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Innovative countermeasures for red light running prevention at signalized intersections: a driving simulator study

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ABSTRACT

The change interval, which includes the yellow and all-red times, plays a crucial role in the safety and operation of signalized intersections. During this interval, drivers not only need to decide to stop or go but also have to interact with drivers both in front and behind, trying to avoid conflicting decisions. Red light running and inconsistent stopping behavior may increase the risk for angular and rear-end crashes. This study aims to investigate the effect of different innovative countermeasures on red light running prevention and safe stopping behavior at signalized intersections. Five different conditions were tested inviting sixty-seven volunteers with a valid driving license. The conditions include a default traffic signal setting (control condition), flashing green signal setting (F-green), red LED ground lights integrated with a traffic signal (R-LED), yellow interval countdown variable message sign (C-VMS), and red light running detection camera warning gantry (RW-gantry). Drivers in each condition were exposed to two different situations based on the distance from the stop line. In the first situation, drivers were located in the indecision zone while in the second situation they were located in the likely stopping zone. A series of logistic regression analyses and linear mixed models were conducted to investigate the overall safety effects of the different countermeasures. The probability of red light running (RLR) was significantly reduced for R-LED in both analyses (i.e. in the total sample, and in the sample of crossed vehicles). Moreover, a clearly inconsistent stopping behavior was observed for the flashing green condition. Furthermore, a unit increase in speed (kph) at the onset of yellow interval significantly increases the probability of RLR by 5.3%.

The study showed that R-LED was the most effective solution for improving red light running prevention and encouraging a consistent stopping behavior at the intersection. In conclusion, the R-LED and the RW-gantry treatments are recommended as effective tools to improve safety at signalized intersections.

KEY WORDS: red light running; signalized intersection; dynamic ground lights; countdown signal; dilemma zone; variable message sign

1 **1. INTRODUCTION**

2 Signalized intersections are one of the most hazardous locations in road networks, accounting for a
3 substantial proportion of fatalities and injuries due to traffic crashes (Huang et al., 2014; PIARC, 2003).
4 According to research conducted by the University of California, 2.5 million crashes occurred at road
5 intersections in 2004 in the United States, and 20% of these intersection crashes were identified as signal-
6 related (Chang et al., 2007). Red light running (RLR), tailgating and inappropriate stopping at intersections
7 could be the main reasons behind signal-related crashes. Devlin et al. (2011) reported that 20% of the serious
8 injuries at signalized intersections in West Australia were related to rear-end crashes during 2004-2009. In
9 2013, RLR accounted for 697 fatalities and 127,000 injuries in the United States (McCarthy, 2015). Even
10 though, red light running detection camera (RLC) is installed at the majority of signalized intersections in
11 Qatar, a total of 23,152 vehicles in 2016, 26,771 in 2017 and 28,945 in 2018 were caught jumping red lights,
12 with drivers incurring a penalty of QAR 6,000 (Ministry of Development Planning and Statistics, 2017).
13 According to a survey based on a self-report assessment of risky driving behavior in the state of Qatar, 3%
14 of the drivers reported that it is completely acceptable to drive through a traffic signal light that just turned
15 red, while 4% oppose to the implementation of RLR cameras (Timmermans et al., 2019). Since RLR could
16 lead to severe crashes, there is a need to introduce innovative countermeasures to help improving safety at
17 intersections.

18 Yellow interval plays a crucial role in the safety and operation of signalized intersections. More than half of
19 the total number of crashes at signalized intersections occur due to the yellow interval (Yang et al., 2014).
20 Drivers usually find it difficult to make decisions at the onset of the yellow interval at signalized
21 intersections. Decisions to stop or cross an intersection typically have to be taken quickly at the onset of
22 yellow interval, and are subject to various physical (e.g. distance to the stop line), psychological (e.g.,
23 attention allocation, signal comprehension, etc.) and behavioral (e.g. current speed) conditions. Several
24 reasons could explain why drivers decide to cross a red light, such as distraction at the onset of yellow
25 interval (Li et al, 2014; Zhang et al., 2018), biased estimations of distance, time, and speed, aggressive

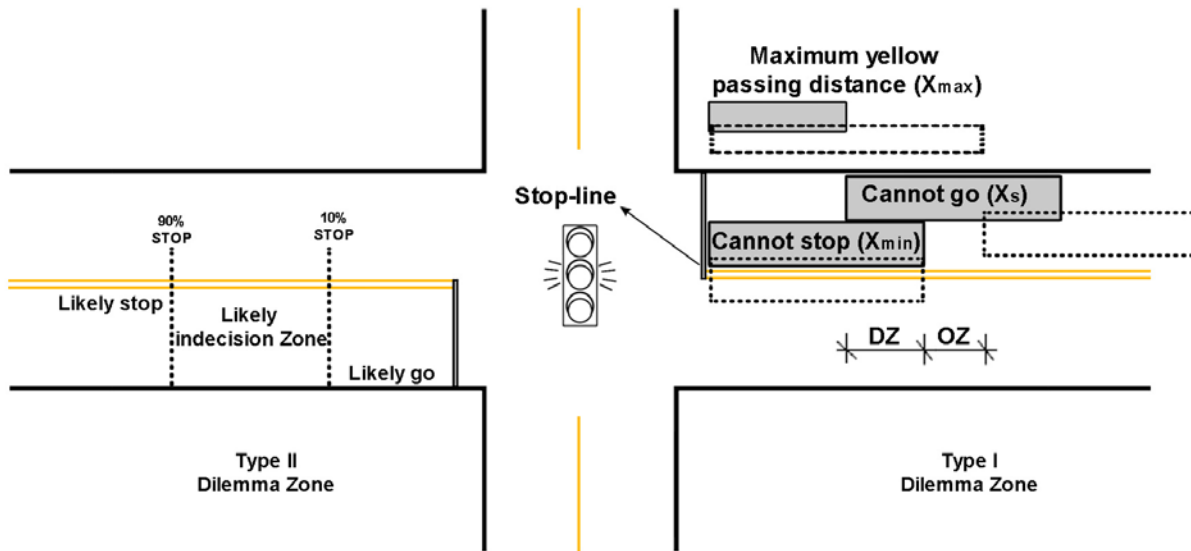


Figure 1. Illustration of the Type I and Type II Dilemma Zones

26 driving (Zhang et al., 2014) and alcohol impairment (Zhang et al., 2018). Drivers also have to interact with
 27 other drivers both in front and behind, trying to avoid conflicting decisions. Inconsistent stopping behavior
 28 may increase the risk for rear-end collisions, for instance, in case a front driver unnecessarily stops while
 29 still able to proceed and cross the intersection. Furthermore, improper signal design or operation may also
 30 promote RLR (Yang et al., 2014). The multitude of potential causative factors explain why RLR and unsafe
 31 stopping are considered as difficult to predict and why it remains a complex problem to be solved (FHWA,
 32 2014). Therefore, it is important to study driving behavior such as inconsistent stopping and red light running
 33 in sufficient details to improve safety at signalized intersections.

34 Sackman et al. (1977) identified two types of dilemma situations based on the vehicle position at the onset
 35 of yellow interval (see Figure 1). The first dilemma situation (type I) was developed back in the 1960s (Gazis
 36 et al., 1960). This situation is based on the maximum distance that can be traveled during the yellow interval
 37 (X_{max}), and the minimum stopping distance (X_{min}). When X_{max} is smaller than X_{min} , a dilemma zone
 38 (DZ) is formed, where drivers positioned in DZ at the onset of the yellow interval can neither stop safely,
 39 nor clear the intersection comfortably during the yellow interval. This can be considered as an indication of
 40 poor signal design (Gazis et al., 1960; Yang et al., 2014). Alternatively, when X_{max} is greater than X_{min} ,

41 an option zone (OZ) forms, where drivers could either stop or clear the intersection comfortably. Using
42 kinematic equations, X_{min} and X_{max} are defined in equations 1 & 2, respectively (Gazis et al., 1960).

$$43 \quad X_{min} = V_o(\delta_1) + \frac{V_o^2}{2a_1} \quad (1)$$

$$44 \quad X_{max} = V_o(t_y) + \frac{a_2(t_y - \delta_2)^2}{2} - L \quad (2)$$

45 In these equations, V_o is the vehicle's speed (m/s); a_1 and a_2 are the maximum deceleration and acceleration
46 rates (m/s²) for stopping and crossing, respectively; δ_1 and δ_2 are the driver's perception-reaction time (s)
47 for stopping and crossing, respectively; t_y is the duration (s) of yellow interval; L is the length (m) of the
48 vehicle. The equations indicate that the perimeters of a dilemma zone are not only depending on vehicle
49 traveling speeds and yellow interval times, but also depend on acceleration and deceleration (ACC/DEC)
50 rates and drivers' reaction times.

