR: reciprocal of the horizontal radius,
SUE: superelevation

```
$UNITS 116
$DATA ONO LES ADT VEH LAI JIT F A RAD SUE GR
$FORMAT F3.0,12X,F4.0,1X,F5.0,1X,F9.0,1X,F2      1X,F2.0,
1X,F4.0,1X,F3.1,1X,I4.1,1X,F1.0 *
$C AGOGIN 4AJOL DATA, nJ(n/IRTn GUGMN2, DnI/   nn111
$C MLR 15.3.1988
$DIM 7$
$CAL A1=365*(LES/1.06)*(D1, 1     n n
$CAL SUO=4*A1$
$CAL S=%LOG(SUO)$
$CAL B1=%IF(RAD>0,RAD,1)$
$CAL B2=1/B1$
$CAL AR=(B2/=1)*B2
$CAL ARA=10+GRA$
$FAC FJ 3$
$CAL FJ=1+%GE(JPF,3)+%GE(JPF,7)$
$CAL CR=(RAD-1)*10000+RAD$
$FAC FCR 3 FSU 2 FGR 3 FLB 3$
$CAL FCR=1+%GE(CR,800)+%GE(CR,3800)$
$CAL FSU=1+%GE(SUE,3.5)$
$CAL FGR=1+%GE(ARA,7)+%GE(ARA,15)$
$CAL FLB=1+%GE(LAI,20)+%GE(LAI,30)$
$YVAR PIA$ERROR P$LINK L$
$FIT :+S:+JPK::+LAI:+AR:+SUE+FGR D E M
$FIT S+JPK+FCR+FCR.SUE$DISP E M$
$C CAL %RE=(%YV-%FV)/%SQRT(SU)$
$C LOOK ONO PIA %FV %RE FJ FLB FCC FLD FGR$
$RETURN $
```

```
$FIT :+S:+LAB:+JPK$DISP D E$
scaled deviance = 300.19 at cycle  4
            d.f. = 115

scaled deviance = 228.96 (change =  -151.2) at cycle  4
          · d.f. = 114    (change =    -1  )

scaled deviance = 221.88 (change =   -7.07) at cycle  4
            d.f. = 113    (change =    -1  )

scaled deviance = 206.23 (change =  -15.650) at cycle  4
            d.f. = 112    (change =    -1  )

Scaled deviance is  206.23  on  112 d.f.   from 116 observations
            change is  -15.65 for   -1 d.f.

            estimate       s.e.       parameter
    1      -0.8228        0.2232        1
    2       0.7636        0.06079       S
    3       0.02125       0.006199      LAB
    4       0.09371       0.02255       JPK
    scale parameter taken as  1.000
```

```
? $FIT +SUE$
scaled deviance = 201.02 (change =   -5.21) at cycle  4
            d.f. = 111    (change =   -1  )
```

```
? $DISP D E$
Scaled deviance is  201.020  on  111 d.f.   from 116 observations
            change is  -5.21 for   -1 d.f.

            estimate       s.e.       parameter
    1      -0.4976        0.2612        1
    2       0.7019        0.05597       S
    3       0.01941       0.006736      LAB
    4       0.09345       0.02269       JPK
    5      -0.1021        0.04583       SUE
    scale parameter taken as  1.000
```

```
? $FIT +SID$
scaled deviance = 200.63 (change =    -0.39) at cycle  4)
              d.f. = 110   (change =   -1   )

? $FIT -SID+AR$
scaled deviance = 196.91 (change =    -3.73) at cycle  4
              d.f. = 110   (change =    0   )

? $DISP D E$
Scaled deviance is  196.908  on  110 d.f.   from  11. observations
              change is   -3.726 for   0 d.f.

        estimate      s.e.      parameter
    1    -0.6274      0.2716     1
    2     0.7532      0.07081    S
    3     0.02132     0.006333   LAB
    4     0.07437     0.02435    JPK
    5    -0.1213      0.04734    SUE
