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# Behavioral and workload measures in real and simulated driving:

# Do they tell us the same thing about the validity of driving simulation?

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The authors declare no conflict of interest.

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# Behavioral and workload measures in real and simulated driving: Do they tell us the same thing about the validity of driving simulation?

#### 78 **1. Introduction**

For the sake of controllability, repeatability, and safety, driving simulators are 79 extensively used for driver behavior research. However, they must have appropriate validity 80 to be useful human factor research tools. Since Blaauw (1982), few studies have compared 81 on-road and simulator driver behavior. While such comparative studies are not easy to carry 82 out (Carsten & Jamson, 2011; Carsten et al., 2013), they remain the most appropriate means 83 to evaluate and improve driving simulator validity. Most of them relied on behavioral 84 measures to capture the differences between on-road and simulated driving (Blana, 1996; 85 86 Mullen et al., 2011; Shechtman, 2010). More recently, few studies assessed the level of mental workload when driving in a real vehicle or in a simulator and showed contradictory 87 results (e.g., Diels et al., 2011; Johnson et al., 2011; Milleville-Pennel & Charron, 2015). On 88 another note, it has been found that mental workload measures were more sensitive to task 89 demands than behavioral measures (Mehler et al., 2009, 2012; Yang et al., 2013). Therefore 90 the present study was aimed at assessing behavioral and mental workload measures to better 91 understand the differences between real and simulated driving. 92

93 Blaauw (1982) proposed to assess driving simulator validity based on physical and behavioral criteria. Physical validity corresponds to the extent to which software and 94 hardware components of the simulator reproduce the physical reality of driving. Behavioral 95 validity has been defined as the correspondence between driver behavior in a simulator and on 96 the road. While some studies have paid attention to hardware components of the simulator to 97 improve physical fidelity (e.g., Chatziastros et al., 1999; Jamson & Jamson, 2010; Kappé et 98 al., 1999; Klüver et al., 2016), it is relatively well established that behavioral validity is a 99 more important criterion than physical validity. This can be explained by the fact that the 100

strict duplication of reality is impossible to obtain in simulated driving (Pinto et al., 2008). In 101 addition, a high level of physical fidelity is not always required to achieve good levels of 102 behavioral validity (Blana, 1996; Klüver et al., 2016; Pinto et al., 2008). Blaauw (1982) also 103 made a distinction between relative and absolute behavioral validity. Absolute validity is 104 established with the sameness of numerical values in the behavioral measures in the two 105 different environments. Relative validity is concluded when the differences due to the 106 experimental manipulation are of the same order and direction and have a similar or identical 107 magnitude in real and simulated driving conditions (for a discussion, see Klüver et al., 2016). 108 Longitudinal and lateral control have been mainly used to assess the behavioral validity 109 of driving simulators, with speed being undoubtedly the most widely used (see 110 Supplementary Table 1 for an overview of results obtained in studies that have compared real 111 and simulated driving under free driving conditions. For a recent review of driving simulator 112 113 validation studies, see Wynne et al., 2019). For lateral position, strong absolute differences between simulated and real driving have been found in the mean and standard deviation of 114 lateral position. Relative validity of simulated driving is however relatively well preserved 115 (see Supplementary Table 1). For speed, 11 out of the 20 studies which examined vehicle 116 speed when driving freely concluded absolute validity while 7 concluded relative validity 117 (two studies showed significant differences between the two systems). When relative validity 118 of the experimental treatment is was found, driving speed is was higher in simulators than on 119 roads (see Supplementary Table 1). Whether for lateral control or speed measures, these 120 results greatly depended on the type of simulator (for a recent review, see Klüver et al., 2016) 121 and driving context, such as the type of road, road geometry (Bella, 2008; Bittner et al., 2002; 122 Branzi et al., 2017), presence of oncoming traffic (Alm, 1995) or traffic in the lane (Wang et 123 al., 2010), and both posted and traffic speed (Wang et al., 2010). 124

While most of the studies evaluated behavioral validity, Blaauw (1982) initially 125 proposed not only to assess behavioral measures but also mental workload measures. For 126 Malaterre and Fréchaux (2001), this corresponds to the psychological validity of driving 127 simulators defined as the sameness of the psychological processes underlying the driving 128 activity, including mental workload. Focusing specifically on mental workload, which is the 129 result of the interaction between task demands and the attentional resource capacity of the 130 operator (Borghini et al., 2014; O'Donnell & Eggemeier, 1986), this distinction finds model-131 based and empirical evidence. De Waard (1996) offered a model which relates the operator's 132 performance and workload to task demands. When the task demands are low, the operator is 133 weakly engaged in the task and the level of mental workload is high while performance is 134 low. When the task demands outmatch the operator's attentional resources, the latter is 135 overloaded and performance is dramatically low. In-between these two extrema, two states 136 137 are of interest here: i) when task demands are high enough to engage the operator whose attentional resources meet the task requirements, performance is high while mental workload 138 level is low; ii) when the task demands increase, the operator has to engage more resources to 139 140 maintain his performance and mental workload level is consequently high. This last state has been highlighted in studies showing that when task demands increased, physiological or 141 visual attention measures of mental workload were sensitive to these changes while driving 142 performance measures were not (e.g., Mehler et al., 2009, 2012; Yang et al., 2013). 143 Based on these considerations, when driving performance, namely vehicle handling, is 144 maintained in a simulator compared to on-road driving, the mental workload level may then 145 be either the same in both environments or different. If the mental workload level is the same, 146

