

Assessment pedestrian crossing safety using vehicle-pedestrian interaction data through two different approaches

Fixed videography (FV) vs In-Motion Videography (IMV)

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1 Assessment pedestrian crossing safety using vehicle-pedestrian
2 interaction data through two different approaches:
3 Fixed Videography (FV) vs In-motion videography (IMV)
4

5 **Abstract**

6 A significant portion of pedestrian accidents occurs in the outskirts areas due to the high vehicle speed and lack
7 of safety facilities for pedestrians. Behavioral study on drivers and pedestrians is the key to better understand the
8 causes of pedestrian accidents in order to develop safety models.

9 Despite numerous studies on pedestrian safety based on various roads, outskirts areas have not been considered.
10 Hence, the present study focuses on evaluating the safety of pedestrian crossing in urban and outskirts areas and to
11 determine the differences of drivers and pedestrians' behaviors between these areas through data based on fixed
12 videography (FV) and in-motion videography (IMV). These approaches may lead to an exact analysis of the
13 behavioral differences of road users behaviors from the perspective of pedestrians (FV data) and drivers (IMV
14 data) in urban and outskirts roads. Accordingly, behavioral studies were conducted at urban and outskirts sites
15 through FV as well as IMV using the behavior of 29 participants in the same roads in Babol city, Iran. The gap
16 acceptance model using linear regression and pedestrian crossing probability model using logistic regression for
17 both approaches showed similarity on results in both urban and outskirts roads. Furthermore, behaviors of
18 pedestrians crossing and drivers' yielding on urban and outskirts roads were very similar. Vehicle speed, the
19 distance of vehicle to pedestrian at the possible collision point, size of pedestrian groups, and waiting time before
20 crossing were the most important behavioral differences of pedestrian for choosing a gap acceptance and
21 probability of crossing on various sites through two different approaches. The inference of the models obtained
22 in this study will lead to a better understanding of the behavior of road users for studies on advanced driving
23 assistance systems (ADAS).

24 **Keywords:** Pedestrians behavior, videography, Drivers behavior, outskirts road, urban road
25
26

27 **1. Introduction**

28 According to the latest report by the World Health Organization (WHO) on the road safety, pedestrians
29 are the most vulnerable group of road users who account for a significant share of road accidents (WHO,
30 2018). According to this report, an average of about 23% of the world's annual 1.35 million road
31 fatalities involves pedestrians. On the other hand, the number of fatalities is not the same in different
32 countries, with the middle-income countries accounting for about 80% of road fatalities despite a share
33 of 59% of the total number of motor vehicles worldwide. Pedestrians' share of road accidents in different
34 countries varies from 14% to 40% of total road accidents. For example, pedestrians account for about
35 27% of Europe's road fatalities, while only 14% of in East Asian countries. On the other hand, pedestrian
36 fatalities in road accidents in the African continent are reported at around 40%, accounting for the
37 highest rate of fatalities among continents. Pedestrian fatalities in the Oceania and Americas are
38 reported at around 22, which is close to the average pedestrian fatalities around the world (23%) (WHO,
39 2018). The remarkable point about pedestrian accidents is that much of the pedestrian accidents occur
40 in some particular areas (OECD, 2017; WHO, 2018). A sample of recent of International Road
41 Assessment Programme (iRAP) assessments from 54 countries, covering 358,000 kilometres of rural
42 and urban roads with over 700 billion vehicle kilometres of travel a year has highlighted that 88% of
43 pedestrian travel is on one or two-star roads (WHO, 2018). According to this, one or two-star roads are
44 referred to roads without any sidewalk or safe crossing while the minimum speed of vehicles is 60
45 kilometers per hour. The importance of this issue is that the excessive growth of the global population
46 as well as the geographical/environmental characteristics have led to a saturation of population density
47 in some cities. So in these cities, people are obliged to live in outskirts areas. Living in these areas and
48 daily commuting to different urban areas are common around the world including in the northern part
49 of Iran. What is important in this regard is the low level of safety on the roads that connect these outskirts
50 areas to urban areas. To put it simply, considering that generally no safety measures are taken for the
51 commuting of pedestrians on these roads, such roads are among the one-star roads, which, as reported
52 by the WHO, are the most important pedestrian accident sites around the world (WHO, 2018).
53 According to the Iranian Legal Medicine Organization report, pedestrian accident rate in the suburban
54 roads are more than 2.5 times that of urban ones. A significant share of these fatalities is on the outskirts
55 areas (Iranian Legal Medicine Organization, 2018). From among 3505 pedestrians killed in the
56 accidents in 2018, more than 66% have been killed on suburban roads, with 74% of them being killed
57 on outskirts roads connecting to urban areas (Iranian Legal Medicine Organization, 2018).
58 Therefore, although several studies have been carried out on pedestrian crossing safety to date, as far
59 as we know, these studies have been on urban areas, and no studies have been carried out on pedestrian
60 safety on outskirts area yet. Considering uneven statistics on pedestrian accidents in both urban and
61 outskirts area, the present study is to investigate the pedestrian decision-making performance in choosing
62 safe gaps on both areas. Therefore, two hypotheses investigated in the present study are: a) diversity of
63 gaps chosen by pedestrians on urban and outskirts roads, and b) Similarity of the results of the data
64 from FV (pedestrian' perspective) and IMV (driver's perspective). In addition, the research questions
65 are:

- 66 • What are the parameters that influence the decision of crossing/not crossing of pedestrians in
67 these two areas? Are these parameters different in these two areas?
- 68 • Is the gap chosen by pedestrians dependent on the behavior of the driver of the approaching
69 car?