    6    67.36       31.50       AR
    scale parameter taken as  1.000

? $FIT -AR+FCR$
scaled deviance = 198.26 (change =    +1.35) at cycle  4
              d.f. = 109   (change =   -1   )

? $DISP D E$
Scaled deviance is  198.258  on  109 d.f.   from  11. observations
              change is   +1.350 for  -1 d.f.

        estimate      s.e.      parameter
    1    -0.2777      0.2984     1
    2     0.7330      0.07412    S
    3     0.02174     0.006450   LAB
    4     0.08607     0.02267    JPK
    5    -0.1278      0.05698    SUE
    6    -0.3608      0.2196     FCR(2)
    7    -0.3133      0.2305     FCR(3)
    scale parameter taken as  1.000

? $FIT +GRA$
scaled deviance = 197.02 (change =    -1.24) at cycle  4
              d.f. = 108   (change =   -1   )

? $FIT :+S:+JPK:+LAB:+SUE:+AR$DISP D E$
scaled deviance = 380.19 at cycle  4
              d.f. = 115

scaled deviance = 228.96 (change =   -151.2) at cycle  4
              d.f. = 114   (change =   -1   )

scaled deviance = 217.69 (change =  -11.271) at cycle  4
              d.f. = 113   (change =  -1    )

scaled deviance = 206.23 (change =  -11.452) at cycle  4
              d.f. = 112   (change =  -1    )

scaled deviance = 201.02 (change =    -5.21) at cycle  4
              d.f. = 111   (change =   -1   )

scaled deviance = 196.91 (change =    -4.11) at cycle  4
              d.f. = 110   (change =   -1   )

Scaled deviance is  196.908  on  110 d.f.   from  11. observations
              change is   -4.112 for  -1 d.f.

        estimate      s.e.      parameter
    1    -0.6274      0.2716     1
    2     0.7532      0.07081    S
    3     0.07437     0.02435    JPK
    4     0.02132     0.006333   LAB
    5    -0.1213      0.04734    SUE
    6    67.36       31.50       AR
    scale parameter taken as  1.000

? $FIT -SUE+GRA$
scaled deviance = 202.78 (change =    +5.87) at cycle  4
              d.f. = 110   (change =    0   )

? $FIT S+JPK+LAB:+AR$
scaled deviance = 206.23 at cycle  4
              d.f. = 112

scaled deviance = 203.88 (change =    -2.36) at cycle  4
              d.f. = 111   (change =  -1   )

? $DISP D E$
Scaled deviance is  203.878  on  111 d.f.   from  11. observations
              change is   -2.356 for  -1 d.f.

        estimate      s.e.      parameter
    1    -0.4721      0.2443     1
    2     0.8113      0.05629    S
    3     0.08006     0.02416    JPK
    4     0.02308     0.006327   LAB
    5    51.36       32.18       AR
    scale parameter taken as  1.000
```

```
? $FIT +AR+FCR:+FCR.SUE$
scaled deviance = 203.39 (change =   -0.40) at cycle
           d.f. = 110    (change =   -1   )

scaled deviance = 194.53 (change =   -8.85) at cycle
           d.f. = 107    (change =   -3   )

? $DISP D E$
Scaled deviance is  194.531  on  107 d.f.   from  117  observations
