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9 **DRIVER WORKLOAD ANALYSIS USING AN INTERACTIVE 3D**  
10 **DRIVING SIMULATOR**  
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**46 ABSTRACT**

47

48 Diagnosing the causes of road accidents and the development of effective countermeasures to  
49 reduce accident rates is of key importance in road safety. Human error is one of the principal  
50 influencing factors that leads to road accidents, and is attributed to increased mental workload  
51 induced by distractions. Workload, however, is characterized by intrinsic properties that are  
52 difficult to observe. Hence, phenotype behaviours, such as lane deviations, could act as good  
53 predictors of driver workload. Driving simulators emerged as a promising technology for the  
54 analysis of driving conditions and road users' behaviour in an attempt to tackle the problem of  
55 road accidents. The work presented herein demonstrates the design and development of a driving  
56 simulator, using a 3D game engine that aims to contribute towards evaluating black spots in road  
57 networks by promoting rapid design of realistic models and facilitating the specification of test  
58 scenarios. The developed simulator was employed to evaluate the impact of distractors on  
59 driving behaviours of local road users for a chosen black spot in Limassol-Cyprus. Data  
60 collected from the experiments are analyzed, and the main findings in regards to the above  
61 relationships are presented and discussed.

62

63 **KEYWORDS:** Driving Simulator, Workload, Distractions.

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65

**66 INTRODUCTION**

67

68 Road accidents have become a daily hazard in Europe and worldwide (Konstantopoulos et al.,  
69 2010). According to Eksler et al. (2008), around 1.2 million fatalities and more than 50 million  
70 injuries occur in roads worldwide every year. Given the current trends, the accident fatalities are  
71 projected to become the second most common cause of death in 2020 if no drastic measures are  
72 taken. To that end, EU set the goal to reduce road fatalities and injuries by 50% by 2020. In  
73 addition to fatalities, traffic accidents result in high economic losses due to traffic congestion  
74 which in turn leads to a wide variety of adverse consequences such as, traffic delays, supply  
75 chain interruptions, travel time unreliability, increased noise pollution, as well as deterioration of  
76 air quality.

77 Road accidents are caused by many factors, and the problem is approach from different  
78 perspectives. Eksler et al. (2008) argue that accidents are influenced by demographic,  
79 infrastructural and political factors. At the other end of the spectrum human factors experts  
80 associate accidents with human error. Human error is defined as the human activity or absence of  
81 activity that leads to incorrect system behaviour (Hollnagel et al., 2004). It may occur due to  
82 human beings' physical, perceptual and cognitive limitations (Montella et al., 2010) and is  
83 directly related to visual attention (Konstantopoulos et al., 2010) and workload (Gregoriades et  
84 al., 2010). The analysis of accidents accounting for human error can be carried out from two  
85 perspectives: the designers' and the users'. The former addresses the system designing flaws that  
86 hinder human activity due to usability problems whereas the latter analyzes internal cognitive  
87 processes of human operators to identify decision making bottlenecks caused by reduced  
88 situational awareness (Endsley, 1995) due to increased workload. Mental workload constitutes  
89 an important influencing factor of road accidents and is directly related to human performance  
90 (Gregoriades & Sutcliffe, 2007) and attention. According to Bailey (1996), workload is defined  
91 as the demand placed upon people which could be a behavioural response to events,

92 communication and interactions among humans or between humans and technology, or humans  
93 and the environment. High levels of workload degrade the driver's concentration, information  
94 processing and decision making, leading to increased errors, which might have catastrophic  
95 effects (Williams, 1988; Endlsey, 1995; Norman, 1988).

96 The focus of this study is on driver workload analysis of local drivers in Cyprus using a driving  
97 simulator that was designed for this purpose and has the capability of easily being reconfigured.  
98 The analysis is based on phenotype driver behaviour data collected through experiments in  
99 controlled settings. The workload assessment method used is task-performance based. In  
100 particular, this investigation focuses on the impact of different types of driver distractions, such  
101 as advertisements, on the primary task (driving). The advertisements are placed along a critical  
102 point of a road network in Limassol Cyprus. Two types of traffic control (i.e. with and without  
103 traffic lights) are examined under dangerous traffic conditions.