51 Due to the behavioral differences among drivers, it is not appropriate to use a standardized dilemma zone
52 as indicated by the type I dilemma situation. Therefore, it could better be positioned between flexible
53 boundaries (Chang et al., 2013) as specified in the type II dilemma situation introduced by Parsonson (1974).
54 The type II dilemma situation defines the approach area where 10% to 90% of drivers decide to stop during
55 the yellow interval (i.e. likely indecision zone), as shown in Figure 1 (Zegeer & Deen, 1978). In this study,
56 we used the concept of dilemma type II to identify the positions of drivers with respect to the stop line for
57 two situations. The first situation focuses on the common 80% of drivers who are driving in the indecision
58 zone, referring to drivers who could make wrong decisions of crossing the intersections and consequently
59 get red light tickets. Drivers in the first situation are also likely to face conflicting decisions with fellow
60 drivers, which can increase the risk of rear-end collisions. The second situation focuses on the rare 10% of
61 aggressive drivers, referring to drivers who decide to cross the intersection from likely stop zone while they
62 were supposed to stop, which can increase the likelihood of running a red light.

63 **2. STUDY OBJECTIVES**

64 The aim of this driving simulator study is to investigate the impact of innovative countermeasures on driving
65 behavior at signalized intersections. To this end, five conditions were compared to measure their
66 effectiveness on RLR prevention and promotion of safe stopping at signalized intersections. The five
67 countermeasures include a default signal setting (control condition) and four treated conditions. Two of the
68 four treated conditions are based on advanced warning: a condition with 3 seconds advance warning by
69 flashing green; and a condition with animation-based variable message sign fixed on a gantry about red light
70 running camera detection warning (RW-gantry). The two other treated conditions are based on innovative
71 countdown systems for yellow interval i.e. red LED ground lights integrated with a traffic signal (R-LED)
72 and yellow interval countdown variable message sign (C-VMS). To the best of our knowledge, the approach
73 of using red light units on the ground (R-LED), the countdown clock (C-VMS), and the RLC warning gantry
74 (RW-gantry) have never been tested in the past.

75 The first objective of this study is to analyze which of the tested conditions have a positive impact on RLR
76 prevention. The second objective is to investigate if the tested conditions reduce variations among drivers'
77 speed and ACC/DEC maneuvers.

78 **3. LITERATURE REVIEW**

79 **3.1 Driving simulator studies**

80 Previous research have proven that driving simulators are effective in the evaluation of human factors on
81 road safety (Fisher et al., 2011; Hussain et al., 2019a; Llopis-Castelló et al., 2016). For instance, driving
82 behavior in response to the road safety hazards can be studied in a safe environment using driving simulator
83 to eliminate the risks of testing in real-world scenarios (Ariën et al., 2013; Daniels et al., 2010; Nilsson,
84 1993).

85 More specifically related to the topic under study here, Newton et al. (1997) studied the efficacy of a
86 proposed signal indication with a sequence of green, followed by flashing yellow in conjunction with green,

87 followed by steady yellow, and then red. The proposed signal indication sequence was compared with the
88 most commonly used signal indication sequence i.e. green-yellow-red. Forty-one subjects (23 female)
89 participated in their driving simulator study ranging in age from 15-58 years. Although the authors observed
90 a substantial reduction in red light violations (RLVs), the new signal indication sequence prolonged the
91 indecision zone and increased potential conflicting decisions between vehicles approaching an intersection
92 at the same time. Therefore, the authors concluded that the proposed signal indication sequence would not
93 improve the safety at intersections. Similar conclusions were drawn in another simulator study investigating
94 the impact of advance warning flasher (AWFs) strategies on driving behavior at signalized intersections
95 (Smith & Harney, 2001). The authors concluded that AWFs often motivate stopping at the intersections, but
96 at the same time prolonged the indecision zone and resulted in a higher probability of risky decisions such
97 as unnecessary or early stopping.

98 Another study examined an in-vehicle stopping decision advisory system in a driving simulator in a sample
99 of 20 participants (Bar-Gera et al., 2013). The system provided an auditory and visual indication on
100 odometer with blinking red light indicator to warn drivers for the upcoming red light phase. The system was
101 designed taking into account the distance to the traffic light and remaining time of the green phase. The in-
102 vehicle advisory system reduced RLVs by 96% and the range of the indecision zone was shortened by 70%.

103 Abbas et al. (2014) devolved an Adaptive Randomized Incomplete Block Split-plot (ARIBS) design to
104 investigate drivers' behavior and factors influencing drivers' decision at the start of yellow interval in a
105 driving simulator research. Results from their study showed that the adaptive process of the design allows
106 the examination of drivers' stop or go decisions prediction mechanism with an accuracy of about 97.3%.

107 **3.2 Field observation studies**

108 The design of traffic signal timing can play an important role in improving safety at signalized intersections.
109 Several studies have analyzed the effect of lengthening the yellow interval on RLR. A cross-sectional study
110 on signalized intersections in three cities found that RLR occurrences were lower at the intersections where
111 yellow interval was longer (Bonneson & Son, 2003). Another study on four urban intersections and two

112 rural intersections found changes in RLVs one year after the yellow signal interval was increased by 1
113 second, demonstrating a reduction of almost 50% in RLVs (Bonneson & Zimmerman, 2004). Retting et al.
114 (2008) assessed RLV based on data collected before and after the yellow signal interval was increased by
115 approximately one second at several intersections. Based on logit models, they found that RLVs reduced by
116 36% when the duration of the yellow interval was increased.

117 Polanis (2002) evaluated the impact of size of the signal heads in a before-after study, where the signal heads
118 of 8" were replaced with 12" signal heads. The results showed an aggregate reduction of 49% in right angle
119 crashes at the intersections where the signal heads were upgraded.

120 Mahalel & Zaidel (1985) examined the indecision zone for warning of a three-second flashing green just
121 prior to the yellow interval. The authors concluded that although flashing green increases the probability of
122 stopping at intersections, it increases the length of the indecision zone and thereby the probability of rear-
123 end collisions as well.

124 Another study investigated the efficacy of a green signal countdown device (GSCD) in an observational
125 field study in Shanghai at two intersections (Ma et al., 2010). The authors reported a significant decrease
126 in RLVs with higher average traffic flow speeds at the intersection with GSCD installed, as compared to the
127 intersection without GSCD. The effectiveness of countdown timers on red light running has also been
128 evaluated in several other studies where potential decrease in RLVs was reported (Chiou & Chang, 2010;
129 Kidwai et al., 2005; Limanond et al., 2010; Lum & Halim, 2006; Rijavec et al., 2013).

130 Tydlacka et al. (2011) investigated the effectiveness of lighted stop bar system (LSBS) installed parallel to
131 the intersection stop line for three intersections in Houston, Texas. The LSBS containing red LED light units
132 was operated in a steady burn mode during the red interval while it was deactivated during the yellow and
133 green intervals. Results from the study showed a statistically significant reduction in RLR for two of the
134 three intersections. However, results from before-and-after study with a comparison site (non-treated) did
135 not show any notable change in the number of RLR violations. According to the authors, the LED light units

136 and connection wiring performed well under adverse weather conditions and in case of any failure of an
137 individual LED light unit, it could easily and quickly be replaced.