           change is   -0.069 for   -3 d.f.

          estimate       s.e.      parameter
   1      -0.09198      0.4942      1
   2       0.7248       0.07462     S
   3       0.08490      0.02295     JPK
   4       0.01904      0.006628    LAB
   5      -0.7595       0.5382      FCR(2)
   6      -0.2860       0.4984      FCR(3)
   7      -0.1692       0.1371      FCR(1).SUE
   8      -0.01629      0.08199     FCR(2).SUE
   9      -0.2639       0.09962     FCR(3).SUE
   scale parameter taken as  1.000

? $FIT -FCR.SUE+FCR.FSU$
scaled deviance = 199.89 (change =   +5.36) at cycle
           d.f. = 108    (change =   +1   )

? $FIT +FSU$
scaled deviance = 199.89 (change =   0.00) at cycle
           d.f. = 108    (change =   0   )

? $FIT :+S:+JPK:+LAB:+SID$
scaled deviance = 380.19 at cycle  4
           d.f. = 115

scaled deviance = 228.96 (change =  -151.2) at cycle  4
           d.f. = 114    (change =   -1   )

scaled deviance = 217.69 (change = -11.271) at cycle  4
           d.f. = 113    (change =   -1   )

scaled deviance = 206.23 (change = -11.452) at cycle  4
           d.f. = 112    (change =   -1   )

scaled deviance = 205.85 (change =   -0.38) at cycle  4
           d.f. = 111    (change =   -1   )

? $FIT -LAB$
scaled deviance = 215.37 (change =   +9.51) at cycle
           d.f. = 112    (change =   +1   )

? $FIT :+S:+JPK:+LAB:+AR$DISP D E$FIT +SUE$
scaled deviance = 380.19 at cycle  4
           d.f. = 115

scaled deviance = 228.96 (change =  -151.2) at cycle  4
           d.f. = 114    (change =   -1   )

scaled deviance = 217.69 (change = -11.271) at cycle  4
           d.f. = 113    (change =   -1   )

scaled deviance = 206.23 (change = -11.452) at cycle  4
           d.f. = 112    (change =   -1   )

scaled deviance = 203.88 (change =   -2.36) at cycle  4
           d.f. = 111    (change =   -1   )

Scaled deviance is  203.878  on  111 d.f.   from  115 observations
           change is   -2.356 for   -1 d.f.

          estimate       s.e.      parameter
   1      -0.9721       0.2443      1
   2       0.8113       0.06829     S
   3       0.08006      0.02416     JPK
   4       0.02308      0.006327    LAB
   5      51.36         32.18       AR
   scale parameter taken as  1.000

scaled deviance = 196.91 (change =   -6.97) at cycle
           d.f. = 110    (change =   -1   )

? $DISP D E$
Scaled deviance is  196.903  on  110 d.f.   from  110 observations
           change is   -6.970 for   -1 d.f.

          estimate       s.e.      parameter
   1      -0.6274       0.2716      1
   2       0.7532       0.07081     S
   3       0.07437      0.02435     JPK
   4       0.02132      0.006393    LAB
   5      67.36         31.50       AR
   6      -0.1213       0.04734     SUE
   scale parameter taken as  1.000
```

