

# Impact of interface design on drivers' behavior in partially automated cars: An on-road study

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# 1 Impact of interface design on drivers' behavior in

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# partially automated cars: an on-road study

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9

#### 10 ABSTRACT

In partially automated vehicles, the driver and the automated system share control of the 11 vehicle. Consequently, the driver may have to switch between driving and monitoring 12 13 activities. This can critically impact the driver's situational awareness. The human-machine 14 interface (HMI) is responsible for efficient collaboration between driver and system. It must keep the driver informed about the status and capabilities of the automated system, so that 15 he or she knows who or what is in charge of the driving. The present study was designed to 16 17 compare the ability of two HMIs with different information displays to inform the driver about the system's status and capabilities: a driving-centered HMI that displayed 18 19 information in a multimodal way, with an exocentric representation of the road scene, and a 20 vehicle-centered HMI that displayed information in a more traditional visual way. The impact 21 of these HMIs on drivers was compared in an on-road study. Drivers' eye movements and 22 response times for questions asked while driving were measured. Their verbalizations during

23	the test were also transcribed and coded. Results revealed shorter response times for
24	questions on speed with the exocentric and multimodal HMI. The duration and number of
25	fixations on the speedometer were also greater with the driving-centered HMI. The
26	exocentric and multimodal HMI helped drivers understand the functioning of the system, but
27	was more visually distracting than the traditional HMI. Both HMIs caused mode confusions.
28	The use of a multimodal HMI can be beneficial and should be prioritized by designers. The
29	use of auditory feedback to provide information about the level of automation needs to be
30	explored in longitudinal studies.
31	
32	KEYWORDS
33	automated vehicles, human-machine interface, multimodal interface, mode transition,
34	verbatim, eye tracking
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36	
37	

#### 38 **1. INTRODUCTION**

The scientific literature identifies inattention while driving as the main cause of road accidents, with more than 70% of car crashes involving inattention or fatigue (Dingus, Neale, Klauer, Petersen, & Carroll, 2006; Lemercier, Pêcher, Berthié, et al., 2014; Victor, Dozza, Bärgman, et al. 2015; Wang, Knipling, & Goodman, 1996). In a society where the priority is to substantially increase road safety, automated cars appear to be the solution for reducing the human factor in road accidents. However, as vehicles become more automated, the collaboration between the driver and the system also becomes more complex. Partially 46 automated vehicles can take charge of some parts of the driving activity. The driver and the 47 system therefore have to work as a team, as they share control of the vehicle. For this team 48 to be efficient, its two members need to communicate via a human-machine interface 49 (HMI). Communication can take place via auditory, haptic and visual displays, as well as via 50 controls. The term HMI is used here to encompass all the explicit and implicit forms of 51 communication between the driver and the system. In automated vehicles, the precise role 52 of the driver depends on the state of the automated system. An increasing level of 53 automation means an increasing amount of information to be displayed. Drivers therefore 54 have to be capable of finding the right information, in order to know which role they need to 55 play in the team. The present study investigated how the design of the HMI in partially 56 automated vehicles influences drivers' understanding of the level of automation and the 57 functioning of the system. In the present exploratory study, two vehicles were driven on 58 road to evaluate the impact of two different HMIs in an ecological setting.

59

#### 60 1.1. Automated driving

61 Highly automated cars are expected to arrive on the market within the next few decades. 62 The idea behind automation is that the system replaces the driver in the operational, 63 tactical, and strategic control of the vehicle (Michon, 1985), allowing the driver to devote attentional resources to other activities (e.g., reading, working, or talking to other 64 65 passengers). Six levels of automation have been established, based on the driver's and system's respective degrees of control (SAE, 2018). The present study focused on vehicles 66 67 equipped with Level 1 and Level 2 systems. At Level 1, the driver is responsible for the 68 operational, tactical and strategic control of the vehicle. Level 2 vehicles are classified as 69 partially automated vehicles, as the car can partially perform the driving activity. Relying on

adaptive cruise control (ACC) and lane centering assist (LCA), the system controls the vehicle's longitudinal and lateral movements. However, even if the vehicle drives itself, drivers are not passive, as they must monitor the vehicle's environment and behavior. Operational control is therefore carried out by the system, but the driver is in charge of strategic and tactical control.

75 Humans do not perform well in prolonged monitoring activities. According to Bainbridge 76 (1985), it is very hard for them to maintain their visual attention on tasks in which very little 77 happens for more than 30 minutes. In highly automated vehicles, this can lead to 78 disengagement in the monitoring activity and engagement in secondary tasks (Carsten, Lai, 79 Barnard, Jamson, & Merat, 2012). The problem is that the system requires specific 80 conditions to operate properly. For LCA, the main condition is to have clear lane markings, 81 so if these markings cease to be visible, the LCA may switch itself off. In this scenario, drivers 82 must realize that the automated system has switched itself off and that they therefore have 83 to take over control of the vehicle.

84

#### 85 1.2. Issues with existing Level 2 vehicles

Existing partially automated vehicles appear to have difficulty communicating the correct 86 87 information to the driver. Drivers have been found to be less inclined to look at the road when using the Tesla Model S with Level 2 systems activated than they are when using a 88 89 Level 1 or Level 0 (Gaspar & Carney, 2019). These authors found that drivers potentially 90 over-relied on the system and were not fully aware of the role they still had to play. In a naturalistic study featuring the Tesla Model S, Banks, Eriksson, O'Donoghue, and Stanton 91 92 (2018) observed confusion between the levels of automation. Participants were filmed and 93 recorded as they drove on the road for 40 minutes. The recorded behaviors were then

94 analyzed. At some point, several drivers let go of the steering wheel, in the mistaken belief 95 that they had activated the Level 2 automation. They only realized that they were confusing 96 modes of automation when they consulted the HMI display. Mode confusions can lead to 97 very hazardous situations, if the driver expects the vehicle to behave in a specific way and it 98 does not (Sarter & Woods, 1995). For example, if drivers think that the vehicle is in charge of 99 lateral control, they will not move the steering wheel in a bend, meaning that the vehicle 100 ploughs straight into the side of the road. Banks, Eriksson, O'Donoghue, and Stanton (2018) 101 suggested that one possible reason for these mode confusions is that drivers place too much 102 faith in the system, but hypothesized that the HMI's lack of transparency is the most likely 103 cause.