138 Numerous studies have investigated the impact of red light running detection cameras (RLCs) on road
139 crashes and frequency of RLVs at signalized intersections. A positive impact of RLCs on overall crashes,
140 and injury crashes has been reported (Aeron-Thomas & Hess, 2005; Retting & Kyrychenko, 2002). Hu et
141 al. (2011) estimated the effect of RLC enforcement on fatal crash rates. The study results showed that fatal
142 crashes significantly reduced by 35% after the installation of RLCs at the intersections. Additionally, several
143 studies evaluated the effectiveness of RLCs in reducing RLVs.

144 Polders et al. (2015) investigated the effect of integrated speed and red light cameras (SRLC) on the
145 occurrence of RLR and the risk of rear-end collisions, observing driving behavior in the real-world in
146 combination with a driving simulator study. Based on the observational field study, the authors reported a
147 reduction in crossing instances during red and yellow intervals after the installation of SRLC, with a decrease
148 in time headway. Furthermore, observations in the driving simulator study revealed that odds of rear-end
149 collisions increased by 6.42 times for the condition with SRLC, compared to the intersections without SRLC.
150 However, the odds of rear-end collisions increased only 4.01 times when an SRLC with an additional static
151 roadside posted sign was installed providing drivers with a warning about the RLC at the intersection.
152 Literature shows the importance of advance warning information, indicating it could motivate drivers to
153 reduce their speeds in advance, resulting in substantial reductions of RLR (Ahmed & Abdel-Aty, 2015;
154 McCart & Hu, 2014; Retting et al., 2008).

155 Gates et al. (2014) compared drivers' reaction and stopping behavior with and without the presence of RLC.
156 They found that drivers tended to react 5% (0.05 s) quicker to a yellow light change in the presence of RLC.
157 Moreover, at RLC intersections, the likelihood of a driver stopping increased by 2.4% in their study.

158 Besides the observational studies, researcher also presented V2I (Xie & Wang, 2018) and I2V (Grembek et
159 al., 2019) models based on the smart in-vehicle support systems to help drivers in making appropriate

160 decisions in the indecision zones at signalized intersections. However, the effectiveness of these systems is
161 not evaluated in the observational studies.

162 As indicated in the literature review, numerical countdown systems were tested for yellow and green
163 intervals in the past, however, this study provides new insights on yellow interval countdown by introducing
164 innovative treatments (C-VMS and R-LED). C-VMS counts down the yellow interval by graphic circular
165 clock while R-LED counts down the yellow interval by means of intelligent LED light units on the road
166 surface. Furthermore, the animation-based VMS displaying warning about RLC has never been tested in
167 previous research. The paper also address the effectiveness of flashing green in the State of Qatar, where
168 the driver population is very heterogeneous with many different cultural backgrounds (Soliman et al., 2018;
169 Timmermans et al., 2019). The implementation of 3 seconds flashing green interval before the onset of
170 yellow interval, as commonly applied at signalized intersections in the state of Qatar, is expected to increase
171 the probability of inconsistent decisions by drivers such as early and unnecessary stopping which may lead
172 to rear end collisions and reduction in the efficiency of signal control. In short, this study initiates the
173 assessment of the three new innovative countermeasures and tests their impact on RLR prevention and
174 promotion of safe stopping.

175 **4. METHODS**

176 **4.1 Driving simulator**

177 The driving simulator at Qatar University from the Qatar Transportation and Traffic Safety Center was
178 utilized to perform the experiment. The driving simulator has been validated for the external (i.e. actual
179 speed & speed perception) and subjective validity (Hussain et al., 2019a). The driving unit and three large
180 screens with geometric field of view of 135 degrees (5760 x 1080 pixels; 60 hertz refresh rate) are the
181 primary elements of the driving simulator (see Figure 2). The fixed-base cockpit was equipped with force-
182 feedback steering wheel with indicators, automatic gearbox, speedometer, and pedals resembling an
183 authentic Range Rover Evoque. The components are interfaced with STISIM Drive® 3 software and the
184 CalPot32 program. The integrated system offers high-speed graphics and sound processing. The simulator



Figure 2. Driving simulator: Qatar Transportation and Traffic Safety Centre, Qatar University.

185 is capable of collecting a wide range of data including speed, lateral/longitudinal acceleration,
 186 lateral/longitudinal position, number of accidents, number of red light tickets, number of speeding tickets,
 187 pedal inputs, reaction time, etc.

188 **4.2 Participants**

189 Sixty-seven volunteers were invited to participate in the driving experiment on the basis of a valid Qatari
 190 driving license type B (allows to drive all types of passenger cars). Sixty-two participants were driving cars
 191 with automatic transmission while five were driving manual transmission cars. The subjects included Qatar
 192 University community (students, faculty and staff) and drivers from outside the university (including some
 193 taxi drivers). With respect to the minimum requirements of the standard simulation sickness questionnaire
 194 (Kennedy et al., 1993), the drivers were told on forehand not to eat or drink (except water) two hours before
 195 the experiment. Despite the provided instructions, three participants were affected by simulation sickness
 196 and excluded from the results. Moreover, two participants were considered as outliers, resulting in a total
 197 study sample of 62 drivers. Descriptive statistics represent 47 male and 15 female drivers from 20 different
 198 countries of whom 30 Arabic and 32 non-Arabic drivers. The mean age was 28.93 years (SD: 7.3 years)
 199 ranging from 19 to 58 years, and the mean driving experience was 8.61 years, ranging from 1 to 30 years
 200 driving experience with a SD of 6.3 years. Furthermore, the proportion of participants' groups who drove
 201 less than 10,000 km, 10,000 – 20,000 km, and more than 20,000 km per annum were 24.2%, 30.6%, and
 202 45.2%, respectively.

203 **4.3 Implementation of the indecision zone**

204 In this study, each driver was confronted with two situations for all conditions on a road with a speed limit
205 of 80 kph. At the onset of the yellow interval, the distance between the vehicle and stop line was 80 m and
206 95 m in the first (S1) and second (S2) situation, respectively. According to Webster & Ellson (1965) the
207 boundaries of the indecision zone in type II dilemma zone are set at 56 m and 91 m prior to the stopping
208 line on a road with a speed limit of 80 kph where the assumption is that 10% and 90% of the drivers are
209 likely to stop, respectively. Therefore, in our study these situations were proposed aiming that S1 would be
210 an indecision zone for a substantial number of drivers (inside the boundaries) while S2 will be only for a
211 limited number of (more aggressive) drivers as it is situated prior to the outer boundary.

212 **4.4 Virtual road environment**

213 The experiment was designed as a 5x2 within-subject factorial design with as independent variables five
214 conditions: control condition, flashing green (F-green), red LED ground lights (R-LED), countdown
215 variable message sign (C-VMS) and red light camera warning gantry (RW-gantry), offered to participants
216 in two situations (s1: indecision zone; and s2: likely stopping zone). This means that every participant was
217 exposed to 10 study intersections, i.e. 5 conditions x 2 situations. The ten study intersections were randomly
218 alternated with filler pieces to create variations in the simulated scenarios. The filler pieces included 12
219 dummy intersections, pedestrians crossing the road, and lane changing and harsh braking maneuvers by
220 front vehicles. As all the study intersections were designed with changing signal phases, most of the dummy
221 intersections had a green signal phase to reduce the probability of learning effects.

222 Two driving scenarios were designed replicating the road layout and surrounding environment of the north
223 bound of the Corniche road in the city of Doha. In total, 11 intersections were incorporated in each driving
224 scenario. The 22 intersections (12 dummy and 10 study intersections) appeared in randomized order in both
225 driving scenarios. To maximize realism of the driving experience in the simulator (Bella, 2005, 2008), the
226 road environment was replicated as naturalistic and detailed as possible with inclusion of exact geometrical
227 alignment, cross-section furniture and roadside elements. To that end, we relied upon the real road

228 environment based on Google Earth images and video footages. SketchUp Pro (version 18.0.16975) was
229 used to design the roadside objects such as buildings, RLC, and streetlights. Furthermore, based on the real-
230 world observations, vehicle composition replicated in the simulation environment was constituted of 47.8%
231 sedan cars, 45.7% SUV, and 6.5% commercial vehicles (trucks, vans and buses).