1	2	KISANJO	500	1023	43	38	4	3	609.6	+2.60	1.5	35.0	1.25	5.00
2	2	KIGANJO	500	1673	11	21	6	1		-0.50	0.0	45.0	2.50	3.50
3	2	KIGANJO	500	1673	71	27	5	5	914.4	-1.10	5.0	50.0	1.25	0.00
4	2	KIGANJO	500	1229	51	28	2	0	609.6	-1.40	5.0	7.5	0.00	1.25
5	2	KIGANJO	500	1229	51	28	2	0	914.4	+0.57	2.5	60.0	0.00	1.25
6	2	KIGANJO	500	1236	54	27	4	1		+1.30	5.0	35.0	0.00	0.00
7	2	KIGANJO	500	1236	54	27	4	8		-0.10	0.0	60.0	0.00	1.25
8	2	CHAKA	500	1509	53	32	0	1	609.6	-0.30	5.0	30.0	1.25	2.50
9	2	CHAKA	500	1509	63	36	6	1		-0.50	6.2	50.0	0.00	0.00
10	2	CHAKA	500	1375	57	29	2	0	1524.0	+1.50	7.3	20.0	0.00	1.25
11	2	KILIMANI	500	1375	57	29	4	5		+3.50	22.5	50.0	0.00	0.00
12	2	KILIMANI	500	1375	57	29	3	2		+5.00	7.3	30.0	1.25	2.50
13	2	KILIMANI	500	1301	56	36	3	2	609.6	+1.30	35.0	30.0	5.00	13.50
14	2	KILIMANI	500	1776	77	38	4	0	1036.3	+2.98	15.0	30.0	0.00	10.00
15	2	KILIMANI	500	1002	47	51	0	2		+2.70	12.5	35.0	10.00	1.25
16	2	KARICHEN	500	1063	44	30	0	0		-0.20	17.5	35.0	0.00	1.25
17	2	KARICHEN	500	1063	44	30	0	1		+1.10	25.0	5.0	5.75	3.75
18	2	KARICHEN	500	1063	44	30	0	0		+1.30	20.0	15.0	1.25	1.25
19	2	KARICHEN	500	1045	44	30	4	1		+1.30	17.5	50.0	2.50	50.00
20	2	KARICHEN	500	1045	44	30	4	4		+1.10	25.0	40.0	0.00	3.12
21	2	NYANGE	500	1065	44	30	6	0	914.4	-2.50	20.0	50.0	1.38	2.00
22	2	NYANGE	500	987	41	30	4	2	914.4	+3.50	10.0	50.0	0.00	1.25
23	2	NYANGE	500	987	41	30	6	2	502.9	+1.00	12.5	50.0	0.00	13.75
24	2	NYANGE	500	987	41	30	0	2	1524.0	+1.40	15.0	50.0	5.00	10.00
25	2	COISOI	500	987	41	30	0	0	1524.0	+2.50	12.5	20.0	11.25	1.25
26	2	COISOI	500	987	41	30	0	0		-1.00	20.0	50.0	0.00	23.75
27	2	COISOI	500	987	41	30	0	2	6096.0	-1.40	7.5	70.0	6.25	13.75
28	2	COISOI	500	937	41	30	0	1		+0.17	20.0	55.0	0.00	1.25
29	2	COISOI	500	987	41	30	0	0		+0.60	10.0	80.0	0.00	2.50
30	2	COISOI	500	987	41	30	0	0		+1.20	12.5	60.0	0.00	11.25
31	2	THUNGARE	500	987	41	30	4	0		+0.10	10.0	50.0	0.00	7.50
32	2	THUNGARE	500	987	41	30	4	0		+1.40	12.5	40.0	0.00	5.00
33	2	THUNGARE	500	987	41	30	4	0		+0.50	12.0	50.0	0.00	2.50
34	2	THUNGARE	500	987	41	30	4	0		+0.31	15.0	55.0	0.00	6.25
35	2	THUNGARE	500	987	41	30	0	1		+0.30	12.5	50.0	0.00	12.50
36	2	THUNGARE	500	1141	48	25	4	1	914.4	-1.95	17.5	50.0	2.50	3.75
37	2	THUNGARE	500	1141	45	25	0	0		+1.70	10.0	25.0	1.25	3.75
38	2	THUNGARE	500	1141	43	25	2	0	914.4	+0.90	17.5	30.0	3.75	8.75
39	2	THUNGARE	500	1141	48	25	0	0		-0.03	7.5	45.0	0.00	26.25
40	2	THUNGARE	500	1141	48	25	4	0		-0.80	17.5	60.0	0.00	5.00
41	2	AGUTHI	500	1141	48	25	2	3		-0.07	12.5	45.0	1.25	2.50
42	2	AGUTHI	500	1141	48	25	0	0		+0.60	22.5	95.0	0.00	1.25
43	2	AGUTHI	500	1141	48	25	0	0		+2.46	22.5	90.0	0.00	0.00
44	2	AGUTHI	500	1141	46	25	0	0		-1.15	6.2	65.0	2.50	12.50
45	2	AGUTHI	500	1141	48	25	2	1		+0.14	20.0	50.0	3.75	7.50
46	2	RUARE	500	1141	48	25	0	1		+2.24	7.5	60.0	2.50	2.50
47	2	RUARE	500	1141	48	25	0	0		+0.90	12.5	35.0	2.50	5.00
49	2	RUARE	500	1141	48	25	0	1		-1.20	12.5	55.0	5.00	15.00