70 **2. Literature review**

71 Investigation of pedestrian behavioral parameters as well as traffic characteristics of pedestrian
72 crossings plays an important role in identifying effective factors in pedestrian accidents. The effects of
73 environmental and traffic characteristics, as well as other road users' behaviors have led to the
74 complexity of the analysis of pedestrian crossing behaviors (Bichicchi et al., 2017; Etehad et al., 2015;
75 Khattak & Tung, 2015; Ram & Chand, 2016; Sheykhfard & Haghighi, 2019; Wang et al., 2018).

76 So far, numerous studies have evaluated pedestrian safety using various methods. Most of the data used
77 in these studies were obtained from accident database (Huang et al., 2018; Johnson et al., 2013; Khattak
78 & Tung, 2015; Mohamed & Bromfield, 2017), questionnaires (Mwakalonge et al., 2015; Ram & Chand,
79 2016; Rolison et al., 2018) and field observations (Brosseau et al., 2013; Kadali & Vedagiri, 2013; Poó
80 et al., 2018; Wang et al., 2018). Pedestrian safety simulation in driving simulators is another approach
81 used by researchers to investigate pedestrian traffic safety (Erkuş & Özkan, 2019; Gómez et al., 2013;
82 Obeid et al., 2017). In numerous studies on pedestrian behaviors, environmental factors such as road
83 width and traffic volume (Granié et al., 2014), road light situation (Liu & Tung, 2014), or weather
84 conditions such as rainfall (Bedeley et al., 2013), have been reported to affect pedestrians decisions to
85 cross the road. Demographic characteristics of pedestrians such as gender and age (Sun & Benekohal,
86 2003; Yannis et al., 2013), as well as pedestrian risky behaviors like running on the road (Zhuang &
87 Wu, 2011) can lead to changes in pedestrians' decisions when cross. On the other hand, the vehicle's
88 speed and the distance to the pedestrian (Hunter et al., 2015), the type of vehicle (Serag, 2014) and the
89 type of traffic control devices (Minhas et al., 2017) have been reported to affect the pedestrians' decision
90 to cross the road.

91 92 *2.1. Pedestrian viewpoint (Fixed videography)*

93 The use of FV at pedestrian crossings has compensated for many of the limitations of traditional
94 methods, such as the use of accident statistics or field observations by the research team. Today, making
95 up for the lack of data in some areas, inaccurate data logging, and important variables of causative
96 accident patterns are the most important advantages of using FV in pedestrian traffic safety research.
97 Due to the importance of pedestrian traffic on various urban roads, extensive studies have been
98 conducted by researchers to investigate the human, environmental and other possible variables of
99 pedestrian accidents. In some of these studies the traffic variables such as vehicle type and pedestrian
100 traffic volume (Qi & Guoguo, 2017), traffic control devices such as smart signs (Xu et al., 2018),
101 environmental and geometrical variables such as weather conditions and pavement situation, number
102 of lanes, or their widths (Bedeley et al., 2013; Kadali & Vedagiri, 2013; Ni et al., 2016) are the main
103 variables affecting pedestrian traffic safety. However, in most studies, the human factors have been
104 reported to be more important than others in the possibility of pedestrian accidents. According to these
105 studies, the set of behaviors and actions of pedestrians and drivers plays a major role in increasing or
106 decreasing the safety of pedestrian. In other words, pedestrians' decision to safely cross the road is
107 highly connected with the gap acceptance by them, and in fact the safe gap selected can ultimately
108 indicate their decision to cross the road. Several studies have shown that various parameters such as
109 pedestrian crossing behaviors, i.e. zigzag movement (Zhang et al., 2017) or running (Zhuang & Wu,
110 2011), pedestrian physical condition (Salamati et al., 2013), pedestrian age (Liu & Tung, 2014; Sun &
111 Benekohal, 2003), pedestrian gender (Yannis et al., 2013), accidents experience (Narváez et al., 2019),
112 pedestrian group size (Hunter et al., 2015; Sheykhfard & Haghghi, 2018) have been investigated to
113 analyze the selection of the safe or unsafe gap by pedestrians. In addition, some studies have
114 recommended the more attention to the behaviors of pedestrians and drivers when encountering each
115 other. Therefore, in recent years, the behavior of drivers has become a common research subject in the
116 study of pedestrian traffic safety. Following are some of these studies.

117 118 *2.2. Driver viewpoint (In motion videography)*

119 Naturalistic Driving Studies (NDS) through IMV data were used to examine the behavior and action of
120 drivers and other road users as they interact with each other. This is a new approach to current traffic
121 investigation methods and provides information that is difficult or even impossible to obtain by other
122 methods. In this realistic approach, user behavior is continuously studied in the natural environment of
123 the road for a long time. Recording from the driver's perspective allows examining the driver's behavior

124 during daily trips and identification and analysis of the relationship between the driver, the vehicle, road
125 and other road users during different driving conditions (Sheykhfard & Haghghi, 2019; van Schagen
126 et al., 2012). The term NDS presents a flexible approach to assess road users' behavior especially
127 drivers. This approach helps researchers to better study driving behavior by monitoring the observance
128 of the driving tasks and the road environment, as well as giving recommendations on what measures
129 the driver should take before or near a collision (Venter, 2014). In NDS, vehicles are driven in real-life
130 traffic conditions and are equipped with cameras and sensors that can record video and provide
131 information on the driving, flow of traffic and the characteristics of environment before and at the time
132 of occurrence of an interaction and traffic accidents. In recent years, studies have been conducted aimed
133 at the pedestrian safety through the driver's perspective, the results of which indicate the effects of the
134 drivers' behavior and action as well as the pedestrians' behavior on pedestrian traffic safety (Habibovic
135 et al., 2013; Lin et al., 2018; Sheykhfard & Haghghi, 2018). The use of videography, especially through
136 in-motion style, to examine the safety of pedestrians has led to a better understanding of the interactions
137 of drivers, pedestrians and other road users in relation to each other. Studies in recent years have shown
138 that driving errors in the form of inappropriate reactions when encountering the pedestrian can increase
139 the potential for collision between them (Cafiso et al., 2017; Habibovic et al., 2013; Sheykhfard &
140 Haghghi, 2018).

