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A Conceptual Model for Persuasive In-Vehicle Technology to Influence Tactical Level Driver Behaviour

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Abstract

Persuasive in-vehicle systems aim to intuitively influence the attitudes and/or behaviour of a driver (i.e. without forcing them). The challenge in using persuasive systems in a driving setting, is to maximise the persuasive effect without infringing upon driver safety.

This paper proposes a conceptual model for driver persuasion at the tactical level (i.e., driver manoeuvring level, such as lane-changing and car-following). The main focus of the conceptual model is to describe how to safely persuade a driver to change his or her behaviour, and how persuasive systems may affect driver behaviour.

First, existing conceptual and theoretical models that describe behaviour are discussed, along with their applicability to the driving task. Next, we investigate the persuasive methods used with a focus on the traffic domain. Based on this we develop a conceptual model which incorporates the behavioural basis and persuasive methods, and which describes how effective and safe driver persuasion functions. Finally, we apply the model to a case study of a lane-specific advice system which aims to reduce travel time delay and congestion by encouraging a better distribution of the vehicles over the available motorway lanes.

1. Introduction

1.1 The Problem and Scope

The way drivers interact with their cars is changing (Damiani, Deregibus, & Andreone, 2009; Ulrich et al., 2013). Modern cars are more and more equipped with advanced driver assistance systems (ADAS) that can assist the driver, as well as in-vehicle information systems (IVIS) that provide the driver with traffic information or driving advice. Increases in IVIS/ADAS in-car systems mean that the driving environment becomes more information rich, and more systems compete for the driver's attention.

One field of development within IVIS is that of persuasive systems. Persuasive systems employ techniques or incentives to voluntarily change drivers' attitudes or behaviours (Fogg, 2010). The implementation of such persuasive systems in the driving environment can for example help reduce speeding and improve driver engagement during monotonous driving (Steinberger, Proppe, Schroeter, & Alt, 2016). Persuasive systems have also been used to encourage drivers to adopt a more eco-friendly driving style (Ecker, Holzer, Broy, & Butz, 2011), or a safer driving style (Shi, Lee, Kurczak, & Lee, 2012).

While persuasive systems can positively influence driver behaviour and increase safety, they might also introduce new risks (van Nes & Duivenvoorden, 2017). For example, the use of these systems can lead to indirect behavioural adaptations (unwanted and unplanned side-effects) (Martens & Jenssen, 2012), such as when the implementation of the anti-lock braking system (ABS) led to reduced headways (Sagberg, Fosser, & Sætermo, 1997). Additionally, increasing the number of in-vehicle systems can negatively influence traffic safety by overloading or distracting the driver at inappropriate times (Reyes & Lee, 2004; M. S. Young, Brookhuis, Wickens, & Hancock, 2015).

To our knowledge, a conceptual model tying driver persuasion to safety and behavioural outcomes has not been developed yet. In this study, we aim to fill this research gap by developing a conceptual model that describes the effects of in-vehicle persuasive systems on driver behaviour, with the goal of effectively and safely persuading the driver. We will focus specifically on IVIS systems aiming at persuading drivers to change their behaviour at the tactical level. Examples of such systems include lane-specific advice to improve traffic flow

50 (Risto & Martens, 2013; Schakel & Van Arem, 2014), and systems that encourage eco-driving with the goal of
51 reducing pollution (Ecker et al., 2011).

52

53 **1.2 Why Target Driver Behaviour at the Tactical Level?**

54 Driver behaviour is often divided into three levels: the strategic, tactical and control level (Evans & Michon,
55 1985). The strategic level considers high-level choices related to driver's route choice behaviour, which is
56 generally constant over longer periods of time. At the tactical level, drivers decide upon and perform
57 manoeuvres (e.g. change lane, take exit, overtake car) considering the observable and anticipated part of the road
58 network to reach their strategic goals. At the control level, the driver performs actions to operate the vehicle (e.g.
59 change gears, press accelerator pedal, turn on blinker).

60 Our conceptual model will focus on safely persuading driver behaviour at the tactical level. From a
61 persuasive perspective, targeting relatively uncomplicated, short-term behavioural responses (e.g. adjusting
62 speed, changing lane) increases the effectiveness of the persuasion (see for example Fogg, 2009a; 2009b, Oinas-
63 Kukkonen, 2013, section 3.2, 4.2). From a safety perspective, it is important to manage the demands placed on
64 the driver. According to the Task-Capability Interface model (TCI) by Fuller (Fuller, 2005), driving demands
65 that exceed driver capability might lead to risky situations such as loss of control or a collision. Managing driver
66 demand is therefore crucial and a key element in ensuring driver safety when applying persuasive approaches, or
67 when communicating information to the driver.

68 In order to keep task demands low, a persuasive system should focus on short term, low effort behavioural
69 responses. These behaviours can be identified through the behaviour taxonomy of Rasmussen (Rasmussen,
70 1983). The taxonomy divides driver behaviour into three levels: skill-based, rule-based, and knowledge-based.
71 Skill-based behaviour is highly automatic and can be performed without much attentional demands. Tasks at the
72 control level fall into this category, and for experienced drivers likely some highly automated behaviours at the
73 tactical level as well in non-complex traffic conditions (e.g. lane changing, overtaking, merging). In rule-based
74 behaviour, a response or a set of responses is selected based on earlier learned rules. Knowledge-based
75 behaviour is applied in mostly unknown situations when novel behavioural responses are needed. Required
76 attentional demands increase from skill-based to rule-based to knowledge-based behaviour. Since behaviour at
77 the tactical level (mostly) consists of skill-based and rule-based behaviours, changing these types of behaviours
78 carries the least risk of imposing high demands on the driver (Birrel, Young, Staton, & Jennings, 2017).
79 However, the context and complexity of the driving environment may influence the difficulty of the tactical level
80 manoeuvres. An example of a low effort behavioural response is requesting a driver to reduce speed in response
81 to downstream traffic disturbance (skill-based, control level). On the other hand, asking a driver to take a
82 different route along a busy unknown road is likely to place higher demands on the driver, since the execution of
83 a task at the strategic level (knowledge-based behaviour) also involves the tactical (rule-based), and operational
84 level (skill-based) (Alexander & Lunenfeld, 1986).