232 **4.5 Design of scenarios**

233 The control condition was an untreated typical signalized intersection with the signal order of green-yellow-
234 red. The yellow interval was set at 4 seconds in accordance with the Qatar Traffic Control Manual QTCM
235 (Qatar Ministry of Transport and Communications MOTC, 2015). Important is that signal order in the
236 control condition was also used in the treatment conditions, except in the F-green where signal order was
237 changed into green-flashing green-yellow-red. The flashing green interval had a duration of 3 seconds with
238 three instances of green light blinking (i.e., at a frequency of 2HZ) (Mahalel & Zaidel, 1985). As described
239 in section 4.3, each condition was programmed for two situations with the distance to the stop line fixed in
240 each situation. However, this distance to the stop line was increased by 66 m (3 seconds x 80 kph \approx 66 m)
241 in both situations in the F-green condition since it included the 3 seconds warning advance to the yellow
242 interval. More precise details on the other three (treated) conditions follow below.

243 *4.5.1 Red LED ground lights (R-LED)*

244 In this condition, red LED ground lights were installed on the road surface, more specifically within
245 pavement marking strips indicating lane division and edge lines (see Figure 3). Activation of these lights
246 was aligned with functioning of the respective traffic light. Light units were operational during the yellow
247 interval and turned to red one by one in a sequential order towards the intersection. The main objective of
248 these moving light units is to provide direct and exact information to the drivers to stop if light units are
249 activated in front of them. Figure 3 shows the simulation views for two hypothetical vehicles V_{S1} & V_{S2} ,
250 positioned at S1 and S2 at the onset of yellow interval, respectively. V_{S1} is pictured in an indecision zone
251 (i.e. Situation 1) where there still is opportunity to lead and stay in front of the LED lights and where safely
252 crossing the intersection remains possible without a harsh acceleration being necessary. On the other hand,



Simulation view of Vs1



Simulation view of Vs2

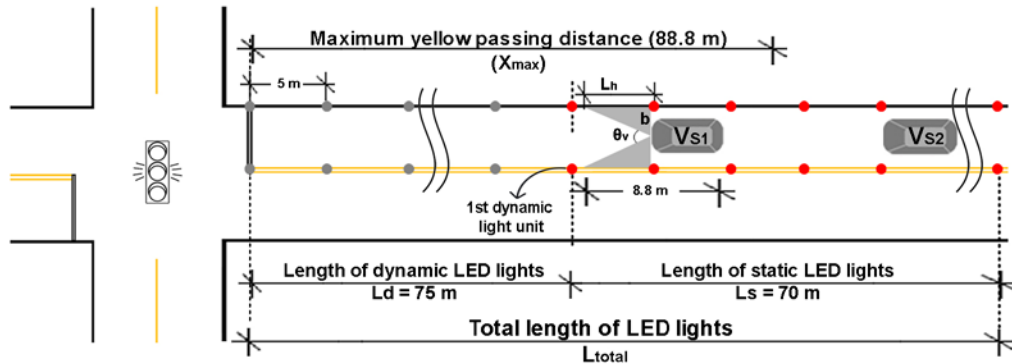


Figure 3. Red LED dynamic ground lights

253 V_{s2} is positioned in the stopping zone (i.e. Situation 2), where it is not possible anymore to cross the stop
 254 line safely without harsh acceleration. The gray dots in Figure 3 represent the lights yet to flash.

255 The whole stretch of the LED lights was divided into two parts, i.e. static and dynamic. A series of static
 256 LED lights were located in the stopping zone, turned red instantly and simultaneously at the onset of yellow
 257 interval, and remained red until the signal phase changed to green. These static lights indicate the area in
 258 which drivers cannot proceed smoothly to cross the intersection and thus they should prepare to stop.
 259 Meanwhile, the dynamic lights were activated one by one approaching to the intersection from a certain
 260 distance starting at the start of the yellow interval and remained red until the signal phase changed to green.
 261 The stretch of static lights (L_s) can be of any appropriate length (i.e. stopping zone), where around 10% of
 262 drivers still decide to cross (Zegeer & Deen, 1978). In our study, we implemented a 70 m long stretch for
 263 static red LED lights aiming to prevent those 10% of aggressive drivers from making unsafe crossing
 264 decisions. In addition, the length of the stretch with dynamic lights should be calculated based on the
 265 maximum yellow passing distance (X_{max}) considering the length of the vehicle (L_v) and the invisible
 266 distance (L_h) due to the front body of the subject vehicle as shown in Figure 3. The last dynamic red LED

267 shall reach to the stop line before the end of yellow interval allowing the subject vehicle to cross the distance
268 $(L_h + L_v)$ safely. Therefore, length of the stretch with dynamic lights can be calculated based on the
269 following equation:

$$270 \quad L_d = X_{max} - (L_h + L_v) \quad (3)$$

271 In this study, we assumed $(L_h + L_v) \approx 8.3$ m based on the average vehicle length (4 to 4.5 m) and 4 to 4.5
272 m of length hidden due to the front of the simulator car.

273 As the speed limit on the road was 80 kph, we calculated the speed of the dynamic LED lights as 75 kph.
274 This was done with a reason that if any driver located closer to the moving light units decide to cross the
275 intersection, he/she will have the opportunity to accelerate up to the speed limit and still cross the red LED
276 lights safely. Accordingly, with a speed of 75 kph and 4 seconds of yellow interval, length of the stretch
277 with dynamic lights can be estimated around 75 m [$L_d = 20.83 \text{ m/s} \times 4 \text{ s} - (8.3 \text{ m})$]. This means that with
278 speed of 75 kph the red lights will reach the intersection in 3.6 seconds ($75 \text{ m} \div 20.83 \text{ m/s}$) allowing 0.4
279 second to the last driver to cover the 8.3 m distance before the yellow interval ends. The intensity of color,
280 dimensions, and longitudinal spacing of the LED lights (i.e., 5 m) were adjusted based on suggestions given
281 by traffic experts involved in the pilot study.

282 *4.5.2 Countdown variable message sign (C-VMS)*

283 As discussed in the literature, the effectiveness of different countdown systems has been tested in simulated
284 environments as well as on the road. Most of those studies used numerical countdown to warn drivers of the
285 upcoming phase change. However, a visual representation of the countdown can be advantageous over the
286 numerical countdown approach, because hypothetically the remaining time interval can be perceived and
287 processed easier with the visual representation. For that reason, we designed a variable message sign (VMS),
288 visually displaying a circular clock that counts down (i.e. ‘ticks away’) the 4 seconds yellow interval (as
289 shown in Figure 4a). The complete circle is green during the green interval, and turns first from green to
290 yellow exactly at the onset of the yellow interval. Then, the bars in the countdown circle gradually change



a) Countdown VMS



b) Red light camera warning gantry

Figure 4. Simulation view of two conditions

291 from yellow to red one by one during the yellow interval. The C-VMS contains a flashing camera pictogram
 292 in the middle to inform drivers of the presence of RLC at the intersection. The C-VMS is installed alongside
 293 the traffic lights.

294 4.5.3 Red light camera warning gantry (RW-gantry)

295 An animated RLC warning VMS was designed and displayed on a gantry in this scenario. This VMS was
 296 aimed at informing drivers about the presence of a RLC at the next intersection and was installed 150 m
 297 prior to the intersection (as shown in Figure 4b). The animated picture shows a moving radar that flashes in
 298 the presence of a moving car leaving the intersection. On the right side of the panel, a text message
 299 mentioning 'camera ahead' is displayed in both Arabic and English.

300 4.6 Procedure

301 After obtaining the Qatar University's Institutional Review Board (QU-IRB) ethical approval, participants
 302 were recruited by putting up announcements on social web-portals with a link to the registration website
 303 (www.qatardrivingsimulator.com). Additional recruitment was done by sending official emails to the staff
 304 and students of Qatar University. The experimental session lasted for about one hour and was organized as
 305 follows:

- 306 1. Upon arrival to the simulation lab, each participant was asked to sign an informed consent form and
 307 to answer a pre-test questionnaire focusing on sociodemographic variables and driving experience.