9	2	RUARE	500	1141	48	25	0	4		-0.20	37.5	35.0	0.00	11.25
0	2	RUARE	500	1141	48	25	8	1		+5.60	10.0	90.0	3.73	2.50
1	2	NAROMORU	500	1141	48	25	8	1	304.8	+4.80	37.5	25.0	0.00	1.25
2	2	NAROMORU	500	1050	44	28	6	1	304.8	-2.70	10.0	70.0	0.00	1.25
3	2	NAROMORU	500	1050	44	28	4	1		+1.06	8.7	35.0	12.50	8.75
4	2	NAROMORU	500	1050	44	28	4	1	609.6	+3.10	25.0	15.0	19.75	12.50
5	2	NAROMORU	500	1050	44	28	2	3	3048.0	-0.55	10.0	72.5	6.25	6.25
6	2	NAROMORU	500	1050	44	28	2	2	3048.0	+2.70	12.5	50.0	10.00	11.25
7	2	MURIRO	500	1050	44	28	0	0	3048.0	-0.67	10.0	35.0	7.50	5.00
8	2	MURIRO	500	1050	44	28	4	1	3048.0	-2.37	20.0	40.0	6.25	2.50
9	2	MURIRO	500	1050	44	28	6	2	3048.0	-3.50	32.5	70.0	10.00	12.50
0	2	MURIRO	500	1050	44	28	4	2	3048.0	+1.29	32.5	35.0	6.25	12.50
1	2	MURIRO	500	1050	44	28	2	1	3048.0	-3.80	20.0	45.0	1.25	6.25
2	2	MURIRO	500	1050	44	28	4	0	3048.0	-4.48	23.7	60.0	3.75	6.25
3	2	MURIRO	500	1050	44	28	8	0		+4.40	22.5	35.0	5.50	5.00
4	2	BANTU	500	1050	44	28	2	0	457.2	-0.29	32.5	45.3	1.25	2.50
5	2	BANTU	500	1050	44	28	2	0	304.8	-2.02	15.0	90.0	1.25	5.00
6	2	BANTU	500	1050	44	28	2	0		-0.44	5.0	45.0	5.00	0.00
7	2	BANTU	500	1050	44	28	4	1		-3.03	40.0	60.0	5.00	5.00
8	2	BANTU	500	1050	44	28	2	1		-2.52	22.5	45.0	7.50	7.50
9	2	BANTU	500	1050	44	28	0	0		+0.15	5.0	75.0	7.50	5.00
0	2	BANTU	500	1202	50	28	6	0	1524.0	+2.84	52.5	25.0	5.00	6.25
1	2	BANTU	500	1202	50	28	2	0		-1.59	25.0	47.5	3.75	7.50
2	2	BANTU	500	1202	50	28	4	1	1524.0	+0.33	15.0	65.0	7.50	7.50
3	2	GWARA	500	1202	50	28	8	0	1524.0	+0.48	17.5	15.0	10.00	3.75
4	2	GWARA	500	1202	50	28	6	0		+0.67	20.0	25.0	5.00	6.37
5	2	GWARA	500	1202	50	28	0	0		+1.78	10.0	30.0	16.25	12.50
6	2	GWARA	300	1202	50	28	2	0		+0.39	47.5	60.0	17.50	16.25
7	2	GWARA	500	1202	50	28	0	2	3048.0	-4.03	37.5	35.0	7.50	10.00
8	2	GWARA	500	1202	50	28	2	0	3048.0	+6.53	15.0	65.0	12.50	5.75
9	2	GWARA	500	1202	50	28	2	0	3048.0	-1.99	27.5	62.5	10.00	3.75
0	2	GWARA	500	1202	50	28	6	0	3048.0	+1.40	15.0	70.0	2.50	2.50
1	2	ICUGA	500	1202	50	28	8	2		+0.60	12.5	17.5	10.00	5.00
2	2	ICUGA	500	1202	50	28	2	0	3048.0	+1.25	10.0	80.0	1.25	1.25
3	2	ICUGA	500	1202	50	28	0	0		-0.15	37.5	80.0	3.12	10.00
4	2	ICUGA	500	1202	50	28	4	0	3048.0	-0.40	30.0	75.0	3.12	10.00
5	2	GATHERI	500	1202	50	28	2	1		+1.55	27.5	20.0	15.00	2.50
6	2	GATHERI	500	1202	50	28	4	1		+4.30	71.2	50.0	7.25	3.75
7	2	GATHERI	500	1202	50	28	4	2		+4.10	62.5	45.0	3.75	2.50
8	2	MOIEQUAT	500	1202	50	28	8	3		+1.20	45.0	40.0	15.62	5.00
9	2	MOIEQUAT	500	1434	60	25	12	0		+0.98	32.5	97.5	2.50	2.50
0	2	MOIEQUAT	500	1434	60	25	6	1		+5.36	25.0	50.0	0.00	2.50
	2	SILVERBE	500	1434	60	25	6	1		+5.00	12.5	65.0	1.25	0.00
2	2	SILVERBE	500	1573	65	25	4	3	914.4	+0.99	12.5	45.0	0.00	1.25
3	2	NANYUKI	500	1573	65	25	6	2		-0.86	12.5	20.0	2.50	8.75
4	2	NANYUKI	500	1573	65	25	2	1		-0.08	12.5	0.00	5.00	7.50
5	2	NANYUKI	500	2363	98	19	0	0		+0.57	5.0	0.00	2.50	5.00
6	2	NANYUKI	500	2363	98	19	0	3		+0.49	5.0	0.00	5.00	2.50