85 We first conduct a critical overview of available behavioural models and select the model most applicable to
86 driver behaviour. We then describe driver behaviour at the tactical level and present the general requirements for
87 an in-vehicle persuasive system. Following this, in section 4, we investigate the different persuasive approaches
88 used in the (traffic) literature and discuss how these approaches fit into the driving environment. Finally, in
89 section 5 we describe the proposed conceptual model and its relation to the current literature. As an example, we
90 apply the conceptual model to the design of a persuasive lane-specific advice system currently in development.

91

92 **2. Describing Behaviour at the Tactical Level**

93 In order to develop our persuasive conceptual model, a behavioural model capable of describing the effects
94 of persuasion on driver behaviour at the tactical level is needed. We have searched the literature for behavioural
95 models that have been used in connection with behavioural change. The search engines used were Google
96 Scholar, Scopus and Web of Science, with the keywords: "behaviour* model AND behaviour* change OR
97 persuasi*". We limited the results to papers of 2005 and newer. Backward snowballing was performed to find
98 the original papers proposing the models. This led to the Social Learning Theory (SLT) (Bandura, 1971), Self-

99 Determination Theory (SDT) (Deci & Ryan, 1985), the Trans-Theoretical Model (Norcross, Krebs, &
100 Prochaska, 2011), and the Theory of Planned Behaviour (TPB) (Ajzen, 1991). For each model, we reviewed
101 their applicability to the driving task, ability to explain the relatively short-term changes in behavioural patterns
102 resulting from persuasion at the tactical level, longer term attitudes towards the use of the system, as well as the
103 ability to accommodate the effects of persuasive efforts.

104 **2.1 Overview of Behavioural Models**

105 The *Social Learning Theory (SLT)*, also known as Social Cognitive Theory, suggests that human behaviour
106 emerges from a constant interaction between environmental, behavioural and cognitive influences (Bandura,
107 1971; Fluegge, 2016). It incorporates elements of operant conditioning to explain how behaviours are learned
108 through social interactions with others (Watkins, 2016). SLT has been applied to a wide range of fields,
109 including how unwanted behaviours may arise (criminal, drug misuse, smoking, traffic violations) and ways to
110 induce a positive change (Hoeben & Weerman, 2016; Lochbuehler, Schuck, Otten, Ringlever, & Hiemstra,
111 2016; Watkins, 2016; Zaso et al., 2016), how public perception is formed and influenced (Fluegge, 2016) and
112 students' tendencies to procrastinate (Gadong & Chavez, 2016). The model is directed at describing how
113 learning experiences are shaped by cognitive and social factors.

114 The *Self-Determination Theory (SDT)* is often cited for its use of intrinsic and extrinsic motivation to explain
115 behaviour (Deci & Ryan, 1985), but actually postulates three basic psychological needs that drive behaviour:
116 autonomy (being in control of one's decisions and behaviour), competence (feeling able to attain behavioural
117 outcomes) and relatedness (feeling understood and respected by others) (Ridgway, Hickson, & Lind, 2016). This
118 model has mostly been applied to behavioural change towards healthier behaviours in the health domain
119 (Friederichs, Bolman, Oenema, Verboon, & Lechner, 2016; Lekes, Houlfort, Milyavskaya, Hope, & Koestner,
120 2016; Niven & Markland, 2015; Sebire et al., 2016; Staunton, Gellert, Knittle, & Sniehotta, 2015), to medical
121 training (Hoffman, 2014), and to volunteering behaviours (Wu, Li, & Khoo, 2015). The SDT describes
122 behavioural motivation at the macro level (Niven & Markland, 2015).

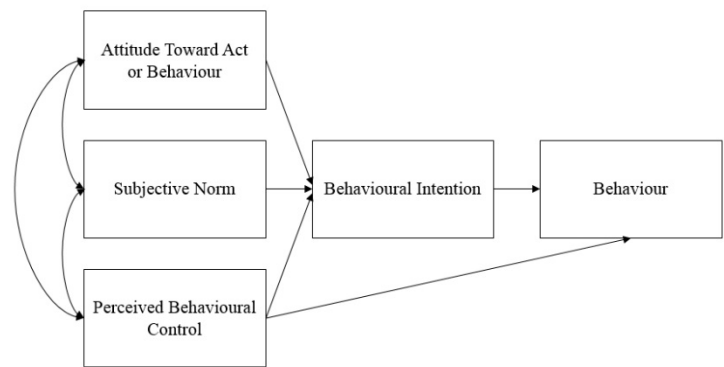
123 The *Trans Theoretical Model (TTM)* describes behaviour as consisting of five stages: pre-contemplation (not
124 thinking about changing behaviour), contemplation (thinking about changing behaviour), preparation (making
125 preparations for changing behaviour), action (changing behaviour) and maintenance (keeping changed
126 behavioural patterns intact) (Norcross et al., 2011). The model originated as a fusion of models from several
127 fields of therapy. Like the SDT, the TTM is a macro model of behaviour, describing high level behavioural
128 processes (see for example Brick, Velicer, Redding, Rossi, & Prochaska, 2016; Kushnir, Godinho, Hodgins,
129 Hendershot, & Cunningham, 2015; Prochaska et al., 1994; Yusufov et al., 2016).