104

#### 105 **1.3. Mental model**

106 As the relation between the driver and the system changes according to the degree of 107 automatization, so the HMI has to change too. Carsten and Martens (2019) recently 108 established goals that need to be achieved when designing HMIs for automated vehicles. 109 The first goal is to ensure that the driver understands the capabilities of the vehicle and the 110 level of automation that has been activated. To achieve this goal, HMIs need to adapt to the 111 new constraints that automation places on the driving activity. In the present study, we 112 compared two HMIs to evaluate which one came closer to meeting Carsten and Martens (2019)'s first goal. HMIs need to foster accurate mental models and avoid mode confusions. 113 114 Mental models are the representations that humans have of a system's purpose, form, 115 functioning, state, and structure (see Seppelt & Victor, 2020, for a more detailed definition). 116 In the context of automated driving, drivers with an accurate mental model know when the

117 automated system has been activated or deactivated. They also understand which situations 118 are appropriate for using the system. Driving-related mental models are influenced by three 119 main factors: experience, training, and HMI transparency (Endsley, 2017). Experience came 120 into play in Forster et al. (2019)'s study, as the more situations the participants encountered, 121 the more accurate their mental models became. These findings were supported by Strand et 122 al. (2018)'s study, where three purchasers of Level 2 vehicles were interviewed as they 123 familiarized themselves with the system. Results revealed that although they refined their 124 mental models, their primary representations of the system influenced the subsequent 125 formation of their mental models. Blömacher et al. (2020) reached the same conclusion. 126 They gave correct or incorrect information to drivers regarding the functioning of the system 127 before they used a Level 3 vehicle. Results revealed that the veracity of the information 128 given at the beginning influenced the formation of mental models and reaction time at take-129 overs. However, although experience with an automated system allows individuals to form accurate mental models, it can sometimes lead them to disengage from the monitoring task 130 131 (Solís-Marcos & Kircher, 2019). These results highlight the influence of training and 132 experience on the formation of mental models. Moreover, they support the idea that the 133 driver's first impression of the system is crucial. The information displayed by the HMI needs 134 to clearly be understood, so that the driver forms an accurate mental model at the very 135 beginning of the interaction. To be understood, the HMI must be tailored to the human's 136 cognitive limits and to the specificity of driving an automated vehicle. Several cognitive 137 models have been built to inform design in transportation and are discussed in the next 138 section.

#### 140 **1.4. Models of driving activity**

141 Michon (1985) proposed a very classic hierarchical categorization of the driving task 142 (operational level, tactical level, and strategic level). Stanton et al. (2001) used this 143 categorization to develop a model of the driving task that focused on the interaction of the 144 driver with the automated system. This model is composed of three subtasks: navigation 145 (choosing the itinerary), control (whether to turn right or left at the next intersection), and 146 hazard identification (avoiding pedestrians, trees, buildings, respecting speed limits). 147 Automated driving consists in supporting or replacing the human in one or more of those 148 tasks. ACC, for example, replaces the driver in the longitudinal control of the car. Mental 149 workload is supposed to decrease as the human has fewer tasks to perform, but appears to 150 stay stable with automation (Stanton et al., 2001). This is because although automation 151 replaces the driver, reducing the number of driving-related tasks, it increases the number of 152 tasks related to monitoring the system's activity, meaning that the mental workload remains 153 the same. This issue could be addressed through the design of HMIs, which are not 154 necessarily adapted to the changes in the driving activity. Interfaces designed to help drivers 155 perform their tasks and which take their cognitive limits into account might be more 156 effective.

157

#### **158 1.5. Design solution to improve interaction with the automated system**

When driving, humans have to process a large amount of visual information from both the cockpit and the external environment. One model that is relevant to HMI design because it takes account of humans' cognitive limitations is the multiple resource model developed by 162 Wickens (2008). Currently used in the design of interfaces for complex environments that 163 require simultaneous information processing, this model can predict the allocation of 164 attentional resources to tasks performed simultaneously according to their qualitative 165 characteristics. For the same amount of information, tasks in different sensory modalities (e.g., visual and auditory) are performed better than tasks in a single sensory modality. 166 167 Accordingly, HMIs that use multiple sensory modalities to communicate should elicit a more 168 effective distribution of attentional resources. Zhang et al. (2019)'s meta-analysis showed 169 that HMIs providing a combination of visual and auditory or tactile information allow take-170 over time to be reduced in Level 3 vehicles. HMIs using two or more sensory channels to 171 inform the driver on the state of automation help the latter understand which actions have 172 to be performed. Use of a similar HMI in Level 2 vehicles could also improve interaction 173 between the automated system and the driver. 174

#### 175 **1.6. Research question and hypotheses**

176 The number of studies carried out on open roads with Level 2 vehicles is increasing (Banks, 177 Eriksson, O'Donoghue, & Stanton, 2018; Endsley, 2017; Solís-Marcos & Kircher, 2019). So far, 178 studies have investigated the impact on the driver of driving an automated vehicle at a 179 subjective or cognitive level. However, none of them have compared different HMI designs, 180 even though some HMIs may be safer than others. According to the multiple resource model 181 and previous studies of Level 3 vehicles (Wickens, 2008; Zhang et al., 2019), a driving-182 centered HMI that takes the limitations of the human cognitive system into account by 183 providing information via multiple resource channels (i.e., visual and auditory) should be 184 more easily understood and elicit more accurate mental models, compared with a classic 185 visual-only vehicle-centered HMI. We therefore asked whether a multimodal HMI centered 8

186	on the driving activity and the limitations of the human cognitive system is more efficient
187	than a visual HMI centered on the state of the advanced driver-assistance system (ADAS). By
188	comparing the on-road usage of two existing vehicle HMIs, we sought to identify which HMI
189	characteristics help drivers to understand the capabilities and automation status of their
190	vehicle. We hypothesized that compared with the classic visual vehicle-centered HMI, an
191	exocentric and multimodal driving-centered HMI allows the driver to have a better
192	understanding of how the vehicle operates when the automated system is activated. This
193	translates as better retrieval of information about the system, fewer visual fixations on the
194	cluster, and fewer mode confusions. To test this hypothesis, participants drove one of two
195	vehicles equipped with HMIs that fitted our description (i.e., driving-centered HMI vs.
196	vehicle-centered HMI) on an open road while using automated systems.
197	

#### 198 2. MATERIAL & METHOD

#### 199 2.1. Participants

200 We recruited 20 volunteers (19 men) aged 27-59 years (*M* = 40.80, *SD* = 8.53). They had held 201 a driving license for a mean period of 21 years. The volunteers had no visual, even corrected 202 (myopia, astigmatism, presbyopia), or auditory impairment. They were recruited among 203 employees at Renault Group's Technocentre site in France. They were not paid, and they all 204 signed an informed consent form. These volunteers were randomly assigned to one of the 205 two vehicles (10 participants for each vehicle). The main criterion for recruitment was to 206 have never driven the vehicle to which they were assigned. Seven participants had already 207 driven a Level 1 vehicle before the experiment: two in the vehicle-centered HMI group, and 208 five in the driving-centered HMI group. None of the participants had driven Level 2 vehicles

209 before. All participants were familiar with an automatic gearbox and cruise control.