104 The use of a driving simulator in studies, like this, is inevitable, firstly due to ethical reasons and  
105 secondly, since controlling infrastructural parameters in the real world requires huge investment  
106 of time and money (Davenne et al., 2012) which is usually prohibitive. Moreover, ruling out  
107 confounding effects to examine the influence of control measures on workload is very difficult in  
108 field experiments. Driving simulators provide the researcher with a powerful tool to test driving  
109 behaviour under controlled settings. Apart from the usually high cost of the simulator,  
110 outsourcing of experiments to analyze driving behavior using native users is difficult, if not  
111 impossible in some cases, due to the large number of subjects needed for reliable results. On the  
112 other hand, low cost driving simulators do not provide a sufficient level of realism to analyse  
113 human factors. Unrealistic conditions may affect the driving behaviour which effectively could  
114 influence the validity of the experimental study. The method proposed herein demonstrates the  
115 design of a driving simulator that exploits 3D modeling tools in a module-based approach to  
116 promote realism and interactive 3D representation of road networks. The approach simplifies the  
117 process of implementing 3D road infrastructure models through the utilization of reusable  
118 modules. This simplifies the process of designing/modifying the simulation model by reusing  
119 model constructs in a plug and play fashion. This enables the analyst to easily design a range of  
120 experimental conditions to evaluate assumptions and hypotheses from different perspectives.

121 The paper is organized as follows: related work is firstly reviewed, followed by the presentation  
122 of the driving simulator design along with the design of the experiment. The paper concludes  
123 with the analysis of the results and a discussion of their implications.

## 125 **RELATED WORK**

127 Road accidents are usually attributed to human error (Gregoriades et al., 2010). Humans have  
128 limited information processing abilities and must rely on three fallible mental functions:  
129 perception, attention and memory (Fuller, 2002). Drivers commit errors because a situation  
130 exceeds the limitations of one or more of these three functions. At the same time, the bandwidth  
131 of their information processing capabilities declines with age. Human factors in driving are  
132 usually distinguished in those addressing driving behaviour and driving task. Task-related factors  
133 relate to information perception, analysis, decision making and action at the operational level  
134 while behaviour-related factors concern the tactical and strategic level. The driving task requires  
135 information processing and motor skills, which improve with time while, driving behaviour  
136 refers to the style of driving that is not necessarily time-dependent. Driving behaviour is  
137 manifested by phenotype behaviours such as speed, gap acceptance and lane changing while the

138 primary driving task is supported or hindered by secondary tasks induced by visual, tactile and  
139 auditory information.

140 Drivers' information processing needs increase with distractions from endogenous and/or  
141 exogenous parameters. Exogenous parameters relate to the environment, the vehicle, the road  
142 infrastructure and the traffic conditions, whilst endogenous parameters include, but are not  
143 limited to, passenger distractions, noise, mobile phones, and using in-vehicle information  
144 systems (Young et al., 2003). According to Miller (1996), people can process 7 (plus or minus 2)  
145 discrete information chunks at a given point in time. This approximates the boundary of our  
146 cognitive capacity in terms of memory. Therefore, increased demand for cognitive resources may  
147 result in drivers failing to attend to critical information on the road. Humans, as information  
148 processing systems, have number of information flow channels (visual, auditory, tactile)  
149 processing various information sources (e.g. a navigation system display, the forward view  
150 through the windscreen) of varied bandwidths (e.g. high-density traffic will require a higher  
151 sampling rate than low-density traffic). Our cognitive capacity is limited, and in return there is an  
152 upper threshold to the amount of information we can process per second and channel  
153 (Gregoriades et al 2007, 2010). Therefore, we tend to share our attention among a few  
154 information sources. An overloaded driver is less likely to deal effectively with an unexpected  
155 event. Fuller (2005 & 2002) also expresses accident risk as a function of the driver's cognitive  
156 resources and task-demand in the driver-road system.

157 The link between overloading and driver distraction is established within the existing literature,  
158 (e.g. Dingus et al., 2006; Jamson et al, 2004; Gregoriades et al., 2010). Distraction in driving is a  
159 frequently reported cause of road accidents. According to Dingus et al. (2006), distractions  
160 contribute to 78% of accidents and to 65% near-crashes. Distractions can emerge from outside or  
161 inside the car. While, much research has investigated in-car distractions (Jamson et al., 2004),  
162 relatively little work has been reported for exogenous distractions emerging from outside the car  
163 (Young et al., 2003). Roadside advertising billboards, are one of the many distractions, which  
164 could pose a crucial risk for road safety. The evidence that accident risk increases with roadside  
165 advertising is increasing, with estimates making advertisements responsible for up to 10% of all  
166 road traffic accidents. According to Young et al. (2009) roadside advertising adversely affects  
167 lateral control which is one manifestation of driver overloading. In the same vein, visual search  
168 reaction times increase with distractors (Holohan, 1978). Hence, distracted drivers take longer to  
169 react to stimulus. This consequently yields support to the claim that drivers' visual attention is  
170 attracted by advertisements (Horberry et al., 2004). This increases significantly the risk of  
171 accident in the case that the driver's visual workload is already compromised. As a result, the  
172 driver may fail to sufficiently attend to the needs of the primary task (i.e. driving) and hence  
173 make errors that could lead to a hazard. Driver phenotype behaviours associated with workload  
174 include, but are not limited to, the following: lane position deviation, number of lane departures,  
175 lane departure durations and speed deviation. Hence, monitoring these phenotypes can give a  
176 good estimation of driver workload.