130 The *Theory of Planned Behaviour (TPB)*, based on the Theory of Reasoned Action (Fishbein & Ajzen,
131 1975), posits that behaviour is directly predicted by 'behavioural intention' and 'perceived behavioural control'
132 (the perceived volitional control over the behaviour). 'Behavioural intention' is predicted by 'attitude towards
133 behaviour', 'social norms regarding the behaviour' as well as 'perceived behavioural control'. The model is
134 displayed in Figure 1. In the traffic domain, the TPB has been used to predict traffic violations (Castanier,
135 Deroche, & Woodman, 2013), speeding behaviour (Elliott, Armitage, & Baughan, 2005) and aggressive driving
136 (Efrat & Shoham, 2013). It has also been used successfully in experiments with the goal of behavioural change
137 (Chorlton & Conner, 2012). It describes how situational constraints and long-term attitudes can influence
138 behaviour.

139

140 2.2 Representing Persuasive effects on Tactical Driver Behaviour

141 We have selected the TRB as a behavioural
142 basis for the conceptual model. This is because
143 this theory can explain both short-term behaviour
144 at the tactical level in the driving setting, as well
145 as the long-term social and attitudinal factors
146 acting on behavioural patterns, which might be
147 relevant when explaining variables like
148 continued system usage. The other reviewed
149 models were either geared more towards
150 changing long-term behavioural patterns (SLT,
151 SDT), describing behaviour at a macro level
152 (SDT, TTM), or describing (changing) behaviour
153 in clinical settings (SDT, TTM). The TPB also plays
154 a central role in models of technology acceptance and trust, such as the Technology Acceptance Model (TAM)
155 (F. D. Davis, 1986; F. D. . Davis et al., 1989) and the UTAUT (Venkatesh, Morris, Davis, & Davis, 2003;
156 Vlassenroot, Brookhuis, Marchau, & Witlox, 2010). In this study, we will utilise the TPB (Figure 1) as a
157 behavioural basis for the conceptual model.



158 **Figure 1: The Theory of Planned Behaviour**

159 In more detail, the TPB posits that behaviour is directly predicted by two factors: ‘Behavioural Intention’
160 (BI) and ‘Perceived Behavioural Control’ (PBC). PBC reflects the degree to which the individual perceives to
161 have volitional control over its own behaviour. In other words, whether the individual believes they are able to
162 successfully perform the target behaviour. PBC directly influences behaviour as well as the intention to perform
163 a behaviour. In some studies, PBC has been split into self-efficacy (perceived ability to perform target
164 behaviour) and perceived controllability (perceptions about whether the person has control over the behaviour or
165 outcomes), with only the self-efficacy component being related to changes in BI and behaviour (Elliott,
166 Thomson, Robertson, Stephenson, & Wicks, 2013). This indicates that PBC is more closely related to ‘ability’
167 from the Fogg Behaviour Model (FBM, see 3.1), rather than to a locus-of-control type of evaluation. BI is
168 predicted by ‘Attitude Towards Behaviour’, ‘Subjective Norms’ regarding the behaviour and PBC. The attitude
169 towards the behaviour represents how the behaviour is appraised not only in terms of the act, but also in relation
170 to the possible outcomes of displaying the behaviour, such as potential rewards, or the averting of negative
171 consequences. ‘Subjective norms’ refers to how displaying the behaviour is evaluated by the social network
172 around the individual, and how displaying the behaviour might affect social relationships.

173 3. Influencing Behaviour at the Tactical Level

174 We searched the literature for persuasive methods that were used or have the potential to be used in the
175 traffic domain. The search engines used were Google Scholar, Scopus and Web of Science, with the keywords:
176 “driver persuasion AND system OR ivis OR adas”, “persuasi* AND traffic OR in-car”, “persuasive systems OR
177 persuasive technology”, “persuasive methods”. We limited the results to experimental papers of 2010 and newer.
178 For methodological papers proposing persuasive methods, no time frame was used. Forward and backward
179 snowballing was performed. This resulted in the persuasive categories of Gamification, Behavioural Economics
180 and Captology. These different methods often overlap to some degree in the persuasive elements used. In this
181 section, we discuss these persuasive methods and motivate our choice for the models we adopt for developing
182 the conceptual model.

183 3.1 Persuasive methods

184 The persuasive methods we reviewed can broadly be divided into Gamification, Behavioural Economics and
185 Captology, although these fields show some overlap in the persuasive elements used or approaches taken.

