Accident Prediction Models for Unsignalised Junction in Malaysia Rural Roadways

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Abstract. The aim of this paper was to developed accident prediction model for unsignalised junction (access points) in Malaysia rural road. This research carried out quite number of surveys to observe traffic behavior including right and left turns from minor or major road and other road geometry characteristic. The dependent variable was modeled as accident point weighting and accident frequency. The research reveals that access point, motorcycles crossing from minor road, motorcar and motorcycle from main road, traffic light, road median, approach speed and vehicle gap, was found to be significance in the Model. The accident prediction model has a potential to improve considerably the quality of engineering aspect of accident reduction and prevention in Malaysia.

Introduction
Traffic accidents have been recognized as one of the major causes for human and economic losses both in developed and developing countries. In the year 2009, Malaysia has recorded 397,194 accidents, resulting in an average 18 deaths from road accident every single day. Meanwhile in year 2006 there were 23.6 deaths in Malaysia for every 100,000 populations recorded according to (Royal Malaysia Police), which is among the highest figures in the world compare to Netherlands and Japan has 4.5 and 5.7 deaths per 100,000 people respectively. In Malaysia, motorcycles constitute nearly half than total registered vehicle in the country with 49% meanwhile cars stated second place consist 45%. Cars casualties contribute the highest rate with 67% followed by motorcycles recorded more than 16% of the total casualties in traffic crashes (Royal Malaysia Police, 2009). Malaysia rural roadways from Batu Pahat to Ayer Hitam were among the rural road in Malaysia that has experienced 8,475 road accidents between the years 2000 and 2010, killing 234 people and injuring 1,703 people.

Road safety engineer can influence traffic safety either through means such as road regulation, legal enforcement and education or road infrastructure improvements. An overall view most of the researcher have indicated the improvement to highway design could produce significance reduction in the number of crashes.

The most common accident indicator that have so far been used are the number of accident per year (accident frequency) and the number of accident per 10,000 kilometre vehicle travelled (accident rate), however accident point weighting have rarely been account in the road accident prediction Models.

Many independent variables affect accident rate, and are generally related to traffic flow, road geometric characteristic, driver behavior and weather. Consequently from the wider set of independent parameters normally authors extort a reduced number of variables for insertion in the accident model. Hence this research will considered the following variables: motorcycle,
car, approach speed, access point, road geometric, motorcycle crossing, vehicles gap and pedestrian.

**Review of Previous Work**

The number of accidents on a given road section during a certain period of time is probabilistic in nature and is a non-negative integer. Despite the fact that accidents are random and unpredictable at a micro level, statistical models can predict reliable estimates of expected accident by relating aggregates of accidents to various explanatory measures of flow, site characteristics, and road geometry at a macro level. Numerous empirical relationships between vehicle accidents and these explanatory variables have been established in several previous studies (Miaou et al 1993). Author (Ceder et al 1993) multiple linear regression models are used. In these models the independent variable (either number of accidents or accident rate) is a function of series of independent variables such as speed or traffic volume.

A number of researchers have investigated this complex interaction in the past. One of the first such studies had analyzed accident and traffic flow on U.S Route 22 through the city of Newark, New Jersey (Gwynn et al 1967). Crash rates were plotted against hourly volume class, and the author found a distinct U-shape relationship, with more accidents observed at higher and lower traffic volumes. Roads with higher ADT and pedestrian traffic are associated with higher accident frequencies for all highway types (Berhanu et al 2004).

Inappropriate or excessive vehicle speed is one of contributory cause of the traffic accident (Barker J et al 1998). Although the connection between speeds and cars is complex, a considerable volume of research evidence shows that lower vehicle speeds result in fewer crashes. In a meta-analysis of Europe studies found that a one mph reduction in mean speed equated to a 5% drop in crash.

The crash occurrence of motorcycles at intersection is affected by their risk as well as their exposure. A number of studies (Hurt et al 1981) have reported that motorcyclist in the traffic stream are often overlooked by other driver. This describe why motorcyclist are over represented in right of way violation crashes in which vehicle from the conflicting stream encroach into the path of an approaching motorcycle. Furthermore driver tends to over-estimate the motorcycle arrival time approaching to the intersection, hence increasing the possibility of a collision (Caurd JK et al 1994).