147 then the simulator does not place more demands on vehicle handling. The control of the

vehicle, which belongs to the operational level of driving activity and requires little

149 attentional resources, is not hampered when driving a simulator. If the mental workload level

is higher in the simulator, this suggests that more attentional resources have to be allocated to 150 151 maintain driving performance. This increase in task demands could come from the lack of vestibular and proprioceptive cues (e.g., Bittner et al., 2002; Boer et al., 2000; Klee et al., 152 1999). Lastly, if the mental workload is higher when driving on a road, the difference could 153 be due to a higher perception of risk and consequences of collisions that are not present in the 154 simulator (Bella, 2008; Deniaud et al., 2015; Engström et al., 2005; Mueller, 2015). 155 Surprisingly, the assessment of mental workload when driving on a road and on in a 156 simulator has given rise to a small pool of studies and to contradictory findings. Alm (1995) 157 and Blaauw (1982) first showed that, while longitudinal control was well preserved whether 158 the simulator was static or dynamic, driving in a simulator was quoted as being more difficult 159 and more effort-demanding than driving for real. Similarly, Diels et al. (2011) found a very 160 close correspondence in speed between the car and the simulator but subjective ratings of 161 162 mental workload tended to be moderately higher in the driving simulator. Engström et al. (2005) compared vehicle-handling measures and physiological measures of mental workload 163 (heart rate and skin conductance) in a static simulator, a dynamic simulator and an 164 instrumented vehicle. Driving measures in the simulators were consistent with those obtained 165 in real driving conditions. However, physiological measures of workload were higher in the 166 field, suggesting an increased effort due to higher actual risk in real traffic. Other studies 167 confirmed that physiological signals of mental workload were higher when driving on the 168 road than driving in a simulator; however, no speed or lateral-control measures were reported 169 (Johnson et al., 2011; Milleville-Pennel & Charron, 2015; Reimer & Mehler, 2011). At the 170 very least, Milleville-Pennel and Charron (2015) reported that the self-rated driving 171 performance was judged to be lower when driving in the simulator than in the field. Using a 172 cognitive secondary task, Reimer and Mehler (2011) and Mehler et al. (2009) showed in two 173 different studies sharing the same manipulation that when the secondary task difficulty 174

increased, the performance decreased by a greater amount in the simulator (Mehler et al., 175 2009) than in real driving conditions (Reimer & Mehler, 2011). This is in keeping with the 176 idea that mental demand coming from simulated driving is higher. A similar result was found 177 by Chen et al. (2015). In this study, participants had to perform a reaching task to virtual and 178 real targets while verbally responding to an auditory tone. Results showed that the response 179 time to the auditory stimulus was significantly longer for the virtual than for the real target 180 condition, confirming that a VR-based task was more attention-demanding than its 181 conventional real-world counterpart. 182

In sum, it appears that measures of behavioral validity may return only a partial view of 183 the correspondence between driving for real and driving in a simulator. As suggested by 184 model-based and empirical evidence, attentional resources required by the driving task during 185 real and simulated driving might differ, while behavioral measures may not (De Waard, 1996; 186 187 Mehler et al., 2009, 2012; Yang et al., 2013). This would lead to wrongfully inferring (behavioral) absolute validity while validity on other dimensions would not be established. 188 The objective of this study was therefore to assess driving simulation validity by 189 confronting measures of behavioral (speed) and psychological (mental workload) validity. In 190 this purpose, we selected a route where participants could drive freely and were likely to 191 adopt a similar speed in real and simulated conditions. This made it possible to examine 192 whether mental workload level differed between real and simulated driving while speed 193 measures had a good correspondence in both driving conditions. It also appears that workload 194 level in virtual environments has received little attention and results are controversial. One 195 reason could be the use of a variety of measures in isolation, which are known to dissociate 196 (e.g., Brookhuis & de Waard, 2002; Faure et al., 2016; Yeh & Wickens, 1988). We therefore 197 used a multiple measurement approach of workload, with subjective, physiological (eye-blink 198 rate) and secondary task performance measures. 199

#### 200 **2. Method**

#### 201 2.1. Participants

Twenty-four participants (8 females) took part in the experiment. They had a mean age 202 of  $43.5 \pm 8.9$  years, ranging from 27 to 56 years old. All participants had their driver's license 203 for a minimum of two years. The self-reported experience ranged from 9 to 36 years (mean = 204  $23 \pm 9.6$  years) and the self-reported total mileage ranged from 20,000 to 420,000 km (mean = 205  $300,000 \pm 196,000$  km). All participants had normal or corrected-to-normal vision, and were 206 207 naïve as to the aims of the experiment. Before they took part in the study, all gave their informed consent which was approved by the local Ethics Committee. The ethical 208 considerations and principles of the World Medical Association Declaration of Helsinki 209 regarding experimentation were respected. 210