186 Gamification is a term that has emerged relatively recently. Video games create an environment in which the
187 player is highly motivated to perform certain behaviours to achieve game-related goals (finishing a level, getting
188 a high score). Gamification takes the elements that elicit this motivational behaviour and applies them to other

189 situations (Deterding, Dixon, Khaled, & Nacke, 2011). The most often and successfully applied game design
190 elements are leader boards, achievements and challenges (Hamari, Koivisto, & Sarsa, 2014). Gamification may
191 work through raising the driver's implicit motivation, by inducing group-effects such as in-group/out-group bias
192 – simply assigning people to a group, induces positive feelings to other group members (Baron & Dunham,
193 2015) and a motivation to help achieve group goals (Musicant & Lotan, 2015) –, as well as through a 'fear of
194 missing out' effect (Przybylski, Murayama, DeHaan, & Gladwell, 2013). A quite extensive review of previous
195 studies found that generally the effects of gamification are positive, although this is moderated by the context in
196 which gamification is used as well as the users that are targeted (Hamari et al., 2014). Gamification effectiveness
197 might also be reduced over time due to a novelty-like-effect (Farzan et al., 2008). Examples of gamification
198 applied to the transportation domain include EcoChallenge (Ecker et al., 2011): a reward and competition-based
199 system to persuade drivers to engage in a more eco-friendly behaviour, I-GEAR (McCall & Koenig, 2012): a
200 system to change driver behaviour by providing small financial and non-financial rewards, and 'Driving Miss
201 Daisy' (Shi et al., 2012): a gamified solution to help drivers improve their driving skills by providing a virtual
202 passenger that occasionally comments on driving styles.

203 Behavioural economics has been defined as the 'body of work seeking to understand behaviour by
204 incorporating insights from behavioural sciences into economics' (Avineri et al., 2010). Rather than being
205 rational thinkers, people use a range of heuristics and display biases that often work well, but can lead to
206 reasoning errors in certain situations (Kahneman, 2003). An overview can be found for instance in the work of
207 Kahneman (Kahneman, 2013) or Cialdini (Cialdini, 2006). Persuasive elements from Behavioural Economics
208 applied to the transportation domain can be found in for example the design of travel information systems
209 (Avineri, 2011), approaches to promoting safe driving behaviours (Millar & Millar, 2000), and methods
210 analysing travel behaviour (Metcalfe & Dolan, 2012).

211 Captology (acronym: computers as persuasive technology) was introduced by Fogg (1998). It is a field of
212 study which uses computers to influence behaviour in various ways (Fogg, 2010). The Fogg Behavioural Model
213 (FBM) (Fogg, 2009a) is prominent in the field of persuasion. It postulates that in order for a persuasive
214 intervention to be successful, three factors need to converge: the person needs to be able to perform the
215 behaviour ('ability'), be motivated to perform the behaviour ('motivation'), and finally a trigger should be
216 present to elicit the behaviour. Targeting simple behaviours has a higher likelihood of success (Fogg, 2009b). In
217 the context of driver persuasion: making sure 'ability' is high means requesting short, simple to perform
218 behaviours such as a speed change, an overtaking manoeuvre, a lane change, or a merging manoeuvre, as well as
219 timing persuasive attempts to moments when driver workload is not high and when traffic conditions allow for
220 the requested behaviour (e.g. don't request a lane change when the neighbouring lane is crowded). 'Motivation'
221 can be raised by using persuasive techniques (see also 3.2). The FBM has been applied to the traffic setting, for
222 instance it has been applied in a persuasive intervention that successfully reduced texting behaviour while
223 driving (Miranda et al., 2013).

224 3.2 Integrating Persuasive Methods

225 The Persuasive Systems Design model (PSD) (Oinas-Kukkonen & Harjumaa, 2008) presents a systematic
226 framework for designing and evaluating persuasive systems. It brings concepts from Gamification, Behavioural
227 Economics and Captology together. The PSD states that a system can be made persuasive by providing the user
228 with support in distinct categories: primary task support, dialogue support, system credibility support and social
229 support.

230 Primary task support shows many of the principles put forth by the FBM and Behavioural Economics. The
231 focus is on supporting the user by making the behavioural tasks more manageable, personal and transparent.
232 Making the tasks more manageable by reducing complex behaviour to a series of steps and then leading the user
233 through them is especially important when considering in-vehicle systems. Apart from increasing the system's
234 persuasive power, this approach reduces task demands placed on the driver, which in turn increases system
235 safety (Fuller, 2005; Wickens, 2002). An example of primary task support can be a lane change system that
236 guides the driver through the steps of finding a gap, matching speed and merging.

237 Dialogue support is aimed at keeping users moving towards their goals. This support level contains elements
238 from Gamification, Behavioural Economics and the FBM. Offering praise and rewards can increase motivation,

239 which is an important factor for persuasion in the FBM (Fogg, 2009a). If applicable, providing reminders for
240 target behaviour or suggesting certain behavioural responses may be a way to increase behavioural effects by
241 facilitating the creation of habits. Habits are a main factor in making persuasive effects last over time (Lally &
242 Gardner, 2013). Further important factors in dialogue support are similarity and liking (Fogg, 2010), which can
243 increase trust and intentions to comply to system requests.

244 System credibility support is mainly important from the perspective of trust and acceptance. It is about
245 showing the driver that the system makes correct decisions and recommendations. Trust and acceptance are
246 major factors in whether a persuasive system's suggestions or advices will be considered by the driver (Risto &
247 Martens, 2013; Vlassenroot et al., 2010). Factors at this support level relate to the accuracy of the information
248 presented, its transparency, and how users will evaluate it. This in turn is important for forming and maintaining
249 trust in the system (Lee & Moray, 1992; Martens & Jenssen, 2012). The need for trust in a persuasive system is
250 underscored by the work of Risto (Risto & Martens, 2013), who reported that, in their study, drivers constantly
251 tried to verify the accuracy of system requests before following them, and refused to follow messages they
252 interpreted as incorrect.