177 While there is evidence that roadside advertisements potentially affect driver attention, it is  
178 difficult to conclude about the specific risks. Results from field studies and controlled  
179 experiments seem to conflict. Hence, more empirical research is required. This work aims to  
180 shed more light into this debate through the empirical analysis of the effect of distractions on  
181 drivers' workload and the consequences on accident risk, using native road users. The use of  
182 Cypriot drivers tailors the analysis on native driving behaviours and helps to investigate how  
183 these could vary in different contexts of domestic relevance.

## 184 **DESIGNING THE DRIVING SIMULATOR**

185

186 The first part of this work involved the design and development of a modular driving simulator  
187 that would enable the analysis of traffic conditions and driving behaviours of native users. Given  
188 these needs it was imperative that the method for designing the simulator and its inherent models  
189 should have been generic, utilizing libraries of components representing assets that make up the  
190 driving conditions and infrastructure in Cyprus. Therefore, the design and development of the  
191 driving simulator utilized software packages to enhance the quality and reduce implementation  
192 time. The software applications that were used include: Unity game engine (Unity  
193 Technologies), City Engine (ESRI), Autodesk Maya (Autodesk inc.), and Tree[d] (Freclé).  
194 Unity is a 3D game engine software application that enables the development of 3D computer  
195 games and interactive virtual environments. The software comes with a combined functionality  
196 that enables changing the graphical environment either by using an editor or by manipulating it  
197 directly in graphical view. It also provides the designer with the ability to define behaviours  
198 through a powerful scripting language. In our case JavaScript scripting was used to define the  
199 interactivity between the user and the assets of the simulator. The main steps followed for the  
200 design of the simulation environment are listed below:

- 201 1. Model objects and design the non-interactive animation objects (static and moving):
  - 202 (a) Create road network infrastructure and surrounding buildings using ESRI CityEngine;
  - 203 (b) Create asset models, such as vehicles, traffic signals, advertising billboard etc, in  
204 Autodesk Maya;
  - 205 (c) Model vegetation using tree[d]
  - 206 (d) Composition of the scene using the above mentioned 3D models in Unity;
- 207 2. Enhance realism by simulating lighting conditions, shadows, and sound effects within  
208 Unity;
- 209 3. Add interactivity within Unity through scripting

210 The first challenge was the modeling of the road network. To achieve this we extracted a section  
211 of the Limassol road network from OpenStreetMap by cropping the area of interest using a  
212 polygon on the map. Subsequently, the extracted XML file was imported into the CityEngine, a  
213 3D procedural modeling software application, specializing in the generation of 3D urban  
214 environments through the manipulation of objects and existing GIS data. In our case, City  
215 Engine was used to manipulate the XML file from OpenStreetmap and the conversion of its 2D  
216 format into 3D using CGA Shape Grammar. Through the use of the grammar, the tool's codes  
217 were adjusted accordingly until the required result was achieved. The final model was exported  
218 into fbx format that is recognized by Unity (Figure 1).

219 Autodesk Maya software was also chosen for 3D modeling and animation. Maya was utilized for  
220 the development of the vehicles and other assets (e.g. traffic lights, advertisement billboards)  
221 imported into the game engine. The next step, after the development of the city model, was the  
222 specification of the static and dynamic models. These refer to traffic lights, street lamps, street  
223 signs, the traffic signs, road stakes and advertising billboards in Unity. In respect to advertising  
224 billboards static and dynamic models, designed in Autodesk Maya, have been used to represent  
225 static and motion-based advertisements. For all the 3D models used in the simulator, attention  
226 was paid to keep the geometric complexity relatively low, thus allowing us to have the simulator  
227 running in high frame rates. Additional car assets have been imported in the simulator through  
228 car models in existing digital libraries, to provide variety. The imported car models were  
229 modified accordingly to abide with the regulations of the Cypriot authorities that state that the

230 steering wheel has to be on the right hand side. The selection of the car models was based on car  
231 types and brands currently used in Cyprus. This was a criterion so as to enhance the realism of  
232 the simulator. Colour modifications and textures manipulations helped the design of the car  
233 models to become more realistic.  
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238 Figure 1 The GIS model of the black-spot (top picture) and its realization in the driving simulator  
239 with the road divided into sections A-G (bottom picture)  
240