- 210
- 211 **2.2. HMIs and embedded systems**

212 We selected two vehicles with similar types of driving assistance (ACC and LCA) but different 213 HMIs. The HMI in the first car provided an exocentric representation of the road scene and a 214 multimodal display interface. It displayed relevant information for driving a partiallyautomated vehicle and took the limitations of the human cognitive system into account 215 216 (Blömacher et al., 2020; Forster et al., 2019; Strand et al., 2018). This HMI is therefore 217 referred to hereafter as the driving-centered HMI. The HMI in the second car only provided 218 visual information about the state of the ADAS, and is therefore referred to hereafter as the 219 vehicle-centered HMI. These HMIs differed on three major points: 1) the representation of 220 the information regarding driving assistance; 2) the location of this information; and 3) the 221 modalities used. The driving-centered HMI represented information in an exocentric form 222 (Tesla Model S, software version 8.0). The detection of road markings and other vehicles was 223 indicated on the cluster. Icons representing the activation of Level 1 or Level 2 automation 224 were displayed at the top of the cluster. The visual feedback was supplemented with 225 auditory signals. When the Level 2 automated system (ACC + LCA) was activated or 226 deactivated, a sound was emitted. The sound of activation was a two-note rising tone. The 227 sound of deactivation was a two-note falling tone. By contrast, the vehicle-centered HMI 228 only provided visual feedback (Volvo XC60). Pictograms representing the activation of 229 automation were located on the cluster and the head-up display (HUD). The steering wheel 230 was slightly stiffer at Level 2 than at Level 1 for both vehicles, but the strength needed to 231 override the lateral control was slightly greater for the driving-centered HMI than for the 10

232 vehicle-centered HMI. A description of the technical features of the two HMIs is provided in 233 Table 1. Both vehicles required specific conditions to activate their automated system. Two 234 clear road markings were necessary for the activation of Level 2. The driving-centered HMI 235 indicated the availability of the automation by highlighting the road marking representation 236 in gray on the cluster. If the system could no longer detect one of the road markings, it 237 switched itself off. The main reason for deactivation was failure to detect a road marking. 238 The one major difference between the two types of automation was that the Tesla was able 239 to perform automatic lane changes. When Level 2 was activated, activation of the turn 240 signals automatically resulted in a lane change.

241

Table 1: Display of onboard system activation by the driving-centered HMI and vehiclecentered HMI. The HUD of the vehicle-centered HMI is shown on the righthand side.

		Driving-centered HMI	Vehicle-centered HMI
No assisted s	systems	Detected vehicle	4 <sup>4</sup> m 10 120 140 10 120 10
Display of activated systems	ACC (Level 1)	ACC icon	60       140       160       130         40       100       200       160         20       230       230       160         ACC icon       ACC icon

	ACC + LCA (Level 2)	ACC icon	€0 €0 €0 €0 €0 €0 €0 €0 €0 €0
Activation	ACC (Level 1)	1 pull for ACC	ACC steering- wheel button
commands	ACC + LCA (Level 2)	2 pulls for LCA A sound is emitted	LCA steering- wheel button



- 246
- Figure 1: Photos showing the instrument panel and forward views of each vehicle with either
- 248 the driving-centered HMI (left) or vehicle-centered HMI (right). The letter A indicates the icon
- 249 displayed when LCA and ACC are activated.

250 **2.3. Task** 

251 Participants had to drive the vehicle they were assigned on a pre-established circuit. They 252 were instructed to drive normally, as though they were driving their own car. They were 253 encouraged to use the automated system whenever they wanted and whenever they felt 254 safe. They were instructed to respect the speed limits and traffic laws. A satnav system 255 indicated the directions both visually (on the central multimedia screen) and verbally in each 256 vehicle. During the drive, the experimenter asked the driver questions. The questions were 257 asked when the driver was cognitively available (e.g., while driving along a straight road), so 258 as not to increase cognitive load. Drivers had to respond as spontaneously and sincerely as 259 possible. These questions were designed to evaluate the drivers' understanding of the 260 vehicle's status and functioning.

261

#### 262 **2.4. Road circuit**

263 The road circuit was a 45-minute round trip along public roads and highways, at an average 264 speed of 75.75 km/hr. First, participants drove along a divided highway for 4 km. Then came 265 a 9-km stretch of motorway (four lanes of traffic travelling in each direction for 4 km, then 266 three lanes in each direction for 5 km), followed by a further 11 km on a divided highway. 267 They then turned round at a roundabout, and followed the same itinerary in the opposite 268 direction until they reached their starting point. For the entire round trip, the two sides of 269 the road were separated from each other by a central reservation, and the road was mostly 270 straight or slightly curved.

#### 272 **2.5. On-road questioning**

273 Participants were asked on-road questions to assess their awareness of the functioning and 274 level of automation of the vehicle they were driving. To answer the questions, drivers either 275 had to look for the requested information in their environment or else retrieve it from 276 memory. The questions covered four topics. The first type of question (system questions) 277 was system-oriented and concerned the status and functioning of the vehicle (e.g. "What is your current speed?" or "Is the ACC activated?"). The second type of question (personal 278 279 questions) covered personal information about the driver (e.g. "Do you have children?"). The 280 third type of question (*interior questions*) covered information accessible inside the car (e.g. 281 "What radio station are we listening to?"). The last type of question (*exterior questions*) 282 covered information accessible outside the car (e.g. "Is the car behind us black?") (see 283 Appendix 1 for the list of questions). The system questions were mixed with the other types 284 of questions, so that participants were not able to prepare their answers. The order of the questions was randomized. The answers to the questions were recorded. The start and end 285 286 times of each question and each answer were extracted with Audacity software for the 287 vehicle-centered HMI group, and BeGaze software for the driving-centered HMI group.

288

#### 289 **2.6. Eye tracker**

290 Our choice of eye-tracking technique took into consideration the areas fixated by drivers 291 during the experiment while answering the questions. This measure allowed us to ensure 292 that drivers knew where to find the required information. We used SensoMotoric 293 Instruments (SMI) Eye Tracking Glasses, which are equipped with infrared sensors to 294 monitor eye movements (saccades, fixations and blinks) and a front camera to record the 295 field of vision. The eye-tracking data were recorded at a sampling frequency of 60 Hz. The 294 14

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296	glasses were connected to a mobile phone (Samsung Galaxy Note 4) that allowed us to
297	power the glasses, calibrate the gaze measures, monitor the visual behavior in real time, and
298	store the video and audio recordings. Eye-tracking data were extracted and processed using
299	BeGaze 3.7 software. We also used this software to map the fixations. This mapping
300	consisted in associating each recorded fixation with an area of interest (AOI). BeGaze
301	software then calculated the gaze count and duration for each AOI. There is no explicit
302	definition of a gaze in the BeGaze software, but we worked on the principle that a gaze is a
303	fixation on an AOI if it remains within a 100-pixel area for around 20 ms.
304	

### 305 2.7. Drivers' comments

306 During the tests, participants were free to comment on their driving experience. The 307 comments made during the entire experiment were recorded with the eye-tracker's 308 microphone. As the microphone was placed on the bridge of the glasses not far from the 309 mouth, the audio recording was good enough to hear the participant's comments. These 310 comments were transcribed and classified. This method served to complement the eye-311 tracking and on-road questioning measures. The goal of the analysis was to identify mode 312 confusions, which was not possible with the quantitative data. It also served to collect 313 information about which aspects of the HMI allowed drivers to understand the automated 314 systems.