141 To summarize, the use of FV and IMV data in recent years has significantly compensated some of the
142 limitations of using previous methods. Applying these data provided a significant opportunity for
143 researchers to study pedestrian safety from different perspectives in relation to influential factors such
144 as human, environment, vehicle, and road. Despite numerous studies based on various scenarios on
145 pedestrian safety through FV (Åbele et al., 2018; Hunter et al., 2015; Minhas et al., 2017b; Salamati et
146 al., 2013; Serag, 2014) and IMV (Alferova et al., 2017; Antić et al., 2016; Habibovic et al., 2013; Lin
147 et al., 2018; Sheykhfard & Haghghi, 2018; Sheykhfard & Haghghi, 2020), to the best of our
148 knowledge, no studies have been conducted on pedestrian safety on the outskirts areas connecting to
149 the urban areas. However, according to statistics (Iranian Legal Medicine Organization, 2018;
150 Organisation for Economic Co-operation and Development (OECD) Staff et al., 2017; WHO, 2018), a
151 significant portion of pedestrian accidents annually occur on these routes. Therefore, the present study
152 is to evaluate the safety of pedestrian crossing in urban and outskirt areas and to determine the
153 differences of drivers and pedestrians' behavior between them, both from the pedestrian's perspective
154 (FV) and from the driver's perspective (IMV).

155 **3. Methodology and materials**

156 *3.1. Study sites*

157 Due to its unique geographical and climatic conditions, there is scattered population in Mazandaran
158 province in northern Iran (area of about 24000 km² and a population of over 3 million). For this reason,
159 various outskirts areas have been built around the cities so that daily commute from outskirts to urban
160 areas has become a habit in these this province. According to Iranian Legal Medicine Organization
161 report, Mazandaran is one of the three provinces with the highest number of road accidents in Iran
162 (Iranian Legal Medicine Organization, 2018). Also, about 700 people are killed on different roads in
163 Mazandaran province annually, with pedestrians accounting to about 28% of the fatalities. About half
164 of these fatalities are related to the urban and outskirts roads of Babol city. The present study is to
165 investigate different routes of urban and outskirts roads in this city in order to evaluate pedestrian safety.
166 Therefore, four urban and outskirts roads (two sites per each area) were selected as the study sites
167 through FV. These sites were chosen due to the high number of traffic accidents as well as the high
168 volume of pedestrian traffic based on the province legal medicine organization statistics. Some
169 characteristics of these roads are given in Table 1.

170

171 **Table1**

172 Geometric and traffic characteristics of outskirts and urban roads

Characteristics	Unmarked Crosswalks	Marked Crosswalks
Crossing type	Uncontrolled crossing	Uncontrolled crossing
Posted speed limit	60km/hr.	40km/hr.
Total of lanes	4+ median	3+ median
Lane width	4m	3.75m
Direction of traffic	Two-way	Two-way

173
174 **3.2. Data collection**

175 **3.2.1 FV data**

176 At each urban and outskirts areas, videography was made by installing a fixed camera on the highest
177 building overlooking the roads. At each site, video was recorded via the camera Panasonic HC-V180
178 Camcorder (30 frames per second) at three 1-hour intervals. Therefore, video was recorded from 7:30
179 to 8:30 in the morning, 12:30 to 13:30 in the afternoon and 18:30 to 19:30 in the evening. Meanwhile,
180 recording was done in July 2019 and the weather was clear and sunny during recording. Fig 1 shows
181 pictures of some of these sites.



182
183 Fig. 1. FV in study sites (Left: Outskirt road; Right: Urban road)

184 **3.2.2. IMV data**

185 NDS studies were carried out by installing cameras in the cars of the participants. A total of 29
186 participants (15 male drivers, 14 female drivers; 18-65 years old) participated in current study in the
187 city of Babol, Iran (Fig 2). In order to avoid the possible impact of the study on their driving behavior,
188 information regarding the objective of the research was not provided to the participants. On average,
189 the NDS was examined about 30 minutes for each participant. IMV was conducted via The CARPA-
190 120 Dual Dash cam Video Resolution 640 x 480; Recording Frame Rate of 30 fps) during August 2019.
191 Also, the variables extracted from the FV and IMV records are listed in Table 2.