253 Social support aims at persuading users by increasing motivation using social factors. This level has parallels
254 with Gamification. It includes factors to incentivise behavioural change by allowing performance comparison
255 with other users, facilitating cooperation and/or competition, creating transparency in behaviour-result
256 relationships of other users and even applying forms of normative social pressure (see for example Lütteken,
257 Zimmermann, & Bengler, 2016). Social factors vary in importance and effects on different age groups
258 (McEachan, Conner, Taylor, & Lawton, 2011), which is important for instance when targeting specific
259 demographic groups.

260 To summarise, Gamification has been shown to be effective in motivating people to change their behaviour.
261 However, some studies report that its effectiveness could reduce over time. Behavioural Economics as a field has
262 many applicable concepts that can persuade drivers effectively, and the FBM presents a view of how driver
263 motivation and ability need to converge in the presence of a trigger for persuasive influence to be effective. The
264 PSD model unifies these persuasive methods using the described four support groupings. These provide
265 persuasive elements that can be used depending on the type of system and the context in which it is intended to
266 be applied. For example, in a cooperative system, which is social by nature, the 'social support level' provides
267 ways to add persuasive elements to the social aspects present in the system (see Lütteken et al., 2016). More
268 generally: system credibility can assist persuasion in most systems by increasing trust in the validity of the
269 messages over time, which has been shown to be a large factor in whether a driver responds to the advice or not
270 (Abe & Richardson, 2006; Risto & Martens, 2013), or even a factor in determining system usage over time
271 (Martens & Jenssen, 2012).

272 **4. Considerations for Safe Driver Persuasion**

273 The driving task is complex, requires constant attention from the driver (de Waard, 1996) and presents
274 frequent distractions. Stutts and Gish (2003) report that drivers engaged in distracting activities for 16.10% of
275 the time the car was moving (31.42% if in-car conversations were included). Poorly designed or implemented
276 persuasive in-vehicle systems may increase this percentage by providing more distractions to a driver (Hibberd,
277 Jamson, & Carsten, 2010), potentially increasing driver workload (Horberrry, Anderson, Regan, Triggs, &
278 Brown, 2006), inducing behavioural adaptation (Martens & Jenssen, 2012), or otherwise creating unsafe
279 situations. Safety, therefore, is an important characteristic of a persuasive in-vehicle system. An effective but
280 unsafe system is not likely to be used long term, either through consumer choice or through changing legislation.
281 In this section, we discuss how improving safety can also increase persuasive effectiveness in the short and long
282 run.

283 **4.1 Safety, Driver Demand and Unsafe Situations**

284 A persuasive system needs to communicate with the driver. At the very least this means transmitting
285 information to the driver, and in more complex cases it may require interaction. One way of limiting negative
286 effects of this communication on driving performance, based on the TCI (Fuller, 2005), is by ensuring that the
287 demands placed on the driver do not create dangerous high workload situations. Although this is a broad

288 statement, this requires considering environmental variables that may affect the driver, such as the proximity of
289 other vehicles, traffic conditions and weather conditions, and driver variables such as driving demand and driver
290 workload as well.

291 Driver workload results from the interplay between the demands placed on the driver by the driving task, the
292 complexity of the environment, and the driver's capacity to meet those demands (de Waard, 1996). It is an
293 important factor in terms of safety, since under- or overload can influence a driver's workload and create
294 hazardous situations (M. S. Young et al., 2015). In section 1.2 we have discussed how targeting the tactical level
295 for persuasive attempts will likely limit the impact on driver demand (compared to targeting the strategic level),
296 and by extension, on driver workload. Despite this, a poorly designed persuasive system targeting tactical-level
297 behaviours may still result in high driving demand and/or workload. The Multiple Resource Theory (MRT) by
298 Wickens (Wickens, 2002) can help understand why, even when a persuasive in-vehicle system targets simple-to-
299 change behavioural tasks, high driver demand or workload may still result.

300 In the MRT, interference from a secondary task is most likely when it accesses the same resources as the
301 primary task. Since driving is mainly a visual task, transmitting information to the driver through a visual
302 channel may cause interference. For instance, diverting the eyes from the road for extended time has serious
303 consequences for driving performance and lane-keeping ability (Peng, Boyle, & Hallmark, 2013). Heads-Up-
304 Displays do not require the driver to take his eyes off the road and can be a better alternative (Liu & Wen, 2004),
305 but do not mitigate all negative effects, and can introduce some new potential problems related to sharing visual
306 resources and to characteristics of the human visual system (Edgar, 2007). Competing resource types are not the
307 only factor in the MRT that can lead to reduced task performance, however: if the demands of one or both tasks
308 are higher than what the driver can handle, two tasks that use very different resources are still likely to cause
309 dual-task interference and degrade driving performance. In terms of a persuasive in-car system, minimising the
310 effect on workload therefore means choosing the correct modality to transmit information to the driver, keeping
311 the cognitive demands of the interaction low to prevent interference with the main driving task, and timing the
312 messages to periods when the driver can accommodate them. If the cognitive demands of the main task (driving)
313 are already high, per the MRT a simple secondary task may create dual-task interference even when using a
314 different modality from the main task, degrading the performance on the main task and thereby potentially
315 compromising driver safety. This is the rationale behind adaptive interfaces (Birrel et al., 2017; Park & Kim,
316 2015): by changing either the complexity of messages presented, the modality used to convey the message to the
317 driver, or by suppressing messages in conditions where safety or workload may be dangerously affected, safety
318 can be improved.