## 211 **2.2. Experimental set-up**

#### 212 **2.2.1. Real vehicle**

The field experiment was run using a Renault Clio III with manual transmission. It was equipped with a GPS sensor (GlobalSat Technology BR-355, Taiwan), a Controller Area Network (CAN) interface (PEAK System PCAN-USB, Germany), and three cameras (4Kam Wideboy Sport, United-Kingdom). GPS data were logged with 1 Hz, and vehicle-related data were logged between 10 and 50 Hz depending on the data. All data sources were synchronized on a computer (Toshiba Satellite Pro A300, Japan) running a data acquisition and synchronization software (Intempora RTMaps, France).

# 220 2.2.2. Fixed-base simulator

The experiment in the virtual environment was run using a fixed-base driving simulator composed of a force-feedback steering wheel, foot pedals, gearbox (Logitech G25), and a Renault Clio driving seat (Figure 1). The simulated environment was displayed on three screens 0.53 m wide and 0.30 m high (resolution: 1,280 x 1,024 pixels; refresh rate: 30 Hz). The distance between the participant's head and the center of each screen was 1 m. The visual angle subtended by this configuration was 97° horizontally and 30° vertically. The images were calculated and projected at the participant's eye height, with the simulated viewing angle aimed at the vanishing point of the simulated scenario. Engine sounds and environmental noises were generated by five speakers. The speedometer was displayed in the lower part of the central screen. The central mirror was superimposed on the visual environment in the upper part of the central screen. Data were collected at a sampling rate of 30 Hz.



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Figure 1. Illustration of the fixed-base driving simulator.

## 234 **2.2.3. Eye-tracker**

Two identical Pertech head-mounted eye-trackers were used to record eye measures in 235 the instrumented vehicle and the driving simulator, respectively. This monocular eve-tracker 236 237 uses a pupil tracking technology with an image processing algorithm to define the ocular direction (50 Hz sampling rate) and has 0.25° of accuracy (according to the manufacturer 238 specification). A seven-point calibration was performed at the beginning of each real and 239 240 simulated experimental trial. Eye measures data were synchronized with simulated and real car data so that they could be analyzed relative to the road geometry and any driving 241 measures. 242

### 243 **2.2.4. Driving environment**

The experimental route (Figure 2) was a 7.5 km long section of the D786 road near 244 Saint-Brieuc (France) between Plérin (GPS coordinates: 48.549687, -2.796007) and Binic 245 (GPS coordinates: 48.6014398, -2.8356539) and consisted of three different road sections. 246 The first road section was a circa 4.2 km long dual 2-lane carriageway road including a 247 roundabout. The posted speed limit was 110 km/h and then 70 km/h before entering the 248 roundabout (limited to 50 km/h). except at the approach to the roundabout where it was set at 249 <del>70 km/h</del>. After the roundabout, the posted speed limit was set again at 110 km/h and then 250 90 km/h at the end of the section (hereafter labeled highway environment). The roundabout 251 was excluded from the analysis as the corresponding speed limit (50 km/h) was not in line 252 with high-speed environment and free driving conditions. This initial decision was reinforced 253 by the fact that in real driving conditions all participants had to stop due to the presence of 254 other vehicles on the roundabout. The second road section was a circa 1.8 km undivided two-255 256 lane rural road. The posted speed limit was 90 km/h in the first section and 70 km/h in the last winding section (hereafter labeled rural environment). The third road section was a circa 257 258 1.5 km urban single carriageway road passing through the town of Binic. The posted speed limit was 50 km/h (hereafter labeled urban environment). 259

With the objective of minimizing the impact of traffic uncertainties and congestion on the driving task, we selected based on traffic surveys carried out on several sections of the experimental road two periods of time (9:30 am to 11 am and 2:30 pm to 4 pm) during which traffic density was the lowest. Furthermore, in case of particular events during the driving sessions in the real environment that could not be reproduced in the simulated driving session (e.g., red traffic light, pedestrians, vehicle overtaking), the corresponding sequences were excluded from data analyses. These sequences represented less than 5% of the data.

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#### Mental workload comparison between on-road and simulated driving





Figure 2. Overview of the experimental route and its different subsections.