Numerous Studies have examined driver behavior at unsignalised intersections and their response to different gaps in the priority stream. Critical gap has been defined for instances as “the minimum time gap in the priority stream that a minor street driver is ready to accept for crossing or entering the major stream conflict zone (Brilon et al 1999). Researchers as (Miller A.J 1995) used the parameter of gap acceptance in the accident model.

Traffic exposure functions such as the cross product of flows, sum of crossing flow product and the sum of encounter flow product produced much better fit to the accident data at the junction in the cities of (Mohammad Salifu 2004).

The paper moves on to develop a number of statistical models by using the Multiple Non-Linear Regression that can be used in the prediction of the expected number of accident on the two lane rural undivided roadways, Malaysia. The statistical programs and software have been created to perform this task. This involved SPSS which will be used in this project for among others, obtaining and validating the coefficients and also the model.
Model Development

A 5 year observation period from 2006 to 2010 was carried out on a four lane undivided roadways. This infrastructure stretch 47 kilometres with high density of driveways and property access. Direct access from frontage properties along the multilane roads such as housing area, school, university factory, commercial building and office lot will influence the safety aspect. Most of the infrastructure didn’t provide the road median except for some kilometre section. Base on the Figure 1, some 2,540 accident were considered in this study, 64 of which were fatal and 253 were injury while 2,223 were damage only accident.

By traversing the entire stretch of the road to observe the number of access point, median and signalized opening in every selected section were obtained. Variables such as traffic volume, approach speed, gap and vehicles movement were obtained over 2 hour time period namely morning (0800-1000), midday (1200-1400) and evening (1600-1800) by using the video camera on each kilometre section at the eleven corresponding access point.

![Figure 1: Number of accident according its type at study location (2006-2010)](image)

Site Selection

In this study, a total of 11 unsignalised junction in section 2,5,8,9 and 10 at Batu Pahat, meanwhile section 19,20,21,22,23 and 24 at Parit Raja as shown in Figure 1. From Figure 1, represent the number of accident at 11 selected unsignalised junction which classified as fatal, serious, slightly and damage over 5 years period (2006:2010). The highest number of crash is section 2 with 339 toll accident, followed by section 5 with 319 cases. Third higher accident frequency were section 23 with 272 cases. Meanwhile section 21, 10, 9, 22, 8, 19, 20 and 24 total accidents are 251, 236, 221, 194, 175, 163, 175, and 163 respectively. The highest number of fatal accident is at section 10 with 18 cases followed by section 21 with 12 deaths recorded and next was section 9 with 7 deaths toll. However section 22 doesn’t have any fatal accident. Serious injuries occurred at certain study location they were sec 5, 8, 9, 10, 21, 23 and section 24. Whilst slight injury and damage only happening at every study location. Almost all the hazardous intersections in this road are classified as access point (unsignalised single T-junction). Therefore, we will concentrate on the access points. This survey was carried out at the most dangerous access point on each of the selected sections. 11 most dangerous access points of each section were surveyed for all the 11 selected sections.
**Dependent Variable**

Accident frequency (AF) and accident point weighting (APW) data were used in this study. Both dependent variable in this modals, described as accident data consisted fatal, serious, slightly and damage base on 2 hour period which is morning (08:00-10:00) midday (12:00-14:00) and evening (16:00-18:00). APW is based on the value contribute by the Transport Research Laboratory (TRL) and it has been used in the Highway Planning Units United Kingdom. APW considering weighting for fatal (6 points), serious injury (3 points), slight injury (0.8 points) and damage (0.2 points), the system include:

\[
APW = X_1(6.0) + X_2(3.0) + X_3(0.8) + X_4(0.2) \tag{1}
\]

Therefore each sections kilometer have three AF and APW, hence they are thirty three AF and APW in 11 section kilometers were consider in this model as shown in Figure 2. The accident frequency range from 7 to 42 while accident weighting point range from 2.2 to 33.

![Figure 2: 33 Accident frequency and accident point weighing at the 11 intersection in section kilometer (2006-2010)](image)

**Independent variable**

Approach speed is defines by the 85th percentile speed measured at a range 50 meter distance from the related access point. Speeds were measured of vehicles selected a random by using video camera until a total 100 vehicles were carried out during the morning peak from 0800-0900hrs, afternoon peak 1200-1300hrs and evening peak from 1700-1800hrs. It was ranges from 59 to 99 km/hr.