319 Unsafe situations can still arise from persuasive in-vehicle systems even when changes induced in driver
320 demand and workload are minimal. A system that distracts the driver at the wrong moment may create a
321 potentially dangerous situation (K. Young & Regan, 2007), highlighting the importance of timing the
322 communication with the driver. Unsafe situations may also arise from the way drivers accommodate the
323 functions of in-vehicle devices into their driving habits, giving rise to behavioural adaptation effects (Martens &
324 Jenssen, 2012; Smiley, 2000). For example, in response to having Anti-Lock Braking (ABS) and Airbag systems
325 installed, headways decreased and seatbelt usage reduced (Sagberg et al., 1997). Overreliance on a system is
326 another potential problem. For example, with a lane-change advice system: if a driver places too much trust in
327 the lane change advice system, a lane change may be initiated when the system gives an advice, without the
328 driver checking whether it is actually safe to change lane.

329 **4.2 Persuasive Attempts and Acceptance**

330 A persuasive in-vehicle system needs to be able to consistently persuade the driver. According to the Fogg
331 Behaviour Model (FBM) (Fogg, 2009a), persuasive interventions timed to periods when both motivation and
332 ability are high, have a higher chance of resulting in changed behavioural outcomes. In terms of an in-vehicle
333 system, an advice that is given to a driver when there is a high motivation to follow it, will have a higher
334 probability to be complied to. Similarly, an advice given at a time when the driver ability is high, i.e. when the
335 driver perceives they can follow the advice, will be more likely to result in the target behaviour. This again
336 underscores the importance of targeting behaviours that require less effort to change, such as tactical level driver
337 behaviour: not only it is safer, persuasive effectiveness is also likely to increase when doing so (Fogg, 2009a). In

338 the traffic context, the FBM's 'ability' to follow a persuasive advice can be impacted by multiple factors and
339 conditions, such as weather conditions, traffic conditions, secondary tasks or driver states (de Waard, Kruizinga,
340 & Brookhuis, 2008). One such driver state is driver workload, which needs to be considered for the effectiveness
341 of persuasion as well as for safety. When driver workload is high, presenting an advice and/or requesting an
342 action from the driver may increase the difficulty of the driving task further, in turn reducing the likelihood that
343 the driver complies to the persuasive request because the requested behaviour is seen as difficult or impossible
344 given the circumstances. In other words, high workload is likely counterproductive when trying to persuade the
345 driver.

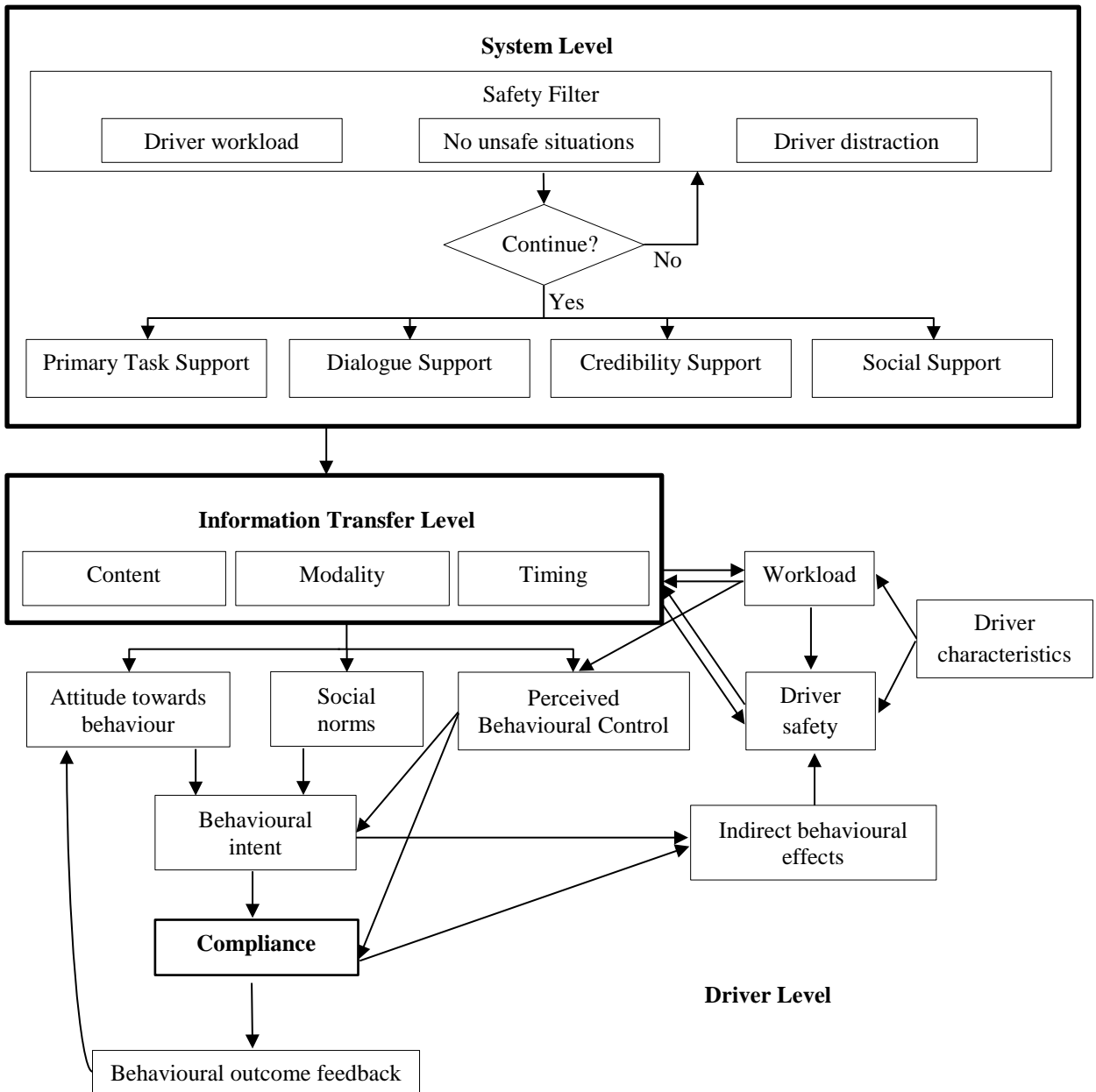
346 In addition to persuading a driver effectively, a persuasive system needs to be, and keep on being, used. To a
347 large degree, this usage will depend on the acceptance of a system (Vlassenroot et al., 2010). Without taking
348 steps to ensure acceptance, there is the risk that a persuasive in-vehicle system falls into disuse or works
349 counterproductively (Martens & Jenssen, 2012). This is especially damaging if the system relies on a user base
350 to function, as for example with cooperative (lane change) systems (Lütteken et al., 2016). To describe the
351 acceptance of new technology several models have been developed, such as the Unified Theory of Acceptance
352 and Use of Technology (UTAUT) (Venkatesh et al., 2003) and the Technology Acceptance Model (TAM) (F. D.
353 Davis, 1986).