192

194 **Table2**
195 Variables

Code	Variables	Definition and calculation
<i>R.G</i>	<i>Rolling Gap</i>	<i>Crossing on a zigzag motion by pedestrian</i>
<i>Group</i>	<i>Group</i>	<i>Whether the pedestrians cross the group or alone; Group: 1, Alone: 0</i>
<i>W.T</i>	<i>Waiting time</i>	<i>Time that each pedestrian waits to accept a gap</i>
<i>H.T.C</i>	<i>How to cross</i>	<i>Running: 1; Walking: 0</i>
<i>Gender</i>	<i>Gender</i>	<i>Male: 1; Female: 0</i>
<i>Choice</i>	<i>Choice</i>	<i>Whether pedestrians start crossing immediately; Cross: 1; Wait: 0</i>
<i>D</i>	<i>Distracted</i>	<i>Whether pedestrian is distracted by a secondary task while crossing (using phones, eating, luggage etc.) Yes: 1, No: 0</i>
<i>ATT</i>	<i>Attention</i>	
<i>V.T</i>	<i>Vehicle type</i>	<i>Looking towards the approaching vehicle; Yes: 1, No: 0</i>
<i>S_p</i>	<i>S_p</i>	<i>Vehicle type</i>
<i>S_v</i>	<i>S_v</i>	<i>Pedestrian speed</i>
<i>X_p</i>	<i>X_p</i>	<i>Vehicle speed</i>
<i>X_v</i>	<i>X_v</i>	<i>The distance from the pedestrian to the potential collision point</i>
<i>T.C</i>	<i>Time to cross</i>	<i>The distance from the vehicle to the potential collision point</i>
<i>S.M</i>	<i>Safety margin</i>	<i>The time takes for a pedestrian to cross the road</i>
<i>Gap</i>	<i>Gap acceptance</i>	<i>The time interval between passage a pedestrian from a point on the road until the next vehicle reaches that point</i> <i>Time difference between the moment the pedestrian is ready to cross the road until the vehicle reaches the collision possible point</i>
<i>P.crossing</i>	<i>Pedestrian crossing</i>	<i>Whether pedestrian could cross or not; Yes: 1; No: 0</i>

196

197 **3.3. Methods**

198

199 In this study, linear regression and logistic regression models were used for analysing gap acceptance
200 behavior and possibility of crossing behaviour of pedestrians, respectively. In the first step, the linear
201 regression model through SPSS (v.24) software were was used to determine the factors affecting gap
202 acceptance behavior. In linear regression, the value of one variable (the dependent variable such as y)
203 are estimated from the values of two or more other variables (independent variables such as $x_1, x_2, x_3,$
204 \dots, x_k). This is done by generating a linear equation as follows:

$$205 \quad Y = b_0 + b_1x_1 + b_2x_2 + b_3x_3 + \dots + b_kx_k \quad (1)$$

206 Where, the parameters b_1, b_2, \dots, b_k are the regression coefficients, and the intercept (b_0) is the
207 regression constant. To prevent correlation between independent variables and its negative effect on
208 dependent variable, Pearson correlation test was used for continuous variables and Chi-square test was
209 used for discrete variables. In the initial model, all variables (except the variables that were correlated
210 with each other) were studied, then the variables that had little effect on the dependent variable
211 (acceptance gap) (significance level greater than 5%) were excluded from the model. In the next step,
212 the remaining variables from the previous step (highly effective variables) were re-analyzed and, the
213 final model was determined. Then, logistic regression was used to examine the correlation between
214 pedestrian crossing probability and the independent variables. In the initial logistic regression model,
215 all variables (except for correlated variables) were analyzed by SPSS software. So, the variables that
216 had little effect on the model depend variable (accepted or rejected gap) ($\text{sig} > 0.05$) were excluded from
217 the model. In the next step, modeling was performed again with the remaining variables and the model
218 was determined with the best fit and the coefficients of the independent variables and their constant
219 values were calculated. Therefore, the probability that the pedestrian crossing the road is obtained by
220 the Eq. (2). Finally, influential variables along with their impact coefficients in the final model were
221 determined.

$$222 \quad \Pr(Y_i = 1 | x) = \frac{e^{\text{logit}(p)}}{1 + e^{\text{logit}(p)}} \quad (2)$$

223 $\text{logit}(p) = \ln\left(\frac{p}{1-p}\right) = \alpha + \beta_1 x_{1,i} + \beta_2 x_{2,i} + \dots + \beta_k x_{k,i} \quad , \quad i=1, 2, \dots, n$