All the important features of the real-world environment were replicated in a simulated 276 277 3D virtual world (see Figure 3). This included the three different driving environments (dual 2-lane carriageways and single carriageways in rural and urban contexts, respectively), 278 roundabout, road signs, and buildings especially in the urban section. The road signs were 279 280 positioned in the visual database at their corresponding locations in the real road. The traffic light in the actual traffic conditions was turned to green in the simulated scenario. In addition, 281 282 traffic running in the opposite direction to that of the participant's car was introduced in the virtual environment. The traffic density corresponded to that surveyed during the two periods 283 of time chosen for the on-road session. The simulated vehicles were of several types 284 (passenger vehicles, commercial vehicles, trucks) and colors and obeyed the speed limit 285 proper to the speed sections in the three driving environments. 286

Based on the classification of the effects of driving task demands on mental workload level (see Patten et al., 2004, 2006) and previous empirical results showing that drivers' mental workload is differently affected depending on the driving contexts (e.g., Faure et al., 2016; Mackenzie & Harris, 2017; Michaels et al., 2017; Patten et al., 2006), these three different driving environments were selected with the aim of triggering different levels of mental workload. The highway environment consisted of a straight carriageway located in the countryside with monotonous surrounding conditions, so that drivers would be undistracted



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- Figure 3. Illustration of the three different driving environments (top: highway environment in
- the first section at 110 km/h; middle: two-lane rural road in the section at 90 km/h; bottom:
- 301 urban environment) in the simulator and in real situation. Views in the real situation were
- 302 taken from Street View database (Google).
- 303

by the roadside environment and could control their vehicle in a consistent manner. This 304 environment was therefore supposed to trigger a low level of mental workload. The rural 305 environment represented a winding single carriageway lined with trees. In this scenario, an 306 intermediate level of workload was expected as it required a greater degree of steering control 307 than the highway environment. Lastly, the urban environment represented a visually rich 308 neighborhood (street furniture, buildings, signs, traffic lights and intersections). Drivers had 309 to adjust their trajectory and their speed and had to pay attention to their surrounding 310 environment as they progressed through the environment. The urban section was therefore 311 assumed to impose a high level of mental workload. The rationale for selecting these road 312 sections was to examine whether the effects of the driving environment on drivers' mental 313 workload varied in an absolute (i.e., sameness of numerical values between the two systems) 314 or relative (i.e., same order and direction and similar or identical magnitude in the two 315 systems) manner during real and simulated driving. 316

#### 317

## 2.3. Experimental design and task

The experimental route was driven twice by all participants, one with the instrumented 318 319 vehicle in real condition and one with the simulator in the 3D world. As the driving environment is known to affect drivers' mental workload level, we used a within-subject 320 design with two driving conditions (real road and driving simulator) and three driving 321 environments (highway environment, rural environment and urban environment). In the real 322 driving condition, participants were asked to drive in the way that they were used to. In the 323 driving simulator condition, they were asked to drive as if they were in a real car on a real 324 road, while obeying the traffic regulations and speed limits. 325

While participants were performing the primary driving task, they also had to perform a tone reaction time task. This subsidiary secondary task consisted of responding verbally ("top") to an auditory stimulus (100 ms duration, 1.5 kHz amplitude) as quickly as possible.

To prevent any anticipation, the inter-stimulus interval varied between 7 and 10 s. It was 329 adapted from the standard subsidiary secondary task (motor response to visual stimulus) as 330 one of its drawbacks is the occurrence of manual and visual interference with the driving task 331 (see Cantin et al., 2009; Jahn et al., 2005). Auditory stimuli were pre-recorded. In the real 332 vehicle, sounds were played by an independent speaker. In the simulator, they were embedded 333 into the engine sound. The participants' responses were recorded by the eye-tracker 334 microphone. Compliant with the subsidiary task paradigm (Cain, 2007), participants were 335 instructed to prioritize the primary driving task while responding to the auditory stimuli as 336 rapidly as possible. 337

### 338 **2.4. Procedure**

On their arrival, the nature and requirements of the experiment were explained to the participants. They signed an informed consent form and demographic data were collected. They drove the real and simulated condition in a counterbalanced order.

For the field test, participants sat in the instrumented vehicle and were invited to adjust 342 the position of the seat, steering wheel and mirrors. They were then equipped with the eye-343 tracker before it was calibrated. Two experimenters also sat in the vehicle. The first one was 344 seated at the front of the vehicle to record traffic events. The second was at the back to 345 supervise data recording. When necessary due to the counterbalanced order of presentation of 346 the two driving conditions, participants were trained with the audio-vocal reaction time task. 347 Then, they drove the vehicle from the lab to a parking lot at the entrance to the course, located 348 15 km away, which gave them time to get accustomed to the experimental vehicle. Once the 349 quality of the ocular tracking had been checked, the experimenter seated in front reminded the 350 participant of the instructions of the driving session. The participants then drove on the 351 experimental course and performed the audio-vocal reaction time task concurrently. Once the 352 total experimental route had been completed, they stopped in a parking lot and completed the 353

mental workload assessment questionnaire (NASA-TLX scale from Hart & Staveland, 1988)
before removing the eye-tracker. Finally, one of the experimenters brought the vehicle back to
its starting point.