315

#### 316 **2.8. Procedure**

Participants began the experiment by filling out a questionnaire. The training phase thenbegan. During this phase, the basic controls were explained to the participants, as were

319 specific aspects of the activation and deactivation of the automated system. The driver's 320 responsibilities regarding the use of each assistive device were explained to participants. 321 Afterwards, the participants were equipped with the eye-tracking glasses. The volunteers 322 then drove the vehicle for about 10 minutes on a straight road. Once the experimenter was 323 sure that the volunteers were able to control the car and its features, the experimental 324 phase began. Participants drove for about 45 minutes, following the pre-established circuit 325 on the satnay. The experimenter sat in the front passenger seat and asked the participant 326 questions without disturbing him or her. Finally, the experiment ended with a questionnaire 327 and an interview.

328

#### 329 2.9. Experimental design

A 2 (between-participants) x 4 (within-participants) experimental mixed design was used. The first factor we manipulated was the driven vehicle, and it had two modalities (drivingcentered HMI vs. vehicle-centered HMI). The second manipulated factor was type of question, and it had four modalities (system questions vs. personal questions vs. interior guestions vs. exterior questions).

335

#### 336 2.10. Measures & analysis

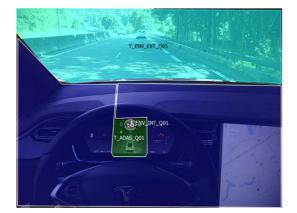
#### 337 **2.10.1. Response times to on-road questions**

For the audio recordings of the drivers answering the on-road questions, we began by extracting the start and end times of each question and each answer. To measure the time needed by participants to access the information and verbalize it, we calculated the interval between the end of the question and the end of the answer. The assumption of normality of

- residuals for mean response times was not met for the driving-centered HMI group. We ran
- 343 a mixed variance analysis on means of median response times, with type of questions as a
- 344 within-participants factor, and HMI type as a between-participants factor. To investigate
- 345 HMI use in greater depth, the questions were analyzed separately for each system. Given
- 346 the small sample, we ran a two-tailed Mann-Whitney *U* test for each system question, with
- 347 HMI type as a between-participants factor. The Mann-Whitney U test is a nonparametric
- 348 test that allows two small independent samples to be compared, in order to decide whether
- 349 they are equally distributed or not (Mann & Whitney, 1947).
- 350

#### 351 **2.10.2. Fixation count and duration**

Gaze positions were coded on a reference image featuring all the AOIs. We used four or five 352 353 AOIs for each vehicle. For the driving-centered HMI group, the AOIs were the external 354 environment, the interior environment, the speed display, and the ADAS. The vehicle-355 centered HMI had the same AOIs, plus the HUD (see Fig. 2). For each AOI, we extracted the 356 mean fixation duration and number of fixations during an 8-second window, extending from 357 4 s before to 4 s after the end of the question. This window allowed us to consider the 358 processing time (i.e. listening to the question, finding the answer, and verbalizing it). The 359 analyses compared the driving-centered HMI group and the vehicle-centered HMI group on 360 the mean duration and number of fixations. Owing to the small sample and the non-respect 361 of the normality of residuals for the driving-centered HMI group, we ran nonparametric 362 analyses. Nonparametric Wilcoxon rank-sum tests were carried out, with group as a 363 between-participants factor for both measures. The same analyses were carried out 364 separately for each system question on each AOI, except for the HUD. No participants were excluded from the analysis (N = 20). 365





366

Figure 2: AOIs used in the driving-centered HMI (left) and vehicle-centered HMI (right).

368 The HUD displayed the vehicle's speed and information about the automated system (see 369 Table 1). The HUD AOI was too small for us to know for certain whether the fixations were 370 on the speed information or on the automation information. We therefore did not include 371 the HUD in the comparison between the two HMIs. We can, however, assume that during 372 the questions about speed or the automated system, drivers looked at the HUD to obtain 373 the relevant information. To consider the fixations on the HUD, we calculated descriptive 374 statistics. For the vehicle-centered HMI group, we calculated the means and standard errors 375 of fixation count and duration on the HUD, speed display, and ADAS AOIs. The same 376 operation was repeated for the driving-centered HMI group, minus the HUD. We then 377 calculated the mean fixation count and duration for each AOI for the questions on speed. 378 We repeated the same operation for the questions on the automated system.

379

#### 380 **2.10.3. Classification of drivers' comments**

381 The drivers' comments about the HMI in relation to the automated system were transcribed and categorized, using a bottom-up approach. Three categories and 15 subcategories 382 383 emerged. The three categories were (1) activation of ACC (Level 1) or ACC combined with 384 LCA (Level 2), (2) deactivation of ACC (Level 1) or ACC combined with LCA (Level 2), and 3) 385 mode confusions. Each category contained five subcategories. These subcategories 386 corresponded to the communication medium between driver and system: (1) visual 387 information display on the cluster; (2) auditory signals; (3) haptic feedback; (4) kinesthetic 388 sensations; (5) commands; and (6) HUD (only for the vehicle-centered HMI group). The 389 analysis consisted in counting the number of comments in each category and subcategory.

391 **4. RESULTS** 

After indicating the response times for each type of question and each vehicle, we compare the vehicles on the system questions. Next, we describe the analysis of the eye-tracking data, comparing the two groups on fixation count and duration for the system questions. Finally, we describe the classification of the drivers' comments for each vehicle.

396

#### 397 4.1. Response times

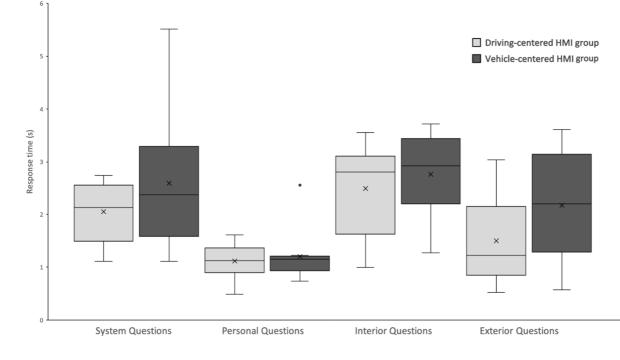
398 We found a difference in the means of median response times between the groups and 399 types of question (see Table 2). The ANOVA revealed an effect of type of question on 400 response times, F(3, 16) = 7.01, p < .001, indicating that participants needed different 401 amounts of time to answer the system questions, personal questions, interior questions, and 402 exterior questions. Less time was needed to answer the system questions than either the 403 personal questions, interior questions, or exterior questions. Regarding the effect of type of 404 HMI on response times for each type of question, the mixed ANOVA revealed no significant 405 effect, F(3, 16) = 3.53, p = .077, d = .67. This means that participants took more or less the 406 same time to answer each type of question regardless of which group they belonged to (see 407 Fig. 3). No interaction effect was observed between type of question and group (p > .1). For 408 the system questions, groups differed slightly on the means of median response times (see 409 Table 2). However, this difference was not significant (p > .1). Regarding the personal 410 questions, we found a slight difference between the groups, but tests indicated that it was 411 not significant (p > .1). This suggests that personal questions were not influenced by type of 412 HMI. For the interior questions, there was a small difference in mean response times, as well 413 as in standard deviations. However, tests indicated that these differences were not 414 significant (p > .1). For the exterior questions, we found a substantial difference. Driving-20