224 **4. Results and discussion**

225 *4.1. Descriptive statistics*

226 4.1.1. FV data

227 Investigating films from FV using Tracker Software showed that there have been 298 accepted gap
228 (cross) and 108 rejected gap (not to cross) in the to the outskirts areas. Also, in the urban areas, 616 gap
229 were accepted and 114 were rejected by pedestrians. Observing IMV data also showed that in the
230 outskirts areas, there were 231 accepted and 87 rejected gaps. There were 176 accepted and 69 rejected
231 gaps on the urban areas. Table 3 shows the descriptive statistics for each site. In addition, the variables
232 considered for the gap acceptance behavior modeling and the crossing probability model are provided
233 in Table 2. According to Table 3, pedestrians are almost 1.7 times more likely to reject the gap in the
234 outskirts than urban areas. Comparing 108 rejected gaps against 298 accepted ones in the outskirts areas
235 indicates that about 27% of attempts in crossing the roads have failed. On the other hand, on urban
236 routes, 16% of pedestrians failed in their first attempt to cross the road. The results also showed that
237 pedestrian women behaved more cautiously than pedestrian men and attempted to use larger gaps to
238 cross the road. Investigating the videos showed that in the same conditions (distance to approaching
239 vehicle and vehicle speed), female pedestrians use gaps 3 and 2 times as more as men in the outskirts
240 and urban areas, respectively. The number of pedestrians waiting to cross the road was another factor
241 that led to the difference in the number of gaps accepted and rejected by pedestrians. As can be seen in
242 Table 2, with the increase in the number of pedestrians waiting on the outskirts and urban roads, the
243 number of gaps accepted by the pedestrians increased. Aggressive behaviors such as running on the
244 outskirts roads were observed more than 2.5 times higher than that on the urban roads while accepting
245 the gap by the pedestrians, indicating a high crossing risk on the outskirts roads. Viewing the films also
246 showed that when accepting the gap, about 21% and 38% of pedestrians lost their attention to road
247 especially to approaching vehicle via an internal or external distraction factor, respectively. Using
248 mobile phones, talking to other pedestrians and also carrying luggage, like backpacks, accounted for
249 69, 22 and 9% of pedestrian distractions in the outskirts areas, respectively. In urban areas, use of mobile
250 phones (61%), advertises of shops and advertising digital billboards (18%), talking to pedestrians (7%),
251 and carrying luggage like backpacks (14%) were the causes of pedestrian distraction. The type of
252 vehicles passing on the outskirts also caused varied numbers of gaps. Analyzing of videos showed that
253 under the same conditions (distance to approaching vehicle and the vehicle's speed) the probability of
254 accepting the gap by the pedestrians lowered with approaching heavy vehicle by about 43% relative to
255 light vehicles. On the other hand, video analysis showed that pedestrians who looked at the approaching
256 vehicles when crossing were more likely to cross than those ignoring the approaching vehicle, even if
257 their chosen gap is smaller. Pedestrian crossings on outskirts roads were significantly different than
258 pedestrian crossings on the urban roads, such that the probability of risky behaviors, like the zigzag
259 movement, during acceptance of the gap by pedestrians was twice as high in outskirts roads as the urban
260 roads.

261

262 **Table3**
 263 Descriptive statistics

Variables		Gap							
		FV				IMV			
		Outskirt roads		Urban roads		Outskirt roads		Urban roads	
		Accepted (298)	Rejected (108)	Accepted (616)	Rejected (114)	Accepted (231)	Rejected (87)	Accepted (176)	Rejected (69)
Gender	Male	181	29	357	38	138	49	119	39
	Female	117	79	259	76	93	28	57	30
Number of pedestrians	1	159	68	319	63	132	38	79	38
	2	69	22	164	30	59	23	51	18
	3	35	11	74	11	21	15	20	11
	4	22	5	49	7	17	8	18	1
	5 and more	13	2	26	3	2	3	8	1
Run	Yes	213	-	167	-	179	-	42	-
	No	85	-	449	-	52	-	134	-
Distracted by internal/external factor	Yes	63	37	234	46	42	50	51	42
	No	235	71	382	68	189	37	125	27
Type of vehicle	Light	209	63	616	114	231	87	176	69
	Heavy	89	45	-	-	-	-	-	-
Attention to approaching vehicle	Yes	266	91	499	101	198	64	157	59
	No	32	17	117	13	33	23	19	10
Crossing path	Perpendicular	64	-	223	-	77	-	93	-
	Oblique	98	-	289	-	49	-	50	-
	Rolling	136	-	104	-	105	-	33	-

264

265 4.1.2. IMV data

266 Data extracted from the videos recorded inside the car show that there is a pattern almost similar to the
 267 variables discussed in the previous section. In this approach, the likelihood of rejected gap by the
 268 pedestrians was also higher on outskirts roads (1.26 times) than the urban ones, which can be attributed
 269 to the high speed of the vehicles as one of the factors. It was also found that pedestrian women use
 270 larger gaps than men, indicating their higher conservativeness when crossing the roads. The distracting
 271 behaviors in this approach were also factors that influenced pedestrian acceptance or rejection of the
 272 gap. It was also observed that pedestrians are more inclined to cross the urban roads in direct and oblique
 273 paths, which can be attributed to the narrowness of the roads. On the outskirts areas, due to the wideness
 274 of the road, the pedestrians attempted to cross the road in a zigzag movement path, which in most cases
 275 accompanies running. It is predictable that doing so by the pedestrians will increase the potential for
 276 accidents.

277 4.2. Pedestrian gap acceptable behavior

278 4.2.1. Modelling

279

280 Table 4 show acceptance gap models through these two data collection approaches in both urban and
 281 outskirts areas. As shown in Table 4, the variables affecting the gap coefficients have different effects
 282 on the dependent variable (acceptance gap). It is important to note that at both areas, the models obtained
 283 through FV and IMV videography data are mostly similar. This result indicates that although a different

284 approach has been considered in the present study, the results are very similar. It should be noted that
 285 based on the values of the Bi column of the coefficients table, it cannot be concluded that the variable
 286 with a higher coefficient (regardless of its sign) has a greater effect on the dependent variable, because
 287 the measurement units of variables in this column are different. Therefore, the standardized beta
 288 coefficients column (regardless of its sign) is used to compare the effects of the variables. Beta
 289 represents the rate of change in the dependent variable as per the change as big as a standard deviation
 290 in the independent variable. According to the Kolmogorov-Smirnov (KS) test value (more than 0.05;
 291 95% significance level) for models through both data collection in case studies; a) outskirt road by FV
 292 (0.316), b) outskirt road by IMV (0.392), c) urban road by FV (0.418) and d) urban road by IMV (0.473),
 293 it can be concluded that the distribution of the dependent variable (acceptance gap) is normal. In
 294 addition, about 87% of the acceptance gap in the outskirts roads is explained by the effective variables
 295 identified on models in outskirt road by FV and outskirt road by IMV which is about 90% in urban road
 296 by FV and urban road by IMV (urban roads). The values of the Durbin-Watson test for outskirt road by
 297 FV (2.057), outskirt road by IMV (1.935), urban road by FV (1.855), and urban road by IMV (1.711)
 298 indicate the independence of errors from each other, considering that their values are limited to the
 299 interval 1.5 to 2.5. Ultimately, the final model is fit considering the values of different tests.