For the simulator test, a similar procedure was used. Participants first made sure they 357 were comfortable when sat in the simulator. One of the experimenters then fitted and 358 calibrated the eye-tracker. When necessary (i.e., when they drove in the simulator condition 359 first), participants familiarized themselves with the subsidiary secondary task. Then, they 360 underwent a training period with the simulator which consisted of driving on a winding rural 361 road with bends varying in length, curvature and direction (13 km long without any other 362 traffic). Finally, a reminder of the instructions for the driving and the reaction time task was 363 given. At the end of the driving session, participants completed the NASA-TLX questionnaire 364 before removing the eye-tracker. The experiment lasted approximately 1 hr. 30. 365

#### 366 **2.5. Dependent Variables**

### 367 **2.5.1. Speed**

The participant-determined speed was used to compute the mean speed (km/h) over the entire experimental route. As the speed limit varied within and between driving environments, the mean speed was also calculated for each speed section (n = 7) for each environment. Since most comparative studies were run in a unique driving environment, this made it possible to examine the potential differences between real and simulated driving for different posted speed limits and sections.

## 374 **2.5.2. Blink rate**

Eye blinking has proven to be a useful and reliable signature of drivers' mental workload variations (e.g., Lohani et al., 2019), with a downward trend in the eye blink rate when the primary task difficulty increased (e.g., Faure et al., 2016; Hancock et al., 1990; Veltman & Gaillard, 1996; Wiberg et al., 2015). Capitalizing on previous research, endogenous blinks corresponded to eye closures lasting between 80 (Benedetto et al., 2011,
2014; Faure et al., 2016; McIntire et al., 2014) and 500 ms (Stern et al., 1984). Blink events
were captured according to a modified version of the algorithm developed by Pedrotti et al.
(2011) for low-speed eye-trackers, and the blink rate (number of blinks per minute) was
computed in each driving environment and condition.

384 2.5.3. Audio-vocal reaction time

Reaction time to the secondary task was defined as the time lag between tone onset and voice onset. Before the mean response time (ms) was computed for each participant, outlier data were removed following a procedure similar to Cantin et al. (2009). Outliers corresponded to reaction times shorter than 150 ms or slower than each participant's mean reaction time plus two standard deviations. The mean response time was computed for each driving environment and condition.

#### 391 **2.5.4. Subjective workload**

The NASA-TLX was used to assess the operator's perceived workload according to six dimensions: mental demand, physical demand, temporal demand, performance, effort, and frustration. The mean ratings on each dimension and the global score were computed for each driving condition (following the unweighted procedure).

## 396 **2.6. Statistical Analysis**

All statistical analyses were done with p set at .05. Speed differences between real and simulated driving were examined for the entire experimental route and each speed section (n = 7). Speed results were first checked for distribution normality using Shapiro-Wilk tests for small samples. Pairwise comparisons were conducted using paired t tests when the data sets were normally distributed or Wilcoxon signed-rank tests when one of the distributions was not normally distributed. Effect sizes were reported as Cohen's d. Because of technical problems during real driving conditions, these analyses were performed with 21 participants.

404	The possible effects of the driving condition and the driving environment on blink rate
405	and reaction time to the secondary task were assessed using 2 (driving conditions: real
406	driving, simulated driving) $\times$ 3 (driving environment: highway, rural, urban) repeated
407	measures analyses of variance (ANOVAs). Mauchly's test of sphericity was run to determine
408	whether the sphericity assumption was violated. When necessary, the degrees of freedom
409	were adjusted with the Greenhouse-Geisser correction. Tukey's honestly significant
410	difference (HSD) procedure was run to follow up on significant effects. For each significant
411	effect, partial eta-squared $(\eta_p^2)$ was computed to determine the proportion of total variability
412	accounting for the effect. Note that the reaction time analysis was performed with 21
413	participants because of technical problems during real driving condition.
414	Subjective workload ratings were analyzed following similar procedures as described
415	above. Differences between real and simulated driving were first compared on global demand.
416	To explore the origin of global demand differences, a repeated-measure ANOVA was run on
417	mean ratings, with driving conditions and NASA-TLX subscales as within-participant factors.

418

### 419 **3. Results**

#### 420 **3.1. Speed**

421 Statistical comparison of the mean driving speed computed for the entire route revealed 422 that speed did not differ between real (73.6 ± 3.4 km/h) and simulated (74.8 ± 5.4 km/h) 423 conditions (t(20) = -1.18, p = .25). Analyses comparing speed between the two driving 424 conditions in each speed section showed that speed did not differ in the highway environment 425 (110 km/h: Z = 1.02, p = .30; 70 km/h: Z = .08, p = .93; 110 km/h: t(20) = 1.93, p = .07; 90 426 km/h: t(20) = -.21, p = .83) nor in the rural environment (90 km/h: t(20) = .81, p = .43; 427 70 km/h: Z = 1.69, p = .09). However, in the urban environment, speed was significantly 428 higher in simulated than real condition (Z = 4.01,  $p \le .001$ , d = -1.74). Mean speed in each

429 driving condition, driving environment and speed section is provided in Figure 4.