- 308 2. To reduce bias due to a lack of knowledge, pictures of the countermeasures were presented to the
309 participants prior to the driving experiment. Subsequently, participants were asked if they
310 understood the functioning of the countermeasures. Additional information (about the functioning
311 of the countermeasures only) was provided when they did not understand, especially in case of the
312 innovative countermeasures. No information was shared about the study objectives or the purpose
313 of applying the countermeasures.
- 314 3. After that, participants were given a practice drive to get familiar with the driving simulator. The
315 drive consisted of an approximately 7 km long stretch of Doha Expressway. Drivers were instructed
316 to stop abruptly a few times to develop a more accurate estimation of the minimum stopping distance.
- 317 4. Before starting the experimental drives, participants received the following instruction: “Drive as
318 you would normally do, follow the traffic rules and continue driving until you are instructed to stop.
319 You have right of choice to quit the experiment anytime and for any reason”. Each participant
320 undertook two test drives with a short break in between.
- 321 5. After completing the test drives, each participant was asked to answer a post-test questionnaire. The
322 questionnaire included feedback/thoughts on the driving experience and on the driving simulator
323 itself.

324 **4.7 Data collection and analysis**

325 Data was collected for several driving parameters using STISIM Drive® Software. The collected data
326 included elapsed time (time passes from the beginning of the driving runs), longitudinal distance,
327 longitudinal speed, total longitudinal ACC/DEC, deceleration due to brake, and occurrence of red
328 light/speed violations. Data were collected for about one month of period with maximum six participants
329 were tested per weekday.

330 Two types of models were used to evaluate the performance of each scenario. For analyzing RLR prevention
331 logistic regression models were fitted, while for analyzing drivers’ stopping behavior linear mixed models
332 were used. In the logistic regression models, the extracted data were fitted to predict the probabilities of

333 RLR for each of the tested conditions. Logistic regression was opted since the dependent variable (RLR
334 event) was dichotomous. To estimate the probability of RLR, red light runners were labelled as “0” while
335 other vehicles were labelled as “1”. The dependent variables considered for the regression analyses were of
336 different types, i.e. categorical variables: situations (0=situation 1, 1=situation 2), conditions (0=Control,
337 1=F-green, 2=R-LED, 3=C-VMS, 4=RW-gantry), and ethnicity (0=Non-Arab, 1=Arab); and continuous
338 variables: speed at the onset of yellow interval in kph, age in years, and experience in years.

339 For analyzing drivers’ stopping behavior, repeated measures analysis of variance ANOVA is often applied
340 to analyze speed and acceleration differences between tested conditions (Ariën et al., 2013; Calvi, 2018;
341 Charlton et al., 2018; Hussain et al., 2018; Reinolsmann et al., 2019). However, in this study it was not
342 appropriate to conduct repeated measures ANOVA tests, as for different conditions some of the drivers
343 stopped at the intersection while other passed. According to Pinheiro and Bates (2000), linear mixed models
344 can be used on unbalanced data where fixed effect factors and random effect factors are considered.
345 Therefore, a series of linear mixed models for speed and acceleration/deceleration were conducted to analyze
346 drivers’ stopping behavior at intersections. For these models, data were extracted for a 500 m analysis
347 section (300 m before and 200 m after the stop line). For the speed models, the analysis section was divided
348 into 11 points with a constant 50 m spacing. Different from that, the analysis section was divided into 10
349 equal zones of 50 m, where SD of ACC/DEC in each 50 m zone were extracted. Standard deviations were
350 estimated for ACC/DEC in each zone to evaluate the homogeneity of acceleration/deceleration pattern
351 among drivers. The independent variables considered for these models include Situation (2), Condition (5),
352 and Point (11) / Zone (10).

353 Outlier analysis was performed separately for each of the 110 possible combinations (i.e. 11 points x 5
354 treatments x 2 situations). Any participant that was identified by SPSS as an extreme outlier (i.e. drove the
355 analysis section faster than 3 interquartile range from the group’s mean) in more than 15% of the total
356 combinations was eliminated from the analysis.

Table 1. Summary statistics of stop/go and RLR observations for each scenario(s)

Scenario	S1: indecision zone			S2: likely stopping zone		
	Stop	Go safely	RLR	Stop	Go safely	RLR
Control (n=62)	44(71.0%)	9(14.5%)	9(14.5%)	57(91.9%)	0	5(8.1%)
F-green (n=62)	56(90.3%)	4(6.5%)	2(3.2%)	59(95.2%)	0	3(4.8%)
R-LED (n=62)	42(67.8%)	17(27.4%)	3(4.8%)	61(98.4%)	0	1(1.6%)
C-VMS (n=62)	37(59.7%)	18(29.0%)	7(11.3%)	53(85.5%)	2(3.2%)	7(11.3%)
RW-gantry (n=62)	56(90.3%)	2(3.2%)	4(6.5%)	62(100%)	0	0

357 **5. RESULTS**

358 Table 1 presents the summary of statistics for stop/go and RLR for 62 subjects. Each subject was exposed
 359 to the 10 analysis intersections (5 conditions x 2 situations) where their decisions were recorded. As for
 360 Situation 1 (i.e. drivers are in the indecision zone), it can be read from the table that the frequency of RLR
 361 is lower for all the treatment conditions as compared to the control condition. The lowest number of RLR
 362 was observed for F-green (3.2%) followed by R-LED (4.8%) and RW-gantry (6.5%). The, highest number
 363 of stopping decisions (90.3%) was made in the F-green and the RW-gantry condition.

364 For situation 2 (i.e. drivers are in the likely stopping zone), all the drivers decided to stop when RW-gantry
 365 was installed before the intersection, while in the R-LED lights condition one driver decided to go and as a
 366 result, got a red light violation ticket. In the other conditions, the RLR occurrence was higher, as visible in
 367 Table 1. The highest number of crossing decisions (14.5%) was observed for C-VMS condition of which
 368 11.3% were RLR.

369 **5.1 Analysis of red light running prevention**

370 Two logistic regression models (Model 1&2) were developed to identify significant predictors for the
 371 probability of RLR (Table 2). Total sample (i.e. stopped and crossed vehicles) was considered for estimating
 372 Model 1. Compared to the control condition, probability of RLR reduced significantly for the R-LED ($\beta=-$
 373 1.453, $V2=5.986$, $p=0.014$) and F-green ($\beta=-1.272$, $V2=5.344$, $p=0.021$) treatments. The highest reduction
 374 was obtained for R-LED i.e. a 4.2 times ($1/\exp\beta$) lower probability of RLR as compared to the control
 375 condition. Furthermore, compared to the male drivers, the probability of red light running increased
 376 significantly for female drivers ($\beta=.829$, $V2=5.316$, $p=0.021$).

Table 2. Probability of red light running (significant p-values at 95% confidence level are indicated in bold)

Variables	Model 1: Sample of stopped and crossed vehicles					Model 2: Sample of crossed vehicles only				
	β	Standard error	Wald	df	Sig.	β	Standard error	Wald	df	Sig.
Constant	-4.735	1.994	5.638	1	.018	4.161	3.315	1.576	1	.209
Condition										
<i>F-green</i>	-1.272	.550	5.344	1	.021	.455	1.067	.182	1	.669
<i>R-LED</i>	-1.453	.594	5.986	1	.014	-1.753	.892	3.866	1	.049
<i>C-VMS</i>	-.030	.410	.005	1	.942	-1.006	.709	2.015	1	.156
<i>RW-gantry</i>	-.979	.601	2.654	1	.103	.292	1.059	.076	1	.783
Situation 2	-.582	.343	2.881	1	.090	4.152	1.122	13.706	1	.000
Gender (Female)	.829	.359	5.316	1	.021	.993	.629	2.495	1	.114
Ethnicity (Arab)	.106	.391	.074	1	.786	.449	.677	.441	1	.507
^a Age	-.021	.049	.187	1	.666	.126	.085	2.183	1	.140
^b Speed	.052	.019	7.477	1	.006	-.077	.033	5.301	1	.021
^a Experience	.009	.050	.031	1	.860	-.159	.089	3.153	1	.076

a: a unit increase in year, b: a unit increase in speed in kph at the onset of yellow interval

377 Model 2 was developed to predict the probability of RLR for crossed vehicles only, without considering
378 vehicles that had stopped. Ninety-three such cases could be identified and were subjected to analysis. Results
379 from the logistic regression models are presented in Table 2, Model 2. Of all the treatments, only the R-
380 LED condition showed a significant difference in comparison to the control condition: the probability of
381 RLR reduced about 6 times ($\beta=-1.753$, $V2=3.866$, $p=0.049$). Interestingly, although not significant, for the
382 F-green condition Model 2 yielded to opposite results compared to Model 1, indicating the probability of
383 RLR to increase. This was perhaps because most drivers decided to stop in this condition.