241 The final step for the simulator environment was the specification of the routing of traffic. For  
242 this task the Maya software was utilized to create car movements as overlay animation. Firstly,  
243 the customized city model was imported into Maya and the cars were added accordingly. To  
244 have a realistic view of the cars while moving, their wheels were manipulated using MEL  
245 scripting language (Expression Editor), to make them rotate according to speed. The movements  
246 of cars were specified through the Motion Path. Upon specifying the motion-path cars were  
247 imported in Unity. The vehicle paths specified were based on a preliminary analysis of traffic  
248 routing on the selected black-spot. The distribution of accidents based on time, at the black spot  
249 was used to pinpoint the most critical time. This was necessary in order to replicate the  
250 conditions when the majority of the accidents occur. In the same manner, some additional  
251 hazardous vehicles manoeuvres were also modelled to mimic critical safety scenarios. These  
252 were used to stress test drivers' behaviour under overloading.

253 The final step of the simulator design was the development of the functionality that would enable  
254 the interactivity between the user and the simulator and this was realised in Unity through  
255 JavaScripting. A screenshot of the simulator's user interface from the driver's perspective is  
256 illustrated in Figure 2.  
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Figure 2 The interface of the developed driving simulator

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## RESEARCH DESIGN

263 The aim of this study was to determine the effects of roadside advertising (billboards) and traffic  
264 signalling on driver attention and crash risk at an accident black-spot in Limassol-Cyprus. Using  
265 this simulator, as presented in the previous section, the research design then employed an  
266 experimental evaluation to test the aforementioned (and other) effects. Participants drove a pre-  
267 specified route in the designed road network both with and without billboards and traffic lights at  
268 a major intersection that has been identified by the police as a critical safety point. The driving  
269 conditions near the billboards were purposefully dangerous to stress-test the effect of distraction  
270 on driver's workload. Specifically, cars were cutting in or crossing the traffic light at high speed.  
271 The conditions tested during the experiment were the use of static and animated advertisement  
272 for the billboards, and absence and presence of traffic lights. Figure 2 illustrates a sample  
273 screenshot of the virtual road design in Unity from the driver's perspective. Prior to the  
274 experiment participants were familiarized with the simulator and briefed on the task they had to  
275 perform. The road network used for the simulator training was different from the model used  
276 during the experiment. Each participant had to complete a set of four scenarios to cover the  
277 different combinations of experimental conditions: with/without traffic lights and with/without  
278 animated advertisements (as also shown in Table 1).

279 Data were collected at different stages: before, during and after the experiment. The pre-  
280 experimental data collection stage concentrated on participants' demographics, driving  
281 experience and historical data relating to driving. The post-experimental data collection focused  
282 on recall of advertisement types and location. During the drivers' engagement with the  
283 experimental conditions, information was recorded relating to driver workload. In particular,  
284 manifestations of workload, such as lateral deviations, crash location and speed, were recorded  
285 on a time-location plot. Data were captured through analyst observations along with video  
286 recordings. Speed was automatically recorded by the simulator, while lateral deviations were  
287 counted based on lane departures during and after the experiment by studying the video  
288 recordings. To facilitate the data collection task the road was divided into 7 sections as illustrated  
289 in figure 1. The specification of these sections was based on infrastructural properties and  
290 billboard locations. With the completion of the experiment, the analyst studied each video and  
291 verified the correctness of the recorded data during the experiment. In sum, the following data  
292 have been collected for each participant, at each road section: speed, lane deviation, and accident  
293 occurrence. It should be noted that the observations were stopped once an accident occurred.  
294 Speed measurements were based on an inflated scale to examine the effect of an exaggerated

295 speedometer reading on driving speed. Under normal conditions drivers will constantly monitor  
 296 their speedometer to maintain the desired/legal limit. In this experiment users were presented  
 297 with overrated speed on the speedometer to find out if they will abide by the law enforcement.  
 298 There were 20 participants (10 male) in the present study, with a mean age of 24 (SD = 4.0). All  
 299 participants had held a full driving license. The age range of participants in the experiment was  
 300 based on the mean age of drivers that had accidents on the black-spot under study as this was  
 301 calculated from historical data provided by the police. All participants were Cypriot residents  
 302 and hence familiar with the right-hand traffic regulations. During the experiment, participants  
 303 were asked to drive as they normally would, given the conditions and the posted speed limits.  
 304 After the experiment, participants were asked to recall if they saw any advertisement, the type of  
 305 advertisements (static/dynamic), their location and content.