#### 430

Driving environment and speed section (km/h)

Figure 4. Mean speeds as a function of driving condition (real and simulated) for each driving
environment and speed section. Bars represent standard error of the mean.

434

## 435 **3.2. Eye blink rate**

The repeated measures ANOVA on eye blink rate revealed a significant main effect of 436 the driving condition (F(1,23) = 17.54, p < .001,  $\eta_p^2 = .43$ ), due to more frequent blinks 437 during real driving  $(11 \pm 9.7 \text{ blinks/min})$  than simulated driving  $(4.1 \pm 4 \text{ blinks/min})$ , a 438 significant main effect of the driving environment ( $F(2,46) = 5.86, p < .01, \eta_p^2 = .20$ ), and a 439 significant interaction between the two factors (F(1.28,29.37) = 7.81, p < .01,  $\eta_p^2 = .25$ ). 440 Pairwise comparisons for the main effect of the driving environment showed that more blinks 441 were made when driving on highway (M =  $8.7 \pm 9.5$  blinks/min) than when driving in rural 442 443  $(M = 7.1 \pm 8.2 \text{ blinks/min})$  and urban settings  $(M = 6.8 \pm 7.5 \text{ blinks/min})$ , which did not differ. The interaction between driving condition and driving environment (see Figure 5) 444 revealed the same effect of the driving environment as above in the real driving condition. In 445

the simulated driving condition, however, the blink rate did not vary significantly with the
driving environment. Note that the difference between real and simulated driving conditions
was significant in the three driving environments. It should also be mentioned that the eye
blink frequency decreased in the simulated as compared to the real driving condition in 21 out
of 24 participants.



452 Figure 5. Interaction between driving condition (real and simulated) and driving environment
453 on eye blink rate. Bars represent standard error of the mean.
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### 456 **3.3. Response time to the audio-vocal task**

The repeated measures ANOVA on the response time to the audio-vocal task, with 457 driving condition and driving environment as within-participant variables, yielded a 458 significant main effect of the driving condition (F(1,20) = 13.18, p < .01,  $\eta_p^2 = .40$ ). No other 459 main effect or interaction was observed. The response time was higher when driving the 460 simulator (419  $\pm$  65 ms) than when driving for real (380  $\pm$  60 ms). Note that the mean 461 response time increased in the simulated as compared to the real driving condition in 18 out of 462 21 participants. The mean response time as a function of driving condition and driving 463 environment is given for information in Figure 6. 464



#### 465

Figure 6. Mean response time (ms) to the audio-vocal task as a function of driving condition
(real and simulated) and driving environment. Bars represent standard error of the mean.

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## 470 **3.4. Subjective workload (NASA-TLX)**

A first *t*-test was run to compare the global score of subjective workload (unweighted 471 472 mean of all six subscales) between real and simulated driving. This analysis revealed a statistically significant difference (t(23) = -3.75, p < .01, d = -.75), with perceived workload 473 higher in simulated (M = 41.9) than in real conditions (M = 29.5). Global score in the 474 simulated condition outmatched the one in the real driving condition in 18 out of 24 475 participants (two participants had the same score in the two conditions). To examine the 476 workload profile and determine whether it changed with driving conditions, a 2 (driving 477 conditions: real, simulated) x 6 (subscales: mental demand, physical demand, temporal 478 demand, performance, effort, and frustration) repeated measures ANOVA was performed on 479 480 the ratings. Results revealed a significant main subscale effect ( $F(5,115) = 15.20, p \le .001$ ,  $\eta_p^2 = .40$ ). Pairwise comparisons indicated that the effect was due to significantly higher 481 ratings of mental demand and performance (inverted scale: the higher the score, the worse the 482 483 performance) compared to physical demand, temporal demand and frustration, and to higher ratings of effort compared to temporal demand and frustration (see Figure 7). 484



486 Figure 7. Workload profile (mean rating for each subscale) as a function of driving condition
487 (real and simulated). Bars represent standard error of the mean.
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### 490 **4. Discussion**

To date, few studies have tracked driving simulator validity using mental workload 491 measures and they have found contradictory findings (e.g., Diels et al., 2011; Johnson et al., 492 2011; Milleville-Pennel & Charron, 2015). In this context, the present study aimed at 493 determining whether mental workload varied between on-road and simulated driving while 494 behavioral measures (speed) did not. To achieve this, we selected a free driving condition 495 expected to induce a good correspondence in speed behavior. If the mental workload level 496 497 differed according to the driving condition, the objective was to examine whether the effect of driving environment on mental workload was similar (relative validity) or not between 498 499 simulated and real driving. The findings confirmed previous studies (e.g., Ahlström et al., 2012), showing that the 500

mean speed over the entire route had a very good correspondence between simulated
(74.8 km/h) and on-road driving (73.6 km/h). However, mental workload level was higher in
the simulator as revealed by scores to the NASA-TLX, response times to the subsidiary task
and eye blink rate.