300 **Table4**
 301 Estimation linear regression model results of effective factors

Code	Road type	Approach	Unstandardized (Bi)	Standardized beta	(t-value)	Sig
Constant	Outskirt	FV	-0.408	-1.025	2.005	0.000
		IMV	-1.394	-3.881	3.115	0.004
	Urban	FV	-3.313	-0.045	3.672	0.002
		IMV	-0.237	-0.091	4.815	0.003
Gender	Outskirt	FV	6.176	1.320	3.216	0.004
		IMV	8.164	0.918	6.705	0.009
	Urban	FV	-	-	-	-
		IMV	-	-	-	-
V.T	Outskirt	FV	4.058	0.497	4.809	0.000
		IMV	-	-	-	-
	Urban	FV	-	-	-	-
		IMV	-	-	-	-
S_p	Outskirt	FV	13.059	0.041	8.002	0.010
		IMV	19.216	0.082	5.061	0.000
	Urban	FV	8.492	0.116	6.662	0.015
		IMV	7.059	0.107	5.608	0.010
X_p	Outskirt	FV	8.467	0.018	6.431	0.005
		IMV	14.149	0.044	4.008	0.000
	Urban	FV	10.025	0.102	2.065	0.003
		IMV	9.803	0.099	3.094	0.006
T.C	Outskirt	FV	1.618	0.731	5.547	0.005
		IMV	3.398	0.485	6.043	0.002
	Urban	FV	3.255	0.158	7.035	0.009
		IMV	2.299	0.197	6.726	0.010
Group	Outskirt	FV	-	-	-	-
		IMV	-	-	-	-
	Urban	FV	1.141	0.716	3.363	0.023
		IMV	1.865	0.828	3.524	0.017
Choice	Outskirt	FV	-	-	-	-
		IMV	-	-	-	-
	Urban	FV	-7.695	-1.690	-2.047	0.034
		IMV	-5.967	-1.484	-2.906	0.029

302

303 Analysis of the significant coefficients for each of the variables shown in the Table 4 indicates that in
304 the outskirts roads, the gender and crossing time variables have the most impact on the acceptance gap.
305 According to the results, pedestrian women tend to use larger gap to cross the road than pedestrian men,
306 which can be attributed to their conservative behavior. Passengers who needed more time to cross tried
307 to avoid smaller gap. Another influencing factor was the type of vehicle, and pedestrians choose larger
308 gap when facing larger vehicles. Although the speed and acceleration of these vehicles are lower than
309 that of cars, and therefore under similar conditions, pedestrians are more likely to cross safely when
310 faced with heavy vehicles. But because of the dimensions of these vehicles, pedestrians prefer to use
311 larger gap. The speed and distance of the vehicle were other variables that the pedestrians used to choose
312 the gap. On the urban roads, the effective variables identified by IMV data analysis were the same as
313 those identified by the analysis of data from FV data, with two differences. First, due to the use of
314 cameras inside the cars, it was not possible to assess the impact of the type of vehicle. The ban on heavy
315 vehicles on certain roads due to specific geometrical features led to the fact that, in the present study,
316 we could not investigate the impact of vehicle type on acceptance gaps by means of IMV. The second
317 difference is about the effectiveness of variables that are based on the results of FV data analysis,
318 although gender and time to cross are still the most important variables for crossing, but compared to
319 the results of IMV data analysis, they have a lesser effect on gap selected. On the other hand, the effect
320 of the distance between the vehicle and the pedestrian was almost similar to that of the FV analysis
321 approach. The speed of the vehicle also had a relatively greater effect on the acceptance gap in the IMV
322 data analysis than FV ones.

323 Acceptance gap model for pedestrians on urban roads not only is influenced by variables such as vehicle
324 speed, distance between vehicle and pedestrian and time required to cross the road (these variables were
325 effective in the model of outskirts roads), but also it showed significant differences with pedestrian
326 group size and choice variables. Data analysis showed that with the increase in the number of
327 pedestrians waiting to cross the road, the probability of choosing larger gap increased. Therefore, it
328 seems likely that the probability of accident for the pedestrians moving in groups decreases due to
329 choosing larger gap. On the other hand, the negative coefficient of choice variable indicates that
330 pedestrians who choose to immediately cross the road (not waiting before moving) use smaller gap,
331 which may increase the likelihood of the pedestrian-vehicle collision under dangerous conditions (high
332 speed cars or slow pedestrians' speed in short distance between them). In summary, the acceptance gap
333 models for urban roads are mostly similar in fixed and in-motion videography data, which is also the
334 case with the obtained models of outskirts roads.