- 415 centered HMI drivers seemed to answer questions about the external environment quicker 416 than vehicle-centered HMI drivers did. Again, however, tests indicated that this difference 417 was not significant (p > .1). We then compared the two groups on response times for each 418 system question.
- 419
- 420

421 Table 2: Mean (standard deviation) response times depending on HMI and question type.

	System	Personal	Interior	Exterior
	Questions	Questions	Questions	Questions
Driving-	1.61 s	1.17 s	2.27 s	1.50 s
centered	(0.46)	(1.08)	(0.84)	(0.86)
HMI group				
Vehicle-	2.16 s	1.13 s	2.55 s	2.17 s
centered	(1.50)	(0.41)	(1.04)	(0.74)
HMI group				

422





423

Figure 3: Boxplot of response times depending on type of question and HMI group.

425

## 426 4.2. System questions

427 The Mann-Whitney tests on each system question revealed significant differences between 428 the two groups (see Fig. 4). For the question "What is your current speed?" when the 429 automated system was active, analysis revealed a significant difference between the groups, U = 14, p = .02, r = -0.55. On average, driving-centered HMI users gave their speeds faster 430 431 (*Mdn* = 1.15) than vehicle-centered HMI users did (*Mdn* = 1.82). For the question "What is your current speed?" when the automated system was inactive, analysis revealed no 432 significant effect, U = 6, p = .07, r = -0.51. Driving-centered HMI users responded slightly 433 434 faster (*Mdn* = 1.43) than vehicle-centered HMI users did (*Mdn* = 2.64). Analysis of responses 435 to the other system questions failed to reveal any significant differences (p > .1; see Table 3). 436

Table 3: Sample sizes and median (interquartile range) response times depending on the question and the HMI. \* p < .05.

System questions	Driving-centered	Vehicle-centered	Test results
	HMI group	HMI group	
"What is your current	<mark>(n = 10)</mark>	<mark>(n = 8)</mark>	U = 14, p =
speed?"	1.08 s	1.53 s	.02, <i>r</i> =55
(Level 2 on)	(0.74) (1.01)		
		*	
"What is your current	( <i>n</i> = 9)	( <i>n</i> = 4)	U = 6, p = .07,
speed?"	1.11 s	2.61 s	<i>r</i> =51
(Level 2 off)	(0.08)	(1.91)	

"Are the ACC and LCA	( <i>n</i> = 10)	( <i>n</i> = 9)	<i>U</i> = 30, <i>p</i> =
activated?" (Level 2 on)	0.49 s	0.56 s	.24, <i>r</i> =28
	(0.41)	(1.90)	
"Are the ACC and LCA	( <i>n</i> = 9)	( <i>n</i> = 9)	<i>U</i> = 43, <i>p</i> =
activated?" (Level 2	1.78 s	1.09 s	.86, <i>r</i> =05
off)	(2.05)	(1.49)	
"Is the ACC activated?"	( <i>n</i> = 10)	( <i>n</i> = 10)	U = 36, p = .32, r =24
	0.62 s	1.06 s	.32,124
	(1.05)	(2.59)	
"Is the LCA activated?"	( <i>n</i> = 9)	( <i>n</i> = 8)	U = 25, p =
	0.76 s	1.05 s	.32, <i>r</i> =26
	(1.14)	(0.60)	

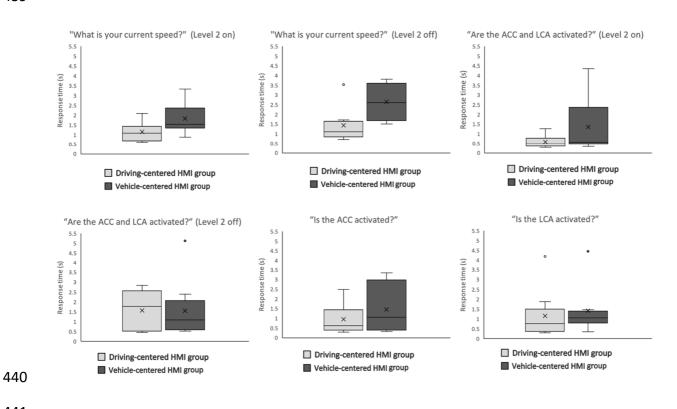


Figure 4: Boxplot of response times depending on system question and HMI group.

443

#### 444 **4.3. Eye tracking**

445 The subsequent analyses served to compare the two groups on each measure, each AOI, and 446 each system question. Because of the large number of analyses (more than 50), only the 447 significant effects and those tending toward significance are reported here. For the question 448 "What is your current speed?" when the ACC and LCA were active, no significant difference 449 was observed between the groups on the mean fixation time on the speedometer, U = 17, p = .06, r = -0.449 (see Table 4). Driving-centered HMI users made slightly longer fixations on 450 451 the speedometer (*Mdn* = 294 ms) than vehicle-centered HMI users did (*Mdn* = 75 ms; see 452 Fig. 5). For the question "What is your current speed?" when in manual driving mode, 453 analysis revealed an effect of group on the number of fixations on the speedometer, U = 3, p 454 = .02, r = -0.656. Driving-centered HMI users looked at the speedometer more often (*Mdn* = 455 2) than vehicle-centered HMI users did (Mdn = 0). Groups did not differ significantly on 456 mean fixation time for this question, U = 5.5, p = 07, r = -0.524. On average, driving-centered 457 HMI users made slightly longer fixations (Mdn = 523 ms) on the speedometer than vehicle-458 centered HMI users did (*Mdn* = 0 ms). For the question "Is the ACC activated?" when it was 459 activated, the test revealed no significant effect on the number of fixations on the ADAS AOI, U = 20.5, p = .07, r = -0.422. Driving-centered HMI users tended to look slightly more often 460 461 (Mdn = 6) at the ADAS than vehicle-centered HMI users did (Mdn = 0). A significant 462 difference was observed for the number of fixations on the interior of the vehicle for the 463 same question, U = 17, p = .03, r = -0.422. Vehicle-centered HMI users (*Mdn* = 1) made more 464 fixations on the interior of the vehicle than driving-centered HMI users did (Mdn = 0). In