The similar speed but increased mental workload results in the simulator are consistent 505 506 with the predictions we made from De Waard's model (1996). In real world driving, controlling the vehicle in free driving conditions is automated and requires few attentional 507 resources. In driving simulation, on the other hand, vehicle handling is not automated. This 508 could be explained by the low physical validity of the simulator (e.g., lack of inertial cues) 509 which disrupts the naturalness of the perceptual-motor loop. This disruption may then call for 510 an increase in the attentional resources allocated to vehicle handling to preserve driving 511 performance. In this vein, Sahami and Sayed (2013) argued that when participants start to 512 drive a simulator, they have to adapt their skills to the simulator outputs. Consequently, they 513 need time to transfer their already existing driving-related perceptual-motor skills to operate 514 the simulator appropriately (Ronen & Yair, 2013; Sahami & Sayed, 2010). However, this 515 period imposes a mental effort, which lasts until driving becomes automatic (Sahami & 516 517 Sayed, 2013). As an example, Benedetto et al. (2014) examined mental workload level during the first periods of exposition to a simulator and showed that mental workload level decreased 518 519 with repetitions. In the present study, the non-significant effect of the driving environment on mental workload measures in the simulator suggests that mental workload level did not 520 decrease with time on task; however, this may be confounded with the order of presentation 521 of the environments, the urban one, which was encountered last, being known to increase task 522 demands (Faure et al., 2016; Michaels et al., 2017). One way forward on this topic would be 523 to examine together the adaptation period and mental workload to explore whether or not 524 workload level decreases with the adaptation of the perceptual-motor loop to the simulator. 525 This has significance as the allocation of additional mental resources to a level of driving 526 activity known to be automated may interfere with the allocation of attention toward other 527 driving tasks (e.g., Mackenzie & Harris, 2017; see also Wynne et al., 2019). 528

To broaden these results, the validity of driving simulators and VR systems for studying 529 human behavior has been assessed through different user-centered dimensions (Malaterre & 530 Fréchaux, 2001; Morice et al., 2008; Pinto et al., 2008). However, some studies have pointed 531 out that these dimensions might be differently affected by the properties of a VR system. 532 Morice et al. (2008) argued that behavioral validity may be deteriorated by the system 533 properties, while subjective validity (i.e., the extent to which the simulated experience is 534 subjectively close to the real situation) remains good. Using a virtual ball-bouncing task, they 535 found that bouncing performance deteriorated for small delays while participants reported a 536 conscious perception of system latency for relatively high ones. In riding simulation, Lobjois 537 et al. (2016) showed that visual roll used to elicit the leaning sensation when driving round a 538 bend was preferred to no roll, although large visual roll hindered steering behavior. On the 539 other hand, the combined evaluation of behavioral and psychological dimensions of validity 540 541 also suggests a discrepancy between these two dimensions. As reported earlier, some studies showed that subjective workload ratings may be higher even when vehicle handling is well 542 543 preserved in a simulator (Alm, 1995; Blaauw, 1982; Diels et al., 2011). Here, we showed that behavioral validity showed a good correspondence between on-road and simulated driving, 544 but mental workload was higher in the simulator. These results thus support the idea that for 545 in-depth evaluation of a VR system, user-centered effects are also to be examined at the level 546 of psychological validity, i.e., at the level of the underlying processes. 547

Results also revealed that speed behavior was similar in 6 out of 7 speed sections. In the highway and rural environments, mean speed in each section did not vary between simulated and real driving. In the urban environment, however, participants drove faster in the simulator than in real driving conditions. In other words, speed was similar between both driving conditions in the less demanding environments, but it was higher when driving the simulator in the more demanding environment. This pattern of results is not consistent with previous

findings showing that participants drive faster in the simulator for the easiest driving 554 situations (Bella, 2008; Bham et al., 2014; Bittner et al., 2002; Branzi et al., 2017). Nor is it 555 consistent with other models which predict that reducing speed is a way to cope with task 556 difficulty (e.g., Fuller, 2005). A potential perspective that could account for these results is a 557 combined effect of increased workload and range of speed. Regarding the workload effect, it 558 is well known that an increase in cognitive workload leads to visual tunneling and reduces 559 glances to dashboard information and mirrors (Engström et al., 2005; Recarte & Nunes, 2000, 560 2003; Reimer, 2009; Reimer et al., 2012; Tsai et al., 2007; Victor et al., 2005). It may then be 561 that increased workload induced by driving a simulator leads to fewer fixations on the 562 speedometer. Regarding the speed range effect, Wu et al. (2017) proposed that the perception 563 of speed might be biased differently according to the range of speed. They showed that 564 participants who passively observed video footage from a driver's perspective overestimated 565 566 low speeds (range from 8 to 40 km/h) but underestimated high speeds (range from 64 to 96 km/h). In a recent study by Hussain et al. (2019), participants had to actively drive a simulator 567 and drive in the field and estimate when they reached the required speed (50, 70, 80, or 100 568 km/h). Without having access to the speedometer, participants, in both driving conditions, 569 underestimated their own speed and so drove faster than the 50 km/h requested speed; 570 however, they drove slower than the 100 km/h requested speed. Interestingly, when they had 571 access to the speedometer, participants drove slower than the 50 km/h requested speed in the 572 field (48 km/h); in the simulator, however, they underestimated their own speed and drove 573 faster than the requested speed (51 km/h). For higher speeds, participants overestimated their 574 own speed in both conditions similarly. Taken together, these results suggest that the possible 575 visual tunneling due to increased workload in the simulator may have led participants to rely 576 577 on speed estimation to select their own speed. As low speeds are underestimated while