306

### 307 DATA ANALYSIS AND RESULTS

308

309 As can be deduced from the aforementioned description of the experiment, this involved a  
 310 repeated measures experimental design since each of the 20 participants took part in 4  
 311 tests/experimental conditions. The data collected during the different stages of the study were  
 312 merged into one dataset, which included the necessary information for both the participants (e.g.  
 313 gender, age, driving experience) as well as their behavioural measures during the four  
 314 experimental conditions (e.g. accident occurrence, speed, lane change, etc). The design, is  
 315 illustrated schematically in Table 1, and shows how we resulted with 80 (correlated)  
 316 observations in total.

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Table 1 The repeated measures experimental design

Traffic Lights YES		Traffic Lights NO	
Advertisement Static	Advertisement Animated	Advertisement Static	Advertisement Animated
Subject 1-20	Subject 1-20	Subject 1-20	Subject 1-20

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319 Both descriptive/exploratory data analysis as well as inferential statistical analysis employing  
 320 regression modeling, were applied to this dataset with the help of the widely used statistical  
 321 package SPSS (version 20). In particular, we explored how (a) relevant factors, such as driving  
 322 speed and lane deviation, which is considered in the literature as good predictor of accident, are  
 323 affected by the experimental conditions, (b) how the outcome variable (i.e. accident occurrence)  
 324 is associated, if so, with the driver's background characteristics, the two experimental conditions  
 325 tested and other behavioral factors relevant to driving (as of (a)), and (c) how all the relevant  
 326 variables can be accounted together for the prediction of accidents in a multivariate repeated  
 327 measures model.

328

329 First we explore some descriptive analysis demonstrating some relationships between some  
 330 background variables of interest with the outcome variable as well as some potentially  
 331 confounding variables. The outcome variable of interest for this analysis is the occurrence of an  
 332 accident. Out of the 80 observations in the above conditions, 25 (31.25%) accidents were  
 333 recorded. The vast majority of them (24 corresponding to 96%) happened at Section D of the  
 334 road. The high accident rate is attributed to the hazardous traffic conditions simulated in the  
 335 scenarios and is considered for the interpretation of the results.. However, it should also be noted  
 336 that a new variable was constructed including information of whether an accident occurred or  
 337 not, independently of location, and this was used for further statistical modeling.

337



338 Table 2 presents the association of accident occurrence with the subject's gender. As can be  
 339 seen, the majority of accidents (68%) observed is attributed to male drivers. This association was  
 340 found to be statistically significant (chi-square=4.713, p=0.03). Of crucial importance is the  
 341 investigation of the association between the two experimental conditions and the outcome  
 342 variable of interest. Table 3 presents these associations.  
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Table 2 Cross-tabulation of accident occurrence by gender

Accident	Gender		Total
	Frequencies (row %)		
	Female	Male	
No	32 (58.2%)	23 (41.8%)	55
Yes	8 (32%)	17 (68%)	25

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Table 3 Two-way cross-tabulation between advertisement and traffic light conditions with the occurrence of accidents

Advertisement	Accident	Traffic Lights		Total
		Frequencies (row %)		
		No	Yes	
Static	No	14 (50%)	14 (50%)	28
	Yes	6 (50%)	6 (50%)	12
Animated	No	10 (37%)	17 (63%)	27
	Yes	10 (76.9%)	3 (23.1%)	13
Total	No	24 (43.6%)	31 (56.4%)	55
	Yes	16 (64%)	9 (36%)	25

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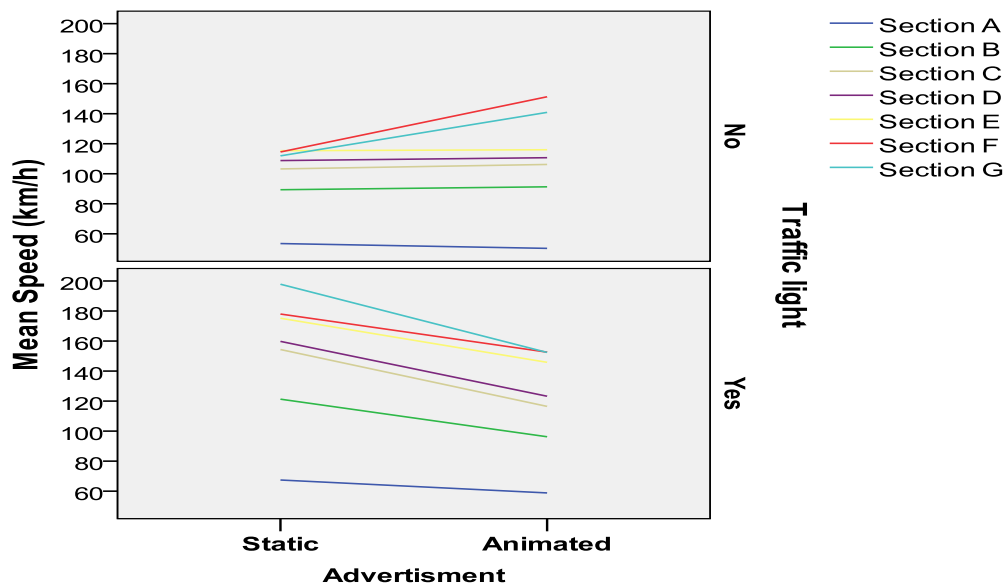
As can be seen, there are no notable differences in the distribution of accidents when we compare traffic light conditions, under the static advertisement scenario (in both cases there were 6 accidents recorded). In contrast, there are some apparent statistically significant differences (chi-square= 5.584, p=0.018) when we consider the distribution of accidents under the animated advertisement condition (i.e. significantly more accidents occur without traffic lights). Other measured variables of interest, which are also considered as explanatory variables in further modeling, are the driving speed and deviation from lane (as an indicator of loss of focus/attention). Speed was recorded separately for each of the seven road sections. The table in Table 4 shows the descriptive statistics for the driving speed at each section, and as can be observed, the average is consistently increased as the driver proceeds in the road section.