Traffic study data used in this paper were based on the hourly traffic volume. Cars and motorcycles data were collected from the hourly traffic volume (disaggregated by non cars and motorcycles), counted on each kilometer section at the selected access point. The cars (car, per hour) on the rural highway range from 1,797 to 6,177, while for motorcycles (Mc, per hour) range from 165 to 1193. Pedestrian crossing counts were also carried out parallel as vehicle counts. Figure 3 shows the traffic temporal fluctuations in traffic volumes at section 5. The highest traffic volumes recorded were 1751 vehicles at 5.45 p.m. to 6.00 p.m. The second
highest was 1316 vehicles recorded at 12.15p.m to 12.30 p.m. While in the morning the highest traffic volume was 1089 vehicles recorded at 8.15a.m to 8.30a.m.

Figure 3: The temporal fluctuations in traffic volumes throughout a typical weekday

Figure 4, illustrate the quantity of traffic volume by vehicle classification at 11 kilometers section for 6 hour survey period. Motorcar and motorcycle contribute the largest proportion of traffic volume with average 61% and 20% respectively. The highest 6 hours traffic volume were section 5 with 39,954 vehicles while section 2 stated second place with 22,206 vehicles and next were section 22 account for 20,555 vehicles. However section 24 recorded lowest traffic volume with 11,223 vehicles.

Figure 4: Traffic volume by vehicles classification at 11 kilometer section for 6 hours survey period.

Access points in this study were describe as un-signalized T junction and it was a road that provides access to a specific destination such as housing area, industrial part, commercial building and schools. Access point (AP, per kilometer section) was range from 1 to 13 on the four lane rural roadway.

6 hours traffic movements data has been undertaken at the eleven corresponding T-junction as shown in Figure 5. Every motors vehicle such as motorcar, motorcycle, van and lorry, and bus were counted separately. There are 6 possible vehicles movement, each movement vehicles describe as type 1: vehicle from minor road turning right into major road, type 2: vehicles
turning right from major road into minor road, type 3: vehicles turning left into minor road while type 4: vehicles turning left form the minor road into major road, type 5: vehicles through eastbound and type 6: vehicles through westbound. Out of four turning vehicles only 2 were selected associate with accident and thus include in the model, they are type 1 and type 2. At the same time motorcycles has been chosen to represent for both movement. This parameter named as motorcycles crossing (McCrs per hour) and it ranges from 7 to 142.

Figure 5: The traffic and pedestrians flow count at Sharp factory junction.

Gap in this study defines as the time between the back of the successive motor vehicles and in front of the following motor vehicles at the corresponding access point. Total gap involved in this study were 3,300 and the range for the average gap were 1.1 seconds to 2.3 seconds.

THE ACCIDENT PREDICTION MODEL

This research applies the “Multiple Regression Method” in order to develop a model which relates accident rate to the road geometry and traffic situation. The theoretical model of accident rate is represented by the following equation:

\[ Y_i = b + b_1X_{1i} + b_2X_{2i} + \ldots + b_nX_{ni} + \varepsilon_i \quad (2) \]

Where \( Y_i \) is the accident frequency or accident point weighting (dependent variable) while all the independent variable are \( X_1, X_2 \) and \( X_{ni} \). The \( b, b_1, b_2 \) and \( b_n \) are the regression coefficients to be estimated and \( \varepsilon_i \) term is the error representing the residual difference between observed and predicted model value. Explanatory variables and the description are presented in Table 1.

The \( (\text{Sig}) \), which indicates the meaningful level to obtained coefficient for the model parameters. Generally, variables with \( \text{Sig} \) value of 0.10, 0.05 and 0.01 with confident interval (CI) of 0.90, 0.95 and 0.99 are statistically meaningful in the model. Table 1 Description of the study variable.
Table 1: Description of the study variable