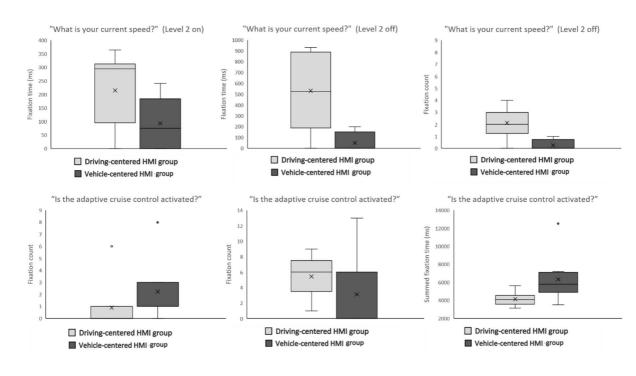
addition, there was a significant difference between the groups on mean fixation duration outside the vehicle for the same question, U = 11, p = .009, r = -0.614. Vehicle-centered HMI users (*Mdn* = 5793 ms) looked outside the vehicle for significantly longer time periods than the driving-centered HMI users did (*Mdn* = 4066 ms).

469

Table 4: Sample sizes, median (interquartile range) numbers and mean durations of fixations depending on the vehicle, and results of nonparametric analysis. Significant differences are in italics. \* p < .05. \*\* p < .01.

Question	AOI	Measure	Driving-	Vehicle-	Test results
			centered HMI	centered HMI	
"What is	Speedometer	Fixation	<mark>(n = 9)</mark>	<mark>(n = 8)</mark>	<mark>U = 17</mark>
your		duration	294 ms	75 ms	<mark>p = .06</mark>
current			(194)	(171)	<mark>r = -0.45</mark>
speed?"					
(Level 2 on)					
"What is	Speedometer	Number	( <i>n</i> = 8)	( <i>n</i> = 4)	<mark>U = 3</mark>
your		of	2 (2.5)	0 (1.25)	<mark>p = .02</mark>
current		fixations	*		<mark>r = -0.66</mark>
speed?"					
(Level 2 off)		Fixation	( <i>n</i> = 8)	( <i>n</i> = 4)	<mark>U = 5.5</mark>
		duration	211.6 ms	0 ms	<mark>p = 07</mark>
			(150)	(171)	<mark>r = -0.52</mark>

"Is the	ADAS	Number	( <i>n</i> = 9)	( <i>n</i> = 9)	<mark>U = 20.5</mark>
adaptive		of	6 (3)	0 (4)	<mark>p = .07</mark>
cruise		fixations			<u>r = -0.42</u>
control	Interior of the	Number	( <i>n</i> = 9)	( <i>n</i> = 9)	<mark>U = 17</mark>
activated?"	vehicle	of	0 (3)	1 (4)	<mark>p = .03</mark>
		fixations	*		<u>r = -0.42</u>
	External	<mark>Summed</mark>	( <i>n</i> = 9)	( <i>n</i> = 9)	<mark>U = 11</mark>
	environment	fixation	4066 ms	5793 ms	<mark>p = .009</mark>
		duration	(63)	(114)	<mark>r = -0.61</mark>





475 Figure 5: Boxplots of fixation counts and durations depending on the question and the HMI

476

group.

For the vehicle-centered HMI group, the number of fixations on the HUD and their mean duration were non-negligible (see Table 5). The number and duration of fixations on the HUD were the same as for the speedometer (for speed questions) and the ADAS (for automated system questions). This suggests that the HUD was used just as much as the speedometer and the ADAS to answer system questions.

483

Table 5: Mean (standard deviation) fixation count and duration on the HUD, speedometer
and ADAS AOIs of the vehicle-centered HMI and driving-centered HMI groups for each type
of question.

		Vehicle	e-centered HMI	Driving-centered HMI		
Question	AOI	Fixation	Mean fixation	Fixation	Mean fixation	
		count	duration	count	duration	
Speed	Speedo	3.20	146 ms (71)	2.40 (1.06)	227 ms (118)	
questions	meter	(2.95)				
	HUD	4.25	280 ms (117)			
		(2.25)				
System	ADAS	4.91	201 ms (124)	2.36 (1.47)	183 ms (103)	
questions		(3.61)				
	HUD	3.20	336 ms (289)			
		(1.67)				

#### 488 **4.4. Drivers' comments**

489 Every participant made spontaneous comments about the driving experience, HMI, and 490 functioning of the automated system. These comments are described separately for each 491 vehicle. We divided them into three categories: (1) activation of automated system, (2) 492 deactivation of automated system, and (3) mode confusions. We counted the number of 493 comments in each category. Most of the vehicle-centered HMI users referred to the cluster 494 to comment on the activation of the automated system (See Table 6). Four of the 10 vehicle-495 centered HMI users referred to the HUD. One vehicle-centered HMI user referred to the 496 commands, and four to the haptic feedback. Finally, four participants in the vehicle-centered 497 HMI group reported relying on kinesthetic sensations. For the deactivation of the automated 498 system, three of the 10 vehicle-centered HMI users mentioned the commands. Only one 499 vehicle-centered HMI user reported using the cluster. The most frequently encountered 500 mode confusions for the vehicle-centered HMI drivers were between Levels 1 and 2. Three 501 vehicle-centered HMI users became confused because of unclear information on the cluster. 502 One vehicle-centered HMI user reported uncertainty or mode confusion between manual 503 driving and ACC (Level 1) when consulting the visual information on the cluster. Another 504 user reported mode confusion between manual driving and the ACC (Level 1) when using 505 the commands.

506

507 Table 6: Number of participants in the vehicle-centered HMI group (total number of 508 participants in the group) who commented about a specific modality of the HMI regarding 509 activation, deactivation, or mode confusions.

Activation	of	Deactivation	of	Mode	Mode

	automated	automated	confusion	confusion
	system	system	between	between Level
			manual driving	1 and Level 2
			and Level 1	systems
			systems	
Visual	8 (10)	1 (10)	1 (10)	3 (10)
Commands	1 (10)	3 (10)	1 (10)	0
Haptic	4 (10)	0	0	0
feedback				
Kinesthetic	4 (10)	0	0	0
sensations				
HUD	4 (10)	0	0	0

511 Five of the 10 users of the driving-centered HMI reported relying on the visual modality of 512 the cluster to know if the automation was activated. Only one participant reported using the 513 auditory feedback. Two participants referred to the commands. One participant referred to 514 the haptic feedback, while another referred to kinesthetic sensations. As for deactivation of 515 the automated system, two of the 10 driving-centered HMI users relied on visual 516 information displayed on the cluster, two referred to the commands, two to the auditory 517 feedback, and one to the haptic feedback. Three participants experienced confusion 518 between Level 2 (ACC + LCA) and either Level 1 (ACC) or manual driving when consulting the 519 visual information. The same kind of confusion was reported by three participants when listening to the auditory feedback. Confusion between Level 1 and either the Level 2 or 520

521 manual driving was experienced by one driving-centered HMI user when consulting the 522 visual information on the cluster. Confusion between Level 1 and Level 2 was experienced by 523 one participant in the driving-centered HMI group when consulting the visual information. 524 The same kind of confusion was reported by one riving-centered HMI user who relied on 525 kinesthetic sensations.