Table 4 Descriptive statistics of speed (in km/h) by gender

Road section	N	Overall Mean (SD)	Male Mean (SD)	Female Mean (SD)
Section A	80	60.70 (19.8)	60.5 (19.9)	60.9 (20)
Section B	80	102.61 (52.1)	94.35 (49.8)	110.9 (53.7)
Section C	80	122.56 (80.9)	100.67 (81.6)	144.5 (75)
Section D	80	130.6 (84.5)	110.12 (86.2)	151.1 (78.7)
Section E	56	143.98 (90.6)	123.78 (81.8)	158.1 (94.8)
Section F	55	149.13 (112.2)	127.8 (95.3)	164.47 (122)
Section G	55	151.58 (126.6)	149.35 (133)	153.2 (123.5)

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362 Speed measurements were based on an overrated scale to examine the effect of an exaggerated  
 363 reading on the speedometer on driver behavior. Results from this analysis revealed that female  
 364 drivers were not affected by the overrated speedometer readings and were driving 24% faster  
 365 than male. This could be attributed to low situational awareness (Endsley, 1995) of female in  
 366 comparison to male drivers and hence, failure to recognize the indicated high speed readings. It  
 367 is probably more interesting to explore how the average speed is affected by the experimental  
 368 conditions. Figure 3 illustrates the association of the two conditions and their interaction with the  
 369 speed.  
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371 Figure 3 Mean speed (km/h) by traffic light and type of advertisement conditions.  
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374 As can be seen, under the 'animated advertisement' condition the speed was constant in most  
 375 sections (apart from Section E) independently of whether there were traffic lights or not. Under  
 376 the static advertisement scenario, however, there was an observed increased speed when there  
 377 were traffic lights compared to the scenario without. This could be attributed to the driver being  
 378 attentive to the animation and hence reverting from driving fast. This however, could be  
 379 considered as an indirect distractor to the main task of driving. Similar behavior is observed  
 380 when drivers revert from the driving task due to distraction, such as mobile phone use. In such  
 381 cases driving speed degrades so as the driver would be able to attend the information  
 382 requirements of the distraction. The increased speed with the presence of traffic lights could also  
 383 be attributed to the fact that the light timing was set to change to red when the simulated vehicle  
 384 was approaching the traffic signals and users were accelerating to pass without stopping. In  
 385 order to simplify further the analysis involving the effect of speed, a new variable was created to  
 386 capture average standardized speed across all sections, up to the point of accident occurrence.  
 387 As mentioned earlier lane deviation, is usually considered an indicator of loss of concentration  
 388 during driving. A binary variable was thus created to account for the occurrence of lane deviation  
 389 (or not) up to the point of accident occurrence (or till the end). Various variables were found to  
 390 influence the probability of lane deviation as shown in the results of a binary logistic regression  
 391 model. Binary logistic is the appropriate regression model of binary outcome variables, as is lane  
 392 deviation for our analysis. Regression modelling allows for the combined effects of various

393 explanatory variables to be taken into account when modelling the outcome variable (Hutcheson  
 394 & Sofroniou, 1999); the ultimate aim sometime is to make predictions for the probability of an  
 395 event happening (i.e. lane deviation) with the regression equation which can be formed with the  
 396 help of the regression coefficients, presented in Table 5.