335 *4.3. Pedestrian crossing behavior*

336 The positive coefficients of the variables indicate their direct impact on the increased crossing
337 probability of pedestrian, and the negative coefficients indicate that any increase in the value of a
338 variable can lead to a lower crossing probability of the pedestrian. It is therefore understood that
339 variables such as distance between the vehicle and the pedestrian, group size of pedestrians waiting to
340 cross, as well as waiting time when selecting an acceptance gap will increase the likelihood of
341 pedestrians crossing. Similarly, it can be concluded that the increase in the speed of vehicles will reduce
342 the crossing probability of pedestrians. Table 5 shows that the coefficients and variables identified for
343 the same road using FV and IMV data are almost identical. Also, the higher the coefficient of a variable
344 (regardless of its positive or negative sign), the greater its effect on the target variable (the crossing
345 probability). Therefore, the group size of pedestrians at the start of movement is one of the most
346 important factors affecting the crossing likelihood which also increases the safety. For example Table
347 5 shows that, using video-based data, the crossing probability of pedestrians is shown to be 6.43 times
348 higher if they choose to cross the road as a group ($\text{Odd} = e^{\beta} = e^{1.862} = 6.43$). On the other hand, the
349 distance between the vehicle and the pedestrian was another important factor that increased the crossing

350 probability of pedestrians. Also, the values obtained for the Nagelkerke R square coefficients for models
 351 were between 0.69 and 0.73, indicating that the independent variables of these models have good
 352 explanatory power regarding the variance and variations of the dependent variable of the crossing
 353 probability. The fitting of the models was also evaluated by the Hosmer-Lemeshow test. This test shows
 354 the correspondence between the numbers of observed and expected cases for the two classes of crossing
 355 or non-crossing, with a significance level greater than 5% ($\text{sig} > 0.05$) indicating that there is a good fit
 356 for the obtained models.

357 **Table5**
 358 Estimation binary logit model results of effective factors

Model	Code	Approach	(Bi)	(t-value)	Sig
Outskirt road	Constant	FV	0.089	3.152	0.015
		IMV	0.126	4.835	0.013
	S_v	FV	-0.132	-5.942	0.000
		IMV	-0.108	-6.186	0.005
	X_v	FV	0.552	3.144	0.031
		IMV	0.716	3.679	0.019
	Group	FV	1.862	3.223	0.006
		IMV	1.428	3.629	0.008
	W.T	FV	0.417	2.992	0.042
		IMV	0.588	3.002	0.035
Urban road	Constant	FV	0.008	2.964	0.026
		IMV	0.035	3.045	0.005
	S_v	FV	-0.095	-5.468	0.002
		IMV	-0.045	-4.982	0.004
	X_v	FV	0.325	4.145	0.020
		IMV	0.314	4.512	0.028
	Group	FV	0.689	4.152	0.000
		IMV	0.476	6.181	0.000
	W.T	FV	0.236	3.943	0.039
		IMV	0.211	4.003	0.038

359
 360 Fig.3 shows the status of pedestrians crossing or not crossing in the outskirts and road urban roads. FV
 361 and IMV-based data are shown concerning the two variables of vehicle speed and vehicle-to-pedestrian
 362 distance. Regard to both plots, Fig.3 shows that there is very similar pattern of pedestrians' crossing
 363 status based on both FV and IMV data. On urban roads, the likelihood of pedestrians crossing (filled
 364 blue circle) is increased when the approaching vehicle speed is low.