actively driving (Hussain et al., 2019), this could explain why participants drove the simulatorfaster than the real car in the urban environment.

Regarding mental workload and the issue of relative validity of driving simulators, 580 different effects of the driving environment were found depending on the mental workload 581 measures. First, the effect of the driving environment on blink frequency was significant in 582 the field. As expected, the blink rate decreased when the complexity of the driving context 583 increased. However, on the simulator, the driving environment was not found to have any 584 effect on the blink frequency. This pattern of results does not support relative validity. With 585 regard to subsidiary task performance, no difference was found deriving from the driving 586 environment either in real or simulated driving. Unlike eye blink frequency, this result 587 supports relative validity. Alternatively, the non-significant effect of the driving environment 588 in real driving conditions may suggest that variations in task demands from one environment 589 590 to another were not sufficient to elicit differences on this variable. This stems from the sensitivity of subsidiary tasks to mental workload variations. These tasks are sensitive to short 591 592 peaks in mental workload (Jahn et al., 2005), unlike other measures that need to integrate raw data over longer intervals. Given free driving conditions with no particular scenario (such as 593 pedestrians stepping out into the road), driving for real in the urban environment may have 594 been carried out without compromising the reaction time task. Beyond this potential limit, the 595 fact that the effects of driving environment on eye blink frequency and response time do not 596 converge in both settings reflects previous studies that concluded that there was (e.g., Reimer 597 & Mehler, 2011) or was not (e.g., Santos et al., 2005) relative validity depending on the 598 mental workload measures. Finally, scores on the NASA-TLX questionnaire may contribute 599 to the issue of relative validity. Galy et al. (2018) highlighted the value of distinguishing 600 601 between the different subscales of the NASA-TLX rather than using only the global score. In addition, scores greater than the midpoint of a subscale (50) indicates substantial workload 602

levels and can be seen as "drivers" of task workload (Greenlee et al., 2018). The analysis
revealed that the global score was mainly driven by scores on the mental demand,
performance and effort subscales. While these scores were above or approached the subscale
midpoint in the simulator, the workload profile was similar in the real and simulated
conditions. Although subjective workload ratings were not collected for each environment,
the similar workload profile in both settings is a reasonable case for overall relative validity of
driving simulation.

### 610 **5. Limitations and conclusion**

It is important to note several limitations to the present study. While we concluded that 611 a low-cost driving simulator has behavioral validity, lateral control measures were not 612 collected. However, these measures are sensitive to the difference between real and simulated 613 driving (see Supplementary Table 1; but see also the recent review by Wynne et al., 2019) and 614 615 are also known to be correlated with mental workload level (e.g., Jamson & Merat, 2005; Faure et al., 2016). If behavioral validity is supported by the present speed behavior data, the 616 617 same may not be true with lateral control measures. The question also arises whether a similar 618 dissociation between behavioral and psychological validity would be observed with a highfidelity driving simulator. Alm (1995) and Klüver et al. (2016) concluded absolute behavioral 619 validity of high-fidelity driving simulators based on lateral control measures, as compared to 620 fixed-base counterparts (but see Hallvig et al., 2013; Harms, 1996). This suggests better 621 attunement of the perceptual-motor loop in high fidelity simulators and it cannot be excluded 622 that mental workload level may be similar in high-fidelity driving simulators and on-road 623 driving. Although Alm (1995) showed that subjective workload ratings were higher in a 624 moving-base simulator than when driving in the field, this topic needs further research. 625 To conclude, the present study renews evidence that in free driving conditions, speed 626 behavior in simulated driving closely matches that in real driving. Nevertheless, workload 627

level, as measured by blink frequency, response time to a subsidiary task and subjective 628 ratings, was consistently higher in the simulator. These results thus suggest that while 629 behavioral validity can be concluded for speed measures, no similar conclusion can be drawn 630 when attentional resources allocated to vehicle handling come into consideration. The 631 allocation of more attentional resources to a level of driving known to be automated may 632 however interfere with other driving tasks. This fact should therefore be taken into account 633 when driving simulators are used to study tactical or strategic aspects of the driving task 634 635 which require attentional resources, or to study drivers interacting with in-vehicle information or assistance systems. A potential follow-up of this study would be to explore whether 636 workload level can be mitigated with the adaptation of the perceptual-motor loop to the 637 simulator. 638

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