526

527 Table 7: Number of participants in the driving-centered HMI group (total number of 528 participants in the group) who commented about a specific modality of the HMI regarding 529 activation, deactivation, or mode confusions.

	Activation	Deactivation	Mode	Mode	Mode
	of	of automated	confusion	confusion	confusion
	automated	systems	between	between	between
	systems		Level 2 and	Level 1 and	Levels 1
			Level 1 or	Level 2 or	and 2
			manual	manual	
			driving	driving	
Visual	5 (10)	2 (10)	3 (10)	1 (10)	1 (10)
Auditory	1 (10)	2 (10)	3 (10)	0	0
Commands	2 (10)	2 (10)	0	0	0
Haptic	2 (10)	1 (10)	0	0	0
feedback					
Kinesthetic	1 (10)	0	0	0	1 (10)

sensations			

#### 531 **5. DISCUSSION**

532 Our study was designed to explore the impact of HMI design on the detection of information 533 about the status and functioning of two partially automated vehicles. We compared two 534 HMIs: a driving-centered HMI and a vehicle-centered HMI. The driving-centered HMI used an 535 exocentric representation of the road scene and multimodal information to inform drivers 536 about the level of automation. The vehicle-centered HMI used a traditional display with 537 visual information only. During a 45-minute on-road driving session, response times and eye 538 movements were recorded while participants answered questions. Spontaneous comments 539 were also recorded, and subsequently classified according to whether they were related to 540 the HMI or the status of the automated systems. In this section, we discuss the extent to 541 which our results answer the question of whether the exocentric and multimodal driving-542 centered HMI allows drivers to have a better understanding of the functioning of the vehicle 543 when using an automated system than a classic visual vehicle-centered HMI.

544 Overall, response times for system questions were shorter for the driving-centered HMI 545 group than for the vehicle-centered HMI group. More specifically, the two groups differed 546 on response times for questions about vehicle speed. When the automation was on, driving-547 centered HMI users responded faster than vehicle-centered HMI users. These results suggest 548 that vehicle speed information is more accessible in the driving-centered HMI than in the 549 vehicle-centered HMI. However, we failed to find any differences in response times to 550 questions about the activation of ACC (Level 1) or ACC + LCA (Level 2). Compared with the 551 classic HMI, the multimodal and realistic driving-centered HMI seemed to inform drivers 552 more efficiently about the functioning of the vehicle, though not about the level of 553 automation. Regarding the eye-tracking data, for the speed questions, the driving-centered 554 HMI group made more/longer fixations than the vehicle-centered HMI group on the 555 speedometer. At first glance, our results suggest that the exocentric HMI captured the 556 drivers' visual resources more than the classic HMI did when it came to vehicle speed. This 557 suggests that it may have been difficult to extract visual information (Jacob & Karn, 2003). 558 However, we observed a high fixation count and duration on the HUD, such that the vehicle-559 centered HMI drivers fixated the HUD at least as many times and for at least as long as they 560 did either the speedometer or the ADAS. The fact that participants looked at both the 561 speedometer and the HUD when answering the speed question suggests that they double-562 checked. They may have seen the information on the HUD and verified it on the 563 speedometer. The time needed to perform this double checking may explain why it took the 564 vehicle-centered HMI users longer to answer questions about vehicle speed. The use of the 565 HUD in this situation may have been more demanding than that of the driving-centered 566 HMI. One way to measure the benefit of the HUD would be to compare the answers of 567 vehicle-centered HMI drivers to speed questions in two HUD conditions: activated versus 568 deactivated. This would make it possible to evaluate the HUD's impact on response times. 569 Regarding ACC activation (Level 1), results diverged. Driving-centered HMI users looked

570 more often at the ADAS, where the information was displayed. By contrast, vehicle-centered 571 HMI users looked more often at the interior of the vehicle, suggesting that they were looking 572 for the information in the wrong places. However, they also looked at the external 573 environment for longer periods than the driving-centered HMI users, which suggests better 574 situational awareness. We had expected driving-centered HMI users to make shorter and 575 fewer fixations on the cluster than the vehicle-centered HMI users when answering 33 questions on automation. The multimodal driving-centered HMI provided auditory feedback about the level of automation, meaning that users could access the information without having to look at the cluster. However, we found no significant differences in fixations compared with the classic HMI. This suggests that the multimodal HMI did not correctly inform drivers about the status and functioning of the vehicle. This result was confirmed by the mode confusions encountered by the driving-centered HMI group.

582 We observed a wider range of mode confusion situations for the driving-centered HMI group 583 than for the vehicle-centered HMI group (see Tables 6 and 7). For the driving-centered HMI 584 users, most of the confusions were related to Level 2 (ACC + LCA), and involved both visual 585 and auditory information. Thus, instead of helping drivers to distinguish between the modes, 586 the multimodal HMI actually induced confusion in some of them. This may be an indication 587 that the auditory information was not correctly perceived or understood by users. In the 588 vehicle-centered HMI group, most of the mode confusions concerned Levels 1 (ACC) and 2 (ACC + LCA). There was only a subtle distinction between these two levels on the cluster, 589 590 which may explain these confusions. Fewer confusions related to the ACC were observed for 591 the driving-centered HMI. The use of a realistic representation of the vehicle being followed 592 may be an efficient clue, helping drivers to understand the activation status of the ACC 593 (Level 1).

According to the multiple resources model, using different sensory channels to convey information is more efficient than using a single one (Wickens, 2008). According to the automated driving model, an HMI that informs drivers about the tasks undertaken by the automated systems is more efficient (Stanton et al., 2001). On the basis of these two models, we had assumed that the multimodal and exocentric driving-centered HMI would allow drivers to understand the vehicle's status and capabilities more efficiently than the