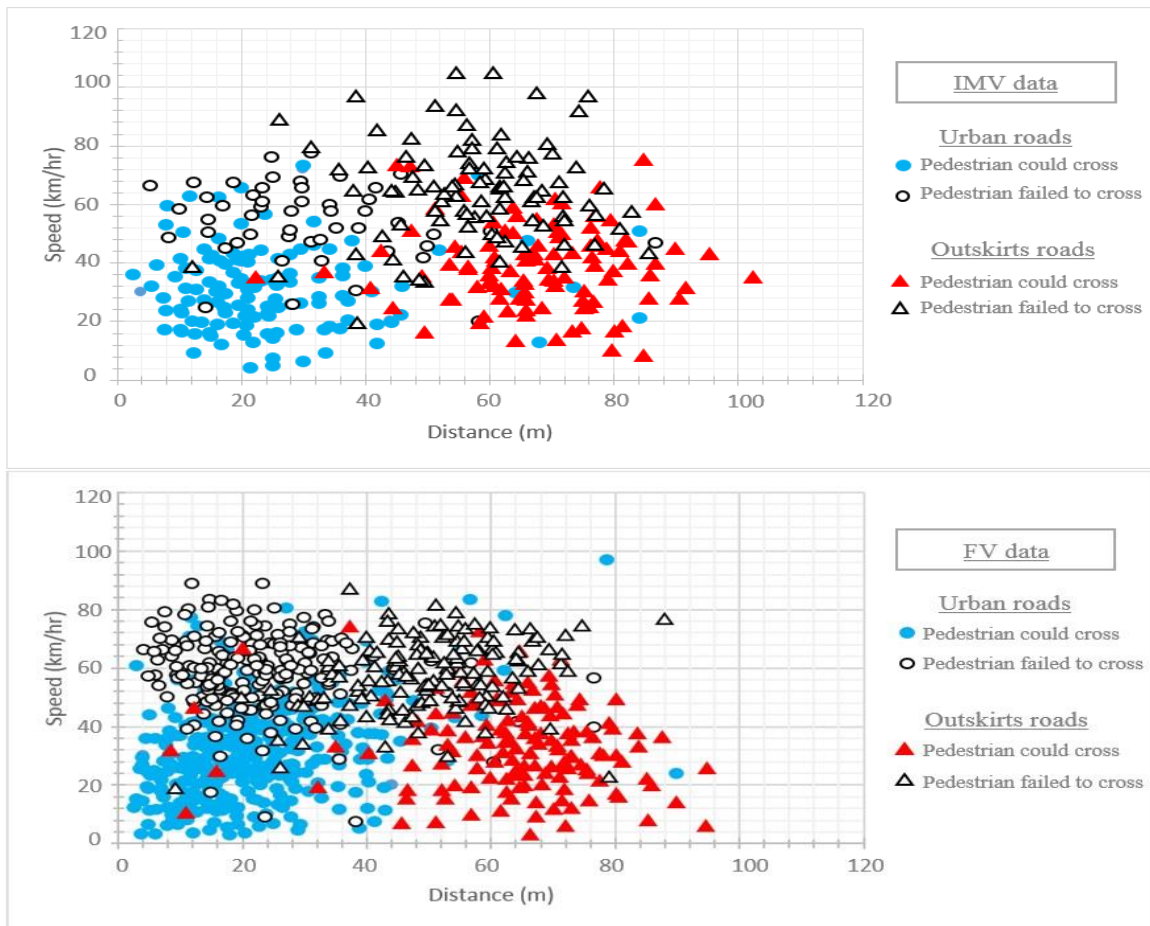


Fig. 3. Plots of pedestrian crossing condition (Up: IMV data; Down: FV data)

365
 366 Also, the probability of conditions in which pedestrians failed to cross (empty circle) increased as the
 367 car speeds up at a constant distance. The reasons for this include drivers' unwillingness to change their
 368 driving behavior, who are not willing to yield encountering pedestrians to allow them to cross the road.
 369 On the other hand, at the same speed, pedestrians on the outskirts roads were able to cross (filled red
 370 triangle) only when the distance between approaching car and them wasn't short. Simply put,
 371 pedestrians are more likely to cross the outskirts roads (filled red triangle) than on urban roads (filled
 372 blue circle) only if there is a longer distance between cars and pedestrians. Otherwise, at a constant
 373 distance and speed, pedestrians are more likely to cross on urban roads than on outskirts roads, which
 374 may be related to reasons such as road width, or more willingness of drivers to yield to pedestrian
 375 through various actions. Furthermore, on outskirts roads, pedestrians at short distances to cars were
 376 more likely to wait on the side of the road (empty triangle) to use a bigger gap size to cross. In this case
 377 (short distances), if the approaching vehicle speed is high, the pedestrians' decisions to wait on the side
 378 of the road increased.
 379

380 5. Conclusions and further research

381 New approaches such as FV and IMV data for better detailed analyzing road user behavior have
 382 compensated some of the limitations of using previous methods. Considering remarkable portion of
 383 pedestrian accidents occur on the outskirts areas connecting to the urban areas, the current study attempts
 384 to evaluate the safety of pedestrians crossing on both urban and outskirts areas through two different
 385 approaches. Data collection based on FV and IMV on the roads provided valuable information on the
 386 behavior of road users at different times and distance on roads. In the first step, data were analyzed to
 387 determine the pedestrian acceptance gap behaviour model through the linear regression model. FV and
 388 IMV data were analyzed for each of the urban and outskirts roads. The findings showed the similarity
 389 of results using these two different approaches. Based on the result of linear regression models, the drivers

390 and pedestrians exhibited relatively similar behavior on both outskirts and urban routes. Also, the results
391 showed that the variables of pedestrian gender, pedestrian willingness to cross without waiting on the
392 side of the roads, inclination of pedestrians to move in groups, and time required to cross the road can
393 be described as pedestrian behaviors that lead to different decisions when crossing the road. On the
394 other hand, the distance between the vehicle and the pedestrian as well as the speed of the approaching
395 vehicle are other influential factors in choosing an accepted gap by pedestrian. In second step, the
396 pedestrian crossing behavior model on both types of studied areas showed similar affects on variables
397 such as vehicle speed, distance of vehicle and pedestrian, pedestrian group size, and waiting time on
398 the both case study area. A similar behavior pattern was observed for pedestrians crossing in the
399 outskirts and urban roads. However, pedestrians on the outskirts roads perform more cautious behavior
400 to cross than in the urban ones. The modeling results also showed that the IMV and FV data are mostly
401 similar to one another and can be deduced that drivers' behavior in NDS studies was not different from
402 their normal behavior. In general, the findings of the present study are very important in two forms.
403 First, using FV and IMV approaches evaluated the issue of pedestrian safety from the perspectives of
404 drivers and pedestrians. Meanwhile, previous studies on road user behavior have always been addressed
405 from a separate perspective. Accordingly, the results of the present study expose a better understanding
406 of the nature of the interaction between the driver and pedestrians. Second, in this study the pedestrian
407 safety was examined on the outskirts areas, meanwhile, this subject had not been addressed through
408 driver or pedestrian behaviors on previous studies up to now. Based on the result, the behavior of road
409 users in both urban and outskirt areas indicates the same behavior in different situations, although slight
410 changes were performed in some of road users' decisions regarding different conditions.

411 It is anticipated that the results of such study will lead to a better understanding of the behavioral and
412 performance differences of drivers and pedestrians on the urban and outskirts routes when encountering
413 each other. In other words, considering that the use of advanced driver assistance systems (ADAS) is
414 increasing rapidly ([Acerra et al., 2019](#); [Bianchi Piccinini et al., 2014](#); [Dumitru et al., 2018](#); [Ghasemi et al., 2020](#); [Schnelle et al., 2018](#); [Varotto et al., 2018](#)), accurate identification of pedestrian behaviors in
415 different urban and outskirts routes can be a positive step in adapting ADAS to enhance pedestrian
416 safety in different areas, including outskirts areas. Taking into account the models presented, future
417 research in this field can focus on the role of Advanced Driver Assistance System (ADAS) to assess the
418 vehicle-pedestrian conflict probability. Studies with such perspectives are expected to lead to a more
419 comprehensive assessment of the role of ADAS in enhancing pedestrian safety on the roads.

421

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