354 **5. The Conceptual Model for Driver Persuasion at the Tactical Level**

355 In this section, we present the proposed conceptual model for driver persuasion at the tactical level using in-
356 vehicle systems. The model has three levels: The System Level, the Information Transfer Level and the Driver
357 Level. The System Level is where the persuasive strategy is formed and safety checks are performed. It
358 incorporates the defined safety criteria (4.1, 4.2) and the four support levels from the Persuasive Systems Design
359 model discussed earlier (3.2). The Information Transfer Level is where communication with the driver takes
360 place, and incorporates elements from Wickens' MRT and Fuller's TCI Model. The Driver Level describes the
361 behavioural effects of the persuasive attempt. It incorporates the TPB (2.2), along with considerations regarding
362 effects on driver workload, indirect behavioural effects and driver safety (4.1, 4.2). The following sub-sections
363 detail these levels and how they are built up from the existing models and theories in the literature.

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387 **Figure 2: Proposed model for influencing tactical driver behaviour**

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389 **5.1 Planning Driver Persuasion: The System Level**

390 The System Level represents the back-end of the persuasive in-vehicle system. It is built up from the PSD
 391 model (3.2) and the considerations of driver safety and the persuasiveness (4).

392 Safety is central to the persuasive system design and operation. This is explicitly reflected in the model,
 393 where the first evaluation made is whether it is safe to initiate an information transfer to the driver. Ideally, the
 394 persuasive system should (either directly or indirectly) take driver workload into account, should not create
 395 unsafe traffic situations, and should aim not to distract the driver at the wrong time. Only then can an attempt to
 396 persuade the driver be made with a high likelihood of being safe (see 4.1). In situations where safety criteria are
 397 not met, they can be re-evaluated until they are met, represented in the model by the conditional loop. These

398 safety criteria can be evaluated from the environment, such as in systems that monitor on-coming traffic (Curry
399 et al., 2010) or weather conditions (Green, 2004), or from the driver, such as in systems that try to estimate
400 driver state (Ferreira et al., 2014; Liang, Reyes, & Lee, 2007).

401 Once it is determined that interacting with the driver does not pose a safety risk, tactical driver advice may be
402 given to persuade the driver. The PSD described earlier combines persuasive techniques into four support levels.
403 These four levels of support are included as possible routes to persuasion (see also (Oinas-kukkonen &
404 Harjumaa, 2009; Oinas-Kukkonen & Harjumaa, 2008), 3.2).

405 **5.2 Interacting with the Driver: The Information Transfer Level**

406 The information transfer level is where the communication between the persuasive system and the driver
407 takes place. Usually this is through a type of interface (visual, auditory, tactile or multimodal). The information
408 transfer level and its effects on behaviour (driver level, 5.3) are built up from the TPB, MRT, TCI and FBM
409 discussed in the previous sections. The information transfer itself is operationalised as having ‘content’ (what’s
410 in the message?), ‘modality’ (how is the message transmitted to the driver?) and ‘timing’ (when is the message
411 transmitted?) as factors. In the model, the information transfer influences driver workload, driver safety and the
412 behavioural determinants of the TPB (attitude, social norms and perceived behavioural control). Here we discuss
413 these effects in terms of the impact on safety and the impact on persuasive potential.

414 From a safety perspective, the model shows an effect of the information transfer on ‘workload’ and
415 ‘perceived behavioural control’ based on the TPB, MRT and TCI. According to the MRT, dual-task interference
416 is likely when two concurrent tasks use the same modality, or when the cognitive load from one or both tasks is
417 high. Dual-task interference reduces performance on the main (driving) task and increases demands placed on
418 the driver, which in turn can raise workload. As demands and workload rise, the perceived behavioural control of
419 the driver reduces: the higher the driver workload, the more difficult it will be to pay attention to or comply with
420 persuasive messages. These effects are crucial, since they can lead to persuasion being ineffective, a degradation
421 of driver performance, or even undesirable situations such as a loss of control or a collision (TCI, 4.1, 5.3). A
422 direct link to driver safety is also included, which includes for example situations where the information transfer
423 leads to eyes-off-road situations (Dozza, 2013; Peng et al., 2013) or to distraction at a critical moment.