384 Moreover, speed at the onset of yellow interval had a direct relation with the likelihood of RLR in both
385 models. A unit increase in speed (kph) at the onset of yellow interval significantly increased the probability
386 of RLR by 5.3%.

387 5.2 Analysis of stopping behavior at intersections

388 5.2.1 Analysis of speed

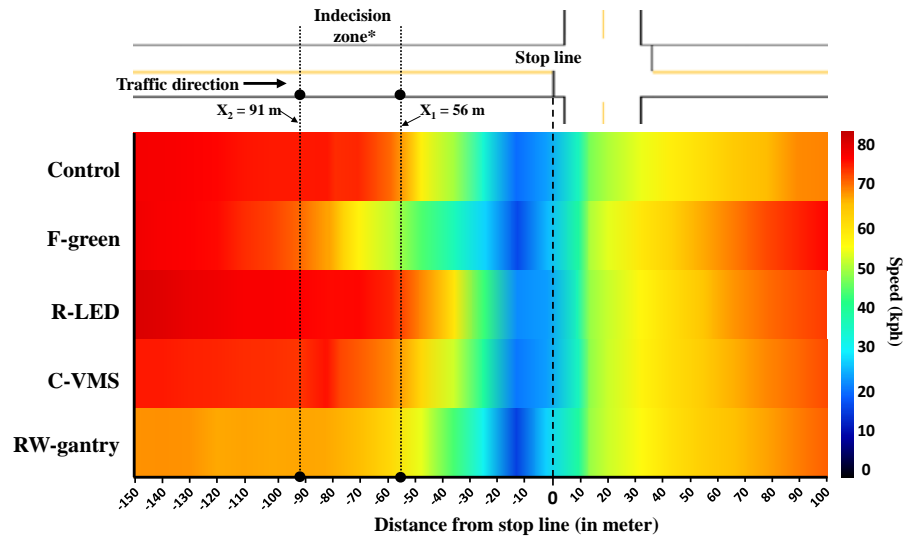
389 Table 3 presents the results for the linear mixed models regarding speed, for stopped and crossed vehicles
390 taken separately (Model 3 and 4, respectively). Results for both models show significant main effects for
391 the factors Point, and Condition. This implies that independent of any other factor, drivers' traveling speed
392 was significantly different across the 11 points, and for the five conditions. All the interaction effects were

Table 3. Linear mixed models – analyses of speed (significant p-values at 95% confidence level are indicated in bold)

	Model 3: Stopped vehicles			Model 4: Crossed vehicles		
	<i>F</i>	<i>dfs</i>	<i>Sig.</i>	<i>F</i>	<i>dfs</i>	<i>Sig.</i>
<i>Point</i>	1866.9	10, 5614	<.001	2.8	10, 882	.002
<i>Condition</i>	61.0	04, 5629	<.001	10.3	4, 921	<.001
<i>Situation</i>	<1	01, 5634	.455	2.6	1, 907	.106
<i>Point x Condition</i>	11.2	40, 5614	<.001	1.1	40, 882	.422
<i>Point x Situation</i>	7.7	10, 5614	<.001	<1	10, 882	.751
<i>Condition x Situation</i>	27.5	04, 5625	<.001	2.5	3, 917	.055
<i>Point x Condition x Situation</i>	3.2	40, 5614	<.001	1.2	30, 882	.184

393 insignificant for Model 4 (crossed vehicles) meaning that speed was not significantly influenced by a
 394 combined effect. For instance, the insignificant interaction effect of ‘Condition x Situation’ shows that speed
 395 was not varying significantly between conditions, for both situations (i.e. indecision zone (S1) and likely
 396 stopping zone (S2)) taken separately. On the contrary, all the interaction effects were significant in Model
 397 3 (vehicles that stopped). Two of these significant interactions are explained in more detail below:

- 398 – Condition x Situation: Independent of the factor ‘Point’, drivers’ traveling speed was significantly
 399 different between the conditions under both situations, separately.
- 400 – Condition x Situation x Point: The results from the post hoc of this interaction effect along 6 points
 401 (150 before and 100 m after the stop line) for S1 (indecision zone) are presented in Figure 5. The figure
 402 illustrates the mean speed profiles for all the conditions (using hue and saturation bars of different
 403 colors) at different points (-150 to 100 m to the stop line with 10 m interval). The dashed line represents
 404 location of the stop line while two dotted lines represent the boundaries of indecision zone (X_1 and X_2)
 405 on a road with speed limit of 80 kph (Webster & Ellson, 1965). It can be seen from the figure that there
 406 is a steep reduction in speed between X_1 (-56 m) and X_2 (-91 m) in the F-green scenario. Mean
 407 differences in speed ($d\mu = V@ X_1 - V@ X_2$), SD of differences in speed, and range of speed differences
 408 (see the table attached to Figure 5) were highest for the F-green condition, indicating high variations
 409 among drivers’ speed in the indecision zone. On the other hand, lowest values for $d\mu$, SD, and ranges
 410 were observed in the R-LED condition, indicating a more consistent speed reduction (stopping



<i>Differences in speed (kph)</i> ($X_1 - X_2$)	Condition				
	<i>Control</i>	<i>F-green</i>	<i>R-LED</i>	<i>C-VMS</i>	<i>RW-gantry</i>
<i>Mean of speed differences</i>	-8.11	-16.09	-1.76	-3.11	-2.13
<i>SD of speed differences</i>	3.8	8.1	2.4	3.7	3.2
<i>Range of speed differences</i>	4.8 to -13.4	-0.1 to -41.4	1.6 to -10.5	2.6 to -12.9	3.7 to -14.5

* Boundaries of indecision zone for 80 kph (Webster & Ellson, 1965)

Figure 5. Mean speed profiles of stopped vehicles for S1 (indecision zone)

Table 4. Linear mixed models – analyses of SD of acc/dec (significant p-values at 95% confidence level are indicated in bold)

	Model 5: Stopped vehicles			Model 6: Crossed vehicles		
	<i>F</i>	<i>dfs</i>	<i>Sig.</i>	<i>F</i>	<i>dfs</i>	<i>Sig.</i>
<i>Point</i>	2053.5	9, 5098	<.001	4.8	9, 792	<.001
<i>Condition</i>	3.4	4, 5148	.008	3.3	4, 666	.011
<i>Situation</i>	1.6	1, 5157	.199	<1	1, 823	.417
<i>Zone x Condition</i>	8.7	36, 5098	<.001	1.2	36, 792	.179
<i>Zone x Situation</i>	32.8	9, 5098	<.001	<1	9, 792	.552
<i>Condition x Situation</i>	1.4	4, 5140	.246	1.2	3, 756	.315
<i>Zone x Condition x Situation</i>	3.5	36, 5098	<.001	<1	27, 792	.587

411 behavior) among drivers. Furthermore, we found an early gradual speed reduction in the RW-gantry
 412 condition where a warning gantry was installed at 150 m prior to the intersection.