600 vehicle-centered HMI would. Despite the presence of an HUD in the vehicle-centered HMI, 601 the driving-centered HMI allowed drivers to respond faster to questions about the vehicle's 602 level of automation. Interestingly, however, the driving-centered HMI captured more visual 603 attention overall than the vehicle-centered HMI did. This result is consistent with the results 604 of Gaspar and Carney (2019). In their study, participants used a Tesla Model S (driving-605 centered HMI) to do daily drives, and their eye movements were recorded. The authors 606 found that users looked at the external environment less frequently when Level 2 was 607 activated. Our results corroborate these findings, suggesting that an exocentric 608 representation tends to be visually distracting. Regarding mental models, the exocentric and 609 multimodal driving-centered HMI seemed to induce mode confusions. Banks et al. (2018) 610 observed similar results with participants using a Tesla Model S with an HMI similar to the 611 one we used in our study. These authors reported mode confusions with the activation of 612 Level 2 automation. Mode confusion with the activation of Level 2, specifically relating to 613 visual information, was also reported in the present study. Interestingly, however, we 614 observed more mode confusion with the deactivation of Level 2. This confusion was related 615 to the auditory feedback intended to inform drivers about the activation or deactivation of 616 the automated system. Participants' verbal comments indicate that some of these sounds 617 were not understood. Carsten and Martens (2019) suggested that an efficient automated 618 vehicle HMI design implies that the functioning of the vehicle can be understood. If the 619 meaning of signals from the HMI is not understood, an accurate mental model cannot be 620 forged. This is particularly important when the signal serves to inform the driver of a return 621 to manual control. Many sounds are emitted nowadays in cars. Most of them are meant to alert drivers of an emergency. Designers should focus on sounds that efficiently inform them 622 623 about the activation level of the automated system.

#### 625 **5.1. Conclusion**

626 The purpose of our study was to compare the characteristics and on-road use of two HMIs. 627 More specifically, our protocol was intended to determine which of the two designs best 628 helped drivers understand the status and functioning of their vehicle. The first HMI was 629 multimodal and featured an exocentric representation of the road scene. The second HMI 630 was more traditional and used codes that are widely used in the automobile industry. The 631 two designs gave rise to similar performances on questions about the status and functioning 632 of the vehicle. However, the driving-centered HMI was more efficient when it came to 633 informing drivers about current speed. Eye movements suggested that its multimodal and 634 realistic representation was more visually demanding. Finally, the driving-centered HMI 635 induced mode confusions related to Level 2, whereas the classic HMI induced mode 636 confusions related to Level 1. Taken together, results indicate that the multimodal driving-637 centered HMI did not help participants driving the vehicle for the first time to gain a better 638 understanding of the functionalities and activation level of the automated system. The mode 639 confusions and failure to use the auditory information suggest that the sounds indicating 640 activation and deactivation of automation could be improved. Future experiments should 641 measure HMI use over an extended period of time.

642

#### 643 **5.2. Limitations**

644 On-road studies have just as many disadvantages as they do advantages. In ecological 645 situations, some factors cannot be controlled. Variables dependent on the surrounding 646 environment can interfere with the protocol. For example, the experiments with the vehicle647 centered HMI took place during vacation periods. The participants who drove with this HMI 648 were therefore more likely to encounter traffic jams. When in a traffic jam, the vehicle's 649 speed is often too low to activate the automation. The vehicle-centered HMI participants 650 therefore used the automated system for shorter periods of time than the driving-centered 651 HMI users did. Another limitation arising from the ecological situation concerns the intrinsic 652 differences between the vehicles. The main difference between the vehicles here was the 653 presence of an HUD in the vehicle-centered HMI but not in the driving-centered HMI. Drivers 654 could obtain several types of information from the HUD, as it centralized information about 655 automation status and speed. Given the broad range of information and the relatively small 656 surface area, we were not able to compare eye-tracking data for this AOI with that of the 657 other AOIs. However, analysis of participants' comments suggested that the HUD was an 658 important source of information on automation status.

659 Another limitation was the small sample size. With only 10 participants per group, caution 660 needs to be exercised when considering the conclusions we drew. Except for one ANOVA, 661 we only carried out nonparametric tests. Several differences were significant, but a larger 662 sample size would have allowed us to increase the statistical power and strengthen the 663 possible effects. This is one reason why we considered the drivers' comments. These 664 supplemented the quantitative data by explaining what the participants experienced. Mode 665 confusions are difficult to reproduce and observe in simulators. They can be accessible if 666 drivers are interviewed, but this is generally done after the experiment, and only rarely while 667 they are driving. Recording and transcribing drivers' comments allowed us to access mode 668 confusions and explain the quantitative data. This exploratory study yielded relevant 669 information about the impact an interface may have on drivers' awareness about mode 670 status. However, it is important to note that this experiment only told us about what

happens while drivers are familiarizing themselves with a vehicle. A 1-hour drive is not long
enough for them to understand every feature of the HMI, and familiarity with automated
systems can have an impact on their utilization (Solís-Marcos & Kircher, 2019).

674

#### 675 5.3. Future research

676 As well as emphasizing the need for the functioning of the vehicle to be understood, Carsten 677 and Martens (2019) underscored the need to carefully calibrate trust. Too much trust leads 678 to disengagement, while too little trust discourages drivers from using the automated 679 system. These authors argued that the exocentric representation of the driving-centered 680 HMI may simply serve to promote trust in the system. It may be esthetically appealing, but 681 possibly not that useful in helping drivers take back control when necessary. It may be more 682 suited to Level 3 vehicles, where the automated system gives drivers more time to take back 683 control. In Level 2 vehicles, drivers need to be ready to do so at all times. The calibration of 684 trust is therefore crucial. Future research should focus on looking for HMIs that foster the 685 right degree of trust. One idea would be to indicate the vehicle's limitations. Beller, Heesen, 686 and Vollrath (2013) explored the idea of warning drivers about the automation's uncertainty 687 in order to bring them back into the loop. In highly automated vehicles, the display features 688 an icon representing uncertainty in situations where the automation may fail. In this 689 condition, drivers are faster at taking back control and have greater trust in the automation. 690 This kind of display could be included in Level 2 vehicles.

Another point to consider in future studies is familiarity with the system. Our study focused
on the first driving session that participants had with the vehicle. A driver who is used to a
Level 2 vehicle should know when the system is able to function properly or not. However,
Solís-Marcos and Kircher (2019) observed a potential effect of familiarity with automation 38

on eye movements. More specifically, the more experienced participants performed more
 secondary tasks, potentially leading to a loss of situational awareness. Future research
 should take experience with the system into account by featuring longitudinal experiments.

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## 786 Appendix: Detailed questions

Target q	uestions
What is	your current speed? (Level 2 on)
What is	your current speed? (Level 2 off)
s the Au	to Pilot activated? (Level 2 on)
s the Au	to Pilot activated? (Level 2 off)
s the ad	aptive cruise control activated?
s the lar	ne centering assist activated?
Persona	questions
Do you ŀ	nave any children/grandchildren?
Do you h	have any pets?
What ar	e their names?
Do you p	lay sports?
nterior	questions
What co	lor is the pen I'm writing with?
What ra	dio station are we listening to?
What co	lor are my clothes?
Exterior	questions
Do you t	hink it will rain today?
s the ve	hicle on our right green?
s the ca	r behind us black?

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#### DECLARATION

As a professional translator for the English language, I hereby declare that I carried out the language editing of the article entitled *Impact of interface design on drivers' behavior in partially automated cars: an on-road study* (N. Monsaingeon et al.).

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