397 Table 5 A logistic regression model for the probability of lane deviation

Variables	Coefficient B	S.E.	Wald	df	p-value (Significance)	Exp(B)
Light(Yes)	-3.662	1.069	11.748	1	.001	.026
Animated(Yes)	-1.298	.841	2.382	1	.123	.273
Gender(F)	1.407	.621	5.124	1	.024	4.082
Animated(1) by Light(1)	4.000	1.446	7.651	1	.006	54.624
Z-Average Speed	1.540	.454	11.487	1	.001	4.664
Constant	.667	.618	1.164	1	.281	1.949

398 According to the results of this model, average driving speed, gender, the traffic light  
 399 experimental condition as well the interaction of traffic lights with animation (or not) have a  
 400 significant effect on the probability of lane deviation, as indicated by the significance values for  
 401 this variables (i.e.  $p\text{-value} < 0.05$ ). The direction of this effect is better explained with the  
 402 exponential of the coefficients. For example after controlling for all other variables in the model,  
 403 female drivers are four times more likely ( $\exp(b)=4.082$ ) to deviate from the lane compared to  
 404 male drivers. Similarly when average speed increases by 1 (standard) unit the odds of lane  
 405 deviation increase by more than 4 times (again assuming we control for all other variables). The  
 406 effect of the interaction between the two experimental conditions is also significant and can be  
 407 interpreted as follows: even though without traffic lights the probability of lane deviation is  
 408 significantly higher for the static advertisement condition, this trend is reversed under the traffic  
 409 lights condition, when the probability of lane deviation becomes higher for the animated advert  
 410 condition.

411 What is also interesting is to explore the association of lane deviation and speed with the main  
 412 outcome of interest; that is accident occurrence. To check for this we split the 80 observations  
 413 into those with accident observed and those without and we then compared the means of average  
 414 speed and lane deviation (i.e. the probability of lane deviation) for the two groups. The results  
 415 are presented in Table 6, and show a statistically significant higher probability for lane deviation  
 416 when no accident was observed. The increased speed in the case of no accident compared to the  
 417 accident cases was not found to be statistically significant.

418

Table 6 Comparison of average speed and probability of lane deviation by accident occurrence

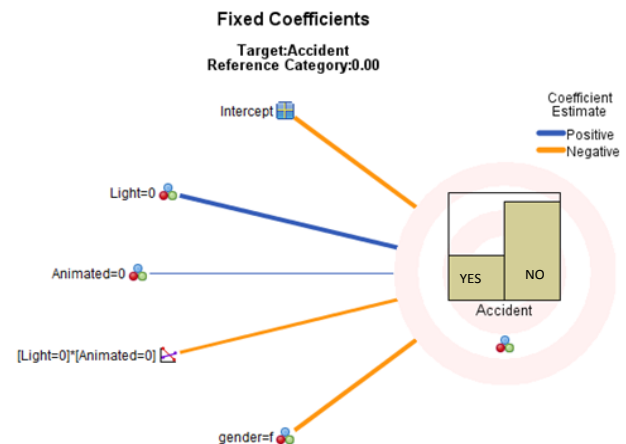
	Accident	N	Mean	Std. Deviation	Std. Error Mean	t-test (significance)
<b>Lane Deviation</b>	No	55	0.636	.49	.065	5.541 ( $p < 0.01$ )
	Yes	25	0.12	.33	.066	
<b>Average Speed</b>	No	55	120.9	66.36	8.95	0.649 ( $p = 0.52$ )
	Yes	25	111.10	60.92	12.18	

419

420 The above associations between the accident occurrence and other variables, even though  
 421 indicative of some patterns, have two main limitations: they fail to account for the inter-  
 422 relationships between different variables, and they do not take into account the repeated  
 423 measures design and the resulting correlations between the observations for each subject. These

424 limitations are overcome with a more advanced regression model. In particular, we employed a  
 425 generalized linear modeling framework (Hutcheson & Sofroniou, 1999) to model the binary  
 426 variable of interest (i.e. accident occurrence) based on various explanatory variables including  
 427 the experimental conditions and participants' characteristics. The appropriate model was a  
 428 logistic regression model, which also deals with the repeated nature of our experimental design.  
 429 Technically, we used the mixed generalized linear modeling tool of SPSS and resulted in the  
 430 model with the main effects shown in in Figure 4 It should be noted that different variations of  
 431 the model were run checking for the effects of other background and behavioral measures (e.g.  
 432 speed and lane deviation) however these were not found to be effective for the response variable  
 433 (i.e. accident occurrence). The model in Figure 4 is considered to be robust and concise  
 434 accounting for both experimental conditions and their interaction as well as driver's gender  
 435 which was found to have a significant effect. Since the explanatory variables were all  
 436 categorical, dummy variables are used in the model to define the categories. The estimated  
 437 coefficients are then interpreted comparatively to the reference categories (i.e. the dummies  
 438 whose coefficients are set to zero: Light=1, Animated=1, Gender=m). Hence controlling of the  
 439 categorical variables in the model is achieved by setting a dummy variable for one of the  
 440 categories to zero and check the effect of the other category(ies) on accident.  
 441