413 5.2.2 Analysis of ACC/DEC

414 Table 4 presents the results for the linear mixed models regarding SD of ACC/DEC, for vehicles that stopped
 415 and vehicles that crossed taken separately (Model 5 and 6, respectively). The results show significant main
 416 effects for the factor Zone and factor Condition in both models. Again, there was no significant interaction

417 effect observed for vehicles that crossed (Model 6). The interaction effect of Condition x Situation was not
418 significant in Model 5 which indicates that variations in ACC/DEC were not significantly different for
419 vehicles that stopped in the different conditions for both situations taken separately. However, the three-
420 way interaction effect (Condition x Situation x Point) was significant, which means if we analyze the results
421 on separate zones, the variations in ACC/DEC become significant between the conditions for the two

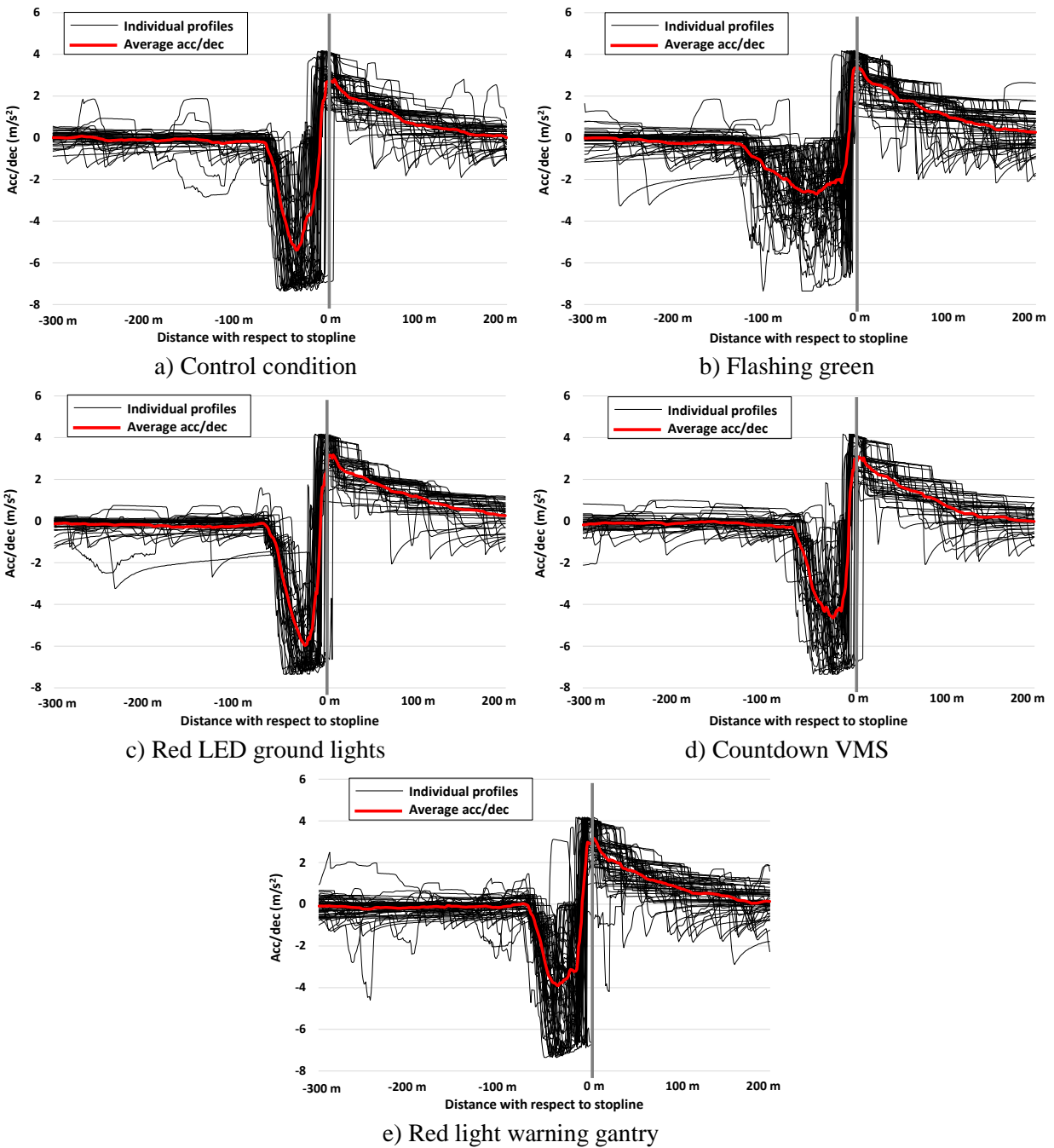


Figure 6. ACC/DEC profiles of stopped vehicles for S1 (indecision zone)

422 situations taken separately. Consequently, Figure 6 (a-e) presents the ACC/DEC profiles of drivers who
423 stopped at intersections in the first situation (indecision zone) for all conditions separately. Except for the
424 F-green condition, most drivers started to decelerate from between 100 - 70 m, with a comparable consistent
425 trend in all the other conditions. However, in the F-green condition the consistency in drivers' stopping
426 behavior was lower. Some drivers started to decelerate in advance probably due to the advance flashing
427 green warning, while others continued for a while and started to decelerate in their approach to the
428 intersection.

429 **5.3 Subjective evaluation of the driving simulator**

430 The results for subjective evaluation of the experience in driving simulator and the performance of the
431 different components of the driving simulator are presented in Table 5. A 5-point Likert scale was used to
432 measure all the items. The results indicate that drivers were comfortable with the driving simulator (Mean:
433 3.51) and gave highest rating for the overall general experience of using the driving simulator (Mean: 4.21).
434 Regarding the performance comparison of the components of the driving simulator, all examined measures
435 were higher than the middle value (i.e. 3). The highest rating was given to the steering wheel (Mean: 4.03)
436 followed by the overall performance (Mean: 3.61). The lowest rating was given for the engine sound
437 produced by the sound system of the simulator.

438 **6. DISCUSSION**

439 This study aims at investigating the effectiveness of different countermeasures to prevent RLR violations
440 and help drivers to stop safely and consistently at signalized intersections. RLR prevention was investigated
441 by means of two analyses, i.e. one for the overall sample (i.e. both the stopped and crossed vehicles), and
442 another for only vehicles that had crossed the intersection. The second analysis was done to reveal if the
443 tested conditions could assist drivers in making correct decision of crossing the intersections or not. Stopping
444 at the signalized intersections indeed completely eliminates the hazard related to running a red signal light.
445 However, inconsistency or differences among drivers' stopping behavior who are approaching the

Table 5. Post-test Questionnaire for the evaluation of the driving simulator (1 = not effective, 5 = highly effective)

		No. of responses for each level of rating					Mean Ratings	SD Ratings	Sample Size
		1	2	3	4	5			
General experience of using the simulator	<i>Overall</i>	1	3	2	31	24	4.21	0.8	61
	<i>Comfort</i>	7	7	11	20	16	3.51	1.3	61
Performance comparison	<i>Overall</i>	3	7	12	28	11	3.61	1.1	61
	<i>Steering wheel</i>	1	7	4	26	23	4.03	1.0	61
	<i>Accelerator</i>	3	9	15	25	9	3.46	1.1	61
	<i>Brake</i>	6	17	7	22	9	3.18	1.3	61
	<i>Sound of engine</i>	9	8	12	23	9	3.25	1.3	61

446 intersections at the same time could increase conflicting decisions and the risk of rear-end collisions. In this
 447 study, inconsistency among drivers' stopping behavior is assessed by variations in drivers' speed and
 448 ACC/DEC maneuvers.