<table>
<thead>
<tr>
<th>Abbr.</th>
<th>Description</th>
<th>Coding</th>
</tr>
</thead>
<tbody>
<tr>
<td>AF</td>
<td>Accident Frequency</td>
<td>AF (per hour)</td>
</tr>
<tr>
<td>APW</td>
<td>Accident point weighting</td>
<td>APW (per hour)</td>
</tr>
<tr>
<td>AP</td>
<td>Access point</td>
<td>AP (per km)</td>
</tr>
<tr>
<td>Pd</td>
<td>Pedestrian</td>
<td>Pd (per hour)</td>
</tr>
<tr>
<td>Mc</td>
<td>Motorcycle</td>
<td>Mc (per hour)</td>
</tr>
<tr>
<td>Car</td>
<td>Car</td>
<td>Car (per hour)</td>
</tr>
<tr>
<td>AS</td>
<td>Approach speed</td>
<td>AS (km per hour)</td>
</tr>
<tr>
<td>Gp</td>
<td>Vehicle gap</td>
<td>Gp (second)</td>
</tr>
<tr>
<td>MD</td>
<td>Median</td>
<td>MD (divided=1, undivided=0)</td>
</tr>
<tr>
<td>SO</td>
<td>Signelized Opening</td>
<td>SO (install=1, uninstall =0)</td>
</tr>
<tr>
<td>McCrs</td>
<td>Motorcycle Crossing</td>
<td>McCros per hour</td>
</tr>
</tbody>
</table>

Table 2: Accident Factor Estimation

<table>
<thead>
<tr>
<th>Variable</th>
<th>Model 1</th>
<th></th>
<th>Model 2</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Estimate</td>
<td>T-Stat</td>
<td>Estimate</td>
<td>T-Stat</td>
</tr>
<tr>
<td>(Constant)</td>
<td>-10.830</td>
<td>-2.679</td>
<td>-18.616</td>
<td>-1.485</td>
</tr>
<tr>
<td>(AP)0.5</td>
<td>4.904</td>
<td>4.782***</td>
<td>4.374</td>
<td>2.104**</td>
</tr>
<tr>
<td>So</td>
<td>-6.012</td>
<td>3.134***</td>
<td>3.286</td>
<td>-1.073</td>
</tr>
<tr>
<td>McCrs</td>
<td>0.031</td>
<td>1.954*</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Car</td>
<td>0.002</td>
<td>2.597***</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Mc</td>
<td>-</td>
<td>-</td>
<td>0.012</td>
<td>2.792***</td>
</tr>
<tr>
<td>(AS)(Gp)</td>
<td>-</td>
<td>-</td>
<td>0.119</td>
<td>2.049**</td>
</tr>
<tr>
<td>Pd</td>
<td>-</td>
<td>-</td>
<td>-0.116</td>
<td>-1.528</td>
</tr>
<tr>
<td>MD</td>
<td>-</td>
<td>-</td>
<td>8.588</td>
<td>1.731*</td>
</tr>
</tbody>
</table>

| R           | 0.745 | 0.788 |
| R²          | 0.555 | 0.621 |
| F           | 8.75  | 4.92  |
| Sample      | 33    | 33    |

*, **, ***=Significant at the 90%, 95% and 99% level, respectively

The final models of accident prediction are formulated from equation (3) and (4) by using non-linear regression model:

Model 1
APW = -10.830 + 4.904(AP)^0.5 - 6.012(SO) + 0.031 (McCrs) + 0.002(Car) (3)

Model 2
AF = -18.616 + 4.374(AP)^0.5 + 0.012(Mc) + 0.119 (AS)(Gp) - 0.116(Pd) + 8.588(MD) - 3.286(SO) + 0.012(McCrs) + 0.001(Car) (4)

Table 2 for model 1 shown the value multiple R and R² are 74.5% and 55.5% respectively. The effect of the access point, signalized opening and car were statistically significant (beyond 1% level) whereas motorcycles crossing variable influenced at 10% level. Model 2 represent
the value of multiple R and $R^2$ were 78.8% and 62.13% respectively. The impacts of motorcycles were highly effected with significant at 1% level meanwhile approach speed, gap and median influenced at 10% significance level.

**DISCUSSION AND CONCLUSION**

By using Model 1, the percent accident reduction through changing the measures of each parameter are, one access point per kilometer reduction can reduce accident by 18% and installing of 1 signalized opening at the access point would reduce number of crash by 42%. Meanwhile using the model 2, 10 kilometer per hour reduction in speed equated to a 9% drop in crash, per long the 0.1 second in gap will cut down number of accident by 3%. Furthermore constructing the median would lessen the accidents by 35%.

The model developed in this paper appear to be useful for many application such as the existing number of access point, increasing number of motorcycle and motorcycle crossing, rise in speed and shorted in gap are among the potential contribute of increment accident rate. Moreover this paper revealed that by constructing the road median and installing the opening signalized at the hazardous access point would drastically reduce the number of accident. Thus, these studies are realistically convinced that this latter may contribute a point of reference for the engineer in improving or designing multilane rural roads.

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**References**