model ... units is 95

y-variate PIA
weight *
offset *

probability distribution is POISSON
link function is LOGARITHM
scale parameter is 1.000

terms = 1 + S + FJ + FLB + FCC + FED + FCR

— list truncated

```
GLIM SESSION ENDS
C:\GLIM $UNITS 96
$DATA ONO LES ADT VPH LAB JFK PIA RAD LGR RUT EDS CRC FEA$
$FORMAT F3.0,12X,F3.0,1X,F4.0,1X,F3.0,1X,F2.0,1X,F2.0,1X,F2.0,
1X,F6.1,1X,F5.2,1X,F4.1,1X,F4.1,1X,F5.2,1X,F5.2$
!C AGOGIN RAJOR DATA, AJOVIRTA AGGLIM, DATA AGO1.DAT
!C MLR 9.3.1988
$DIN 7$
$CAL A1=365*(LES/1000)*ADT/1000000$
$CAL SUU=4*A1$
$CAL S=%LOG(SUU)$
$CFL B1=%IF(RAD>0,RAD,1)$
$CAL B2=1/B1$
$CAL AR=(B2/-1)*B2
$CAD FJ 3$
$CAL FJ=1+%GE(JFK,3)+%GE(JFK,7)$
$CAL CR=(RAD<1)*10000+RAD$
$CAL FCR 3 FED 3 FCC 3 FLB 2$
$CAL FCR=1+%GE(CR,800)+%GE(CR,3100)$
$CAL FED=1+%GE(EDS,30)+%GE(EDS,60)$
$CAL FCC=1+%GE(CRC,5)+%GE(CRC,10)$
$CAL FLB=1+%GE(LAB,28)$
$YVAR FEA$ERROR P$LINK L$
$FIT :+S:+FJ:+FLB:+FCC:+FED:+FCR$DISP D E M$
$CAL %RE=(%LV-%FV)/%SQRT(%UT)$
$LOOK ONO FEA %FV %RE FJ FLB FCC FED FCR$
$RETURN$
```

Note for Appendix A.16

1:	Constant
S:	to which travel is raised
FJ(2):	3-7 junctions per kilometer,
FJ(3):	> 7 junctions per kilometer,
FLB(2):	< 28% lorries and buses,
FLB(3):	> 28% lorries and buses,
FCC(2):	5%-10% cracking and crazing,
FCC(3):	>10% cracking and crazing,
FED(2):	30%-60% edge spalling,
FED(3):	>60% edge spalling,
FCR(2):	800-3099m horizontal curve radius,
FCR(3):	>3100m horizontal curve radius

GLIM

C:\GLIM>echo off

GLIM 3.77 update 1 (copyright)1985 Royal Statistical Society, London

? $INPUT 9$
File name? AGGLIM
File name? AG01.DAT
scaled deviance = 157.69 at cycle 4
          d.f. = 95

scaled deviance = 149.88 (change = -7.808) at cycle 4
          d.f. = 94   (change = -1   )

scaled deviance = 139.32 (change = -11.566) at cycle 4
          d.f. = 92   (change = -2   )

scaled deviance = 134.81 (change = -3.505) at cycle 4
          d.f. = 91   (change = -1   )

scaled deviance = 131.48 (change = -3.336) at cycle 4
          d.f. = 89   (change = -2   )

scaled deviance = 123.72 (change = -7.754) at cycle 4
          d.f. = 97   (change = -2   )

scaled deviance = 120.85 (change = -2.869) at cycle 4
          d.f. = 85   (change = -2   )

Scaled deviance is 120.85   85 d.f.   1 on 95 observations
          change is   -2.869 for   -2 d.f.

          estimate   s.e.   parameter
  1   -0.1617   0.4770   1
  2   1.138   0.6950   S
  3   0.8154   0.2300   FJ(2)
  4   0.3817   0.4030   FJ(3)
  5   -0.7622   0.2674   FLB(2)
  6   0.6893   0.2863   FCC(2)
  7   0.6791   0.3569   FCC(3)
  8   0.9192   0.3594   FED(2)
  9   0.4871   0.4157   FED(3)
 10   -0.5669   0.3796   FCR(2)
 11   -0.5413   0.3470   FCR(3)
    scale parameter taken as 1.000