Model Term	Coefficient	Sig.	Exp(Coefficient)
Intercept	-1.226	.031	0.293
Light=0	1.883	.014	6.573
Light=1	0 <sup>a</sup>		
Animated=0	0.948	.218	2.580
Animated=1	0 <sup>a</sup>		
[Light=0]*[Animated=0]	-1.883	.083	0.152
[Light=0]*[Animated=1]	0 <sup>a</sup>		
[Light=1]*[Animated=0]	0 <sup>a</sup>		
[Light=1]*[Animated=1]	0 <sup>a</sup>		
gender=f	-1.314	.020	0.269
gender=m	0 <sup>a</sup>		



Probability Distribution: Binomial

Link function: Logit

<sup>a</sup>This coefficient is set to zero because it is redundant

442  
 443 Figure 4 A model for predicting accident occurrence. A zero score for a variable denotes absence  
 444 of the condition (i.e. light=0 implies no traffic light, and animated=0 implies static advert).  
 445

446 The coefficients of the explanatory variables in regards to their effect on the response variable  
 447 are interpreted similarly to those from a multiple logistic regression (i.e. based on the change in  
 448 odds ratio using the exponential of the coefficient, as for Table 5). For instance, what we could  
 449 say for gender given the above results is that, controlling for all other variables in the model,  
 450 when we move from male to female driver we reduce the odds of an accident by 0.269 (in other

451 words, for males the odds of accident are almost 4 times those of female). Similarly the odds of  
452 an accident are increased by more than 6 times when we remove the traffic lights.

453

## 454 **CONCLUSIONS**

455 Driver workload should always be maintained at optimal levels to ensure adequate level of  
456 vigilance. Experimental work in this field is often ethically unacceptable, if not impossible, if  
457 performed in actual driving conditions. Therefore, driving simulators enable the study of realistic  
458 conditions, without any objective risks, while at the same time lend themselves to a variety of  
459 human factors analyses. The three categories of methods used to assess driver workload,  
460 according to the literature are: measures of task performance, subjective reports and  
461 physiological measures (Brookhuis et al., 2002). Our approach falls under the task performance  
462 group. Lateral deviation is also identified in previous research as one of the most critical driving  
463 performance indicator relevant to safety (DOT, 2006). Our contribution lays in the correlational  
464 analysis of phenotype driver behaviors, such as lane departures, for the assessment of the  
465 primary task of driving, under different infrastructural conditions to identify if billboard  
466 distractions are increasing the risk of accidents.

467 To our knowledge this is the first study that was conducted in Cyprus for black spot analysis  
468 using a driving simulator and local road users. The method provides local authorities with a cost  
469 effective solution that enables the involvement of native driving users for the analysis of local  
470 driving behaviors and road design challenges. The method enables the design and customization  
471 of the road infrastructure for what-if analyses in a modular fashion. This enables the design of  
472 the experimental settings for the analysis of a variety of conditions such as the use of dynamic  
473 and static advertisement of billboards and traffic signals in straightforward fashion.

474 Preliminary results from this study highlight a weak relationship between lateral deviations of  
475 road users and advertisement. Overall the results highlight a significant difference between  
476 gender groups on accident likelihood which verifies earlier findings in the literature (DOT,  
477 2011). Additional results highlight the relationship between speed and animated advertisements,  
478 and speed with accident occurrence. The former indicate that animated advertisements consume  
479 more attentional resources than static advertisements and this is manifested by reduced speed to  
480 compromise for the reduced cognitive resources available. The latter reinforces the finding that  
481 speed increases accident risk and this is more critical with male drivers. Finally, fake reading on  
482 the speedometer seems to affect only male drivers who appear more contextually aware, as also  
483 suggested by their observed superior lateral control in contrast to females.

484 The implications from these preliminary results highlight the need to examine in more detail the  
485 impact of animated advertisement on driving behaviour and its link to accidents. Moreover the  
486 situational awareness difference between genders needs to be analysed further. However, due to  
487 the limited immersive properties that laboratory methods provide, the driving behaviour of  
488 participants might be affected. To that end, we are aiming to recreate the experiment in virtual  
489 reality settings for improved realism and hence enhance driving conditions which in return will  
490 affect participants' driving behaviour. This will improve observational accuracy and thus yield  
491 more robust results.

492

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