449 In this regard, the R-LED condition was the best treatment for lowering the probability of RLR at
 450 intersections for both samples. Furthermore, the lowest variations in drivers' speed and ACC/DEC were
 451 observed for the R-LED, which indicate that the R-LED could assist drivers in making consistent stopping
 452 behavior while approaching the intersections. According to Wu et al., (2013), the lower variations in speed
 453 and ACC/DEC could shorten the indecision zone. The advantage of R-LED indicted by both results could
 454 be explained giving that the dynamic series of red light units installed on the ground provide direct
 455 information about the upcoming red phase to assist drivers in making safer stop/go decisions, based on their
 456 location at the intersections. The red ground lights provided by R-LED could reduce cognitive load for
 457 judgment about the stop/go decision, as they create a visual experience that allows drivers to more accurately
 458 estimate space and time, which is also in accordance with the subjects' feedback derived from the post-test
 459 questionnaire. On the other hand, when drivers are confronted with the default yellow signal light, they have
 460 to take into account the distance, speed and the remaining time in the yellow interval before deciding to stop
 461 or go (Gazis et al., 1960). It is important to mention that the functioning of these lights can be modified in
 462 different ways to increase the safety or efficiency at signalized intersections considering local driver
 463 behavior. For instance, a shorter stretch of R-LED with more slowly moving lights could increase the

464 efficiency, as those drivers approaching the intersections will have the chance to accelerate and be in front
465 of the light units.

466 Although F-green was effective in reducing RLR for both crossed and stopped vehicles, opposite results
467 were obtained when only analyzing vehicles that had crossed. Besides, results showed that the F-green
468 treatment stimulated drivers to stop when they were in the indecision zone. This could be explained by the
469 advance warning in the F-green condition motivating most drivers to stop (Mahalel & Zaidel, 1985; Smith
470 & Harney, 2001) and drivers interpret the flashing signal as a call for immediate action to stop rather than
471 as an advance warning for the yellow interval (Factor et al., 2012). Furthermore, observation of the highest
472 speed differences, the highest range of differences in speed across participants and the large variations in
473 the acceleration profiles indicate more inconsistent and unsafe stopping behavior in the F-green condition.
474 This incorrect understanding of the function of the flashing green prolongs the length of the indecision zone,
475 and consequently increases suboptimal decisions among traffic approaching the intersections at the same
476 time (Mahalel & Zaidel, 1985). Moreover, research shows that there is increased potential for conflicting
477 decisions if two consecutive drivers are approaching the intersection with a flashing green signal (Factor et
478 al., 2012; Newton et al., 1997).

479 The countdown system (C-VMS) motivated drivers situated in the stopping zone to cross the intersection
480 which yielded to a large number of red light tickets (11.3%). The results are in line with the previous studies
481 showing that countdown systems could lead more vehicles to cross the intersections (Long et al., 2013; Ma
482 et al., 2010) and more RLR (Chiou et al., 2010; Long et al., 2013). This might be because the remaining
483 time in the countdown systems makes drivers impulsive and aggressive to accelerate and cross the
484 intersection, which results in higher number of RLR. Moreover, an early and smooth mean speed reduction
485 was observed for the RW-gantry treatment indicating the importance of warning messages in reducing risk
486 of rear-end collisions (Høye, 2013; Polders et al., 2015).

487 The results from several previous studies show that female drivers are more likely to be less aggressive than
488 male drivers (Anderson et al., 1999; Lewis et al., 2012, 2013). In contrast, our the results from logistic

489 regression model (Model 1) shows that compared to a male driver, the probability of red light running is
490 higher in case of a female driver approaching the intersection with yellow interval. The results are in line
491 with a study conducted in the state of Qatar showing that female drivers are more aggressive and drove
492 faster than male drivers (Hussain et al., 2019b). Finally, traveling speed is an important parameter
493 contributing to drivers' decision on stop/go at signalized intersection (Gazis et al., 1960). In accordance,
494 this study showed that the higher the speed at the onset of yellow interval is, the higher the probability of
495 RLR will be.

496 **7. LIMITATIONS AND FUTURE RESEARCH**

497 There are several limitations in this study that need to be addressed. The experiment was conducted using a
498 fixed-base driving simulator, which might have reduced the level of realism. However, results from the
499 subjective evaluation of the quality and performance of the driving simulator are in line with a validation
500 study (Hussain et al., 2019a) indicating that the settings of the simulator are comparable with the settings
501 offered by a real car. The sample used in this study was skewed more towards the younger age group with
502 no participants of more than 60 years old age. Forty subjects participated in the study were of age 30 or less
503 including 18 university students. The percentage of Qatar population with more than 60 years old is less
504 than 2.4% in 2018 as reported by the Planning and Statistics Authority Qatar (2019). Most of these elderly
505 (>60 years old) do not drive but rather have chauffeurs for their daily activities. Moreover, it is important to
506 note that the recruited sample of drivers might not be representative of the driving population Qatar. In spite
507 of that, the results of this study may vary for a sample with higher proportion of the old-aged drivers or
508 different representative sample. Traffic driving in the same direction was triggered in a way to isolate the
509 participants' car by not driving in front or parallel while the car was approaching the intersection. This was
510 done to overcome the effect of conformity on stop/go behavior (Nordfjærn & Şimşekoğlu, 2014). Finally
511 yet importantly, due to the length of the driving scenarios, all the participants were tested twice for each
512 condition (once for each situation) during one session of driving. The results may vary for testing the drivers
513 in multiple sessions for the same driving scenarios.

514 In this study, C-VMS and R-gantry were installed at a single location i.e. alongside the traffic lights and 150
515 m before the intersection, respectively. The effectiveness of these treatments can be tested for installing
516 them at different locations, such as installing C-VMS system at the start of dilemma zone. Further research
517 and real-world implementation will allow practitioners to find clearer and long term effect of the proposed
518 treatments such as the R-LED on driving behavior. Another feasible option for real-world implementation
519 of R-LED is to alternate the ground lights with the roadside light poles. However, in this case it is needed
520 to calculate the hidden length concealed due to the peripheral visual field instead of the hidden length
521 concealed by the front of the vehicle (see Figure 3), which can estimated as follows;

522

$$523 \quad L_h = (b) \tan \frac{180 - \theta_v}{2} \quad (4)$$

524 where θ_v is the peripheral visual field angle, which is dependent on the driving speed; and b is the half width
525 of one direction road.

526 **8. CONCLUSION**

527 In this study, different types of countermeasures have been proposed and compared to evaluate the safety at
528 signalized intersections. Two logistic regression models were used as an indicator for RLR prevention while
529 variations in speed differences were measured as an indicator for (in)consistent stopping behavior at the
530 intersections. The results showed that none of the tested conditions was effective in reducing the probability
531 of RLR in both models except R-LED treatment for which the probability of RLR reduced significantly in
532 both models (i.e. Model 1: $\beta = -1.453$, $V2 = 5.986$, $p = 0.014$; Model 2: $\beta = -1.753$, $V2 = 3.866$, $p = 0.049$).
533 Furthermore, the lowest variations in speed differences (mean of -1.76 kph; SD of 2.4 kph; range varies
534 from 1.6 to -10.5 kph) were also observed for R-LED, showing the most consistent stopping behavior at
535 signalized intersections among all the tested conditions. Based on these outcomes we conclude that R-LED
536 can be an effective treatment to improve safety at signalized intersections. According to FHWA (2009), such

537 kind of in-roadway LED light units have already been used by several agencies with different purposes, and
538 an LED unit costs approximately \$50 including labor and material costs.

539 Although, flashing green is the default traffic signal setting in Qatar, highest variations in acc/dec profiles
540 and speed differences (mean of -16.09 kph; SD of 8.1 kph; range varies from -0.1 to -41.4 kph) were
541 observed in this condition. The results clearly indicate that F-green could prolong the length of indecision
542 zone and hence a potential for conflicting decisions among traffic approaching the intersection (such as
543 unnecessary or early stopping). Therefore, it is important for policy makers to consider the results from this
544 study while making decisions about traffic signal settings at signalized intersections.

545 Furthermore, results showed that the higher the speed at the onset of yellow interval, the higher would be
546 the probability of RLR. A unit increase in speed (kph) at the onset of yellow interval would significantly
547 increase the probability of RLR by 5.3%. Therefore, the stopping zone and start of the indecision zone can
548 be supported by appropriate speed calming countermeasures such as perceptual countermeasures, road
549 markings, and rumble strips to motivate drivers lowering their traveling speeds at the onset of yellow interval.

550

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