424 From the persuasion perspective, the FBM (Fogg, 2009a) specifies that motivation and ability need to be high
425 at the moment of a behavioural trigger, in order for persuasion to have a high chance of being successful. The
426 goal of the persuasive techniques used (‘content’) is to raise motivation to perform a behaviour, for instance by
427 using social support to increase motivation to comply to a message. Making sure ‘ability’ is high, essentially
428 means timing the information transfer to situations where the driver’s PBC is high (Elliott et al., 2013, see also
429 section 5.3). In a driving setting, the PBC term implicitly includes an environmental component (e.g. give a lane
430 change request only when there is sufficient room on the adjacent lane), and a driver component (a high
431 workload will result in lower PBC). Both components are important for persuasion and safety: when a driver
432 does not feel capable of performing the requested behaviour, it is unlikely the persuasion will have an effect.
433 Alternatively, if an already overloaded driver complies with a requested behaviour, this may have a detrimental
434 effect on safety.

435 **5.3 Human Factors: The Driver Level**

436 The driver level provides a basis to describe expected behavioural effects of the persuasion. In this section,
437 we describe how the TPB fits into the model, how workload relates to both safety and persuasion, its dependence
438 on driver characteristics and factors on the information transfer level, possible behavioural effects and the
439 importance of outcome feedback.

440 As argued in the previous section, both motivation and ability need to be high in order for persuasive systems
441 to actually persuade (Fogg, 2009a). In the conceptual model, motivation is captured by the TPB terms ‘attitude
442 towards behaviour’ and ‘social norms’, and ability is captured by ‘perceived behavioural control’ (PBC). The
443 attitude and social norms influence driver behaviour through the ‘behavioural intent’ (BI) (Ajzen, 1991;
444 Armitage & Conner, 2001; McEachan et al., 2011). PBC affects both BI and the actual behaviour (Armitage &
445 Conner, 2001; McEachan et al., 2011). Additionally, PBC acts as a modulator of workload on behaviour. As
446 discussed in 2.2, PBC relates to the perceived ability a person has to perform a given behaviour, rather than a

447 locus of control-like evaluation of whether the behaviour lies within the control of the individual (see also Elliott
448 et al., 2013). The relationship between workload and behavioural outcomes is inverse: a high PBC means a
449 driver feels competent and able to perform a requested behaviour, whereas a low PBC will negatively influence
450 the likelihood of a behavioural result occurring.

451 Apart from the information transfer (5.2), driver workload is also affected by ‘driver characteristics’. Driver
452 ability is not static and varies between and within individuals over time (M. S. Young et al., 2015), which may
453 cause workload experienced by two different drivers or a single driver in two comparable situations to be very
454 different. ‘Driver characteristics’ also includes differences in inherent driver safety. For example, some age
455 groups display more risky behaviour (Carter, Bingham, Zakrajsek, Shope, & Sayer, 2014), there may be sex
456 differences or geographical differences in driver behaviour and capability (Twisk & Stacey, 2007; Vlakveld,
457 2011), or individual differences in driver aggression (Hennessy & Wiesenthal, 2001). These characteristics may
458 result in some classes of drivers being exposed to higher risk while driving.

459 ‘Indirect behavioural effects’ (Martens & Jenssen, 2012) were discussed in 4.2, meaning changes in driver
460 behaviour or intentions to perform behaviours that are not intended by the designers of the (persuasive) system.
461 An often-cited example of indirect behavioural effects is that of the anti-lock braking system (ABS), which helps
462 reduce stopping distances of the cars in which it is installed. Positive effects were offset by behavioural effects:
463 adaptation was reported from drivers choosing to driver faster on wet surfaces (Smiley, 2000) or with shorter
464 headway and varying seatbelt usage (Sagberg et al., 1997).

465 The last undiscussed term in the model is feedback about behavioural outcomes. This feedback, including
466 information on the behaviour-result relationships in other drivers, is expected to influence the driver’s attitude
467 towards future behaviours in a feedback loop (see also Lütteken et al., 2016). For instance, if a driver observes
468 that complying to an in-vehicle system has resulted in shorter travel times on previous occasions or with other
469 drivers, this might bias the driver to comply more with the system’s advices in the future. This ties into the
470 “system credibility support” level of the PSD (Oinas-Kukkonen & Harjumaa, 2008). It is also in line with an
471 earlier study into compliance to tactical driving advice (Risto & Martens, 2013), where drivers were observed
472 attempting to evaluate the validity of tactical advice in the context of what they observed on the road and the
473 history of the system’s accuracy.

474 **6. Application to a Lane-Specific Advice System**

475 In this last section, we present a case study based on a lane-specific advice system, in which we apply the
476 developed model to the system and discuss how this helps structure system design for safety and persuasion.

477 The goal of this system is to reduce travel time delay and congestion by encouraging a better distribution of
478 the vehicles over the available motorway lanes. This means advising drivers on what lane to take, depending on
479 external factors. For instance, an unbalanced distribution, an upcoming on-ramp or lane drop, or an incident
480 upstream may require a redistribution of traffic to ensure continued flow and avoid congestion. The system’s
481 advices will be in the collective benefit of drivers on a specific stretch of road (minimised total travel time), but
482 will sometimes not be in the benefit of individual drivers receiving the advice (e.g. stay behind this slow truck
483 for now), creating a potential problem (Risto & Martens, 2012). Persuasive techniques will be used to engage
484 drivers with the system and to also stimulate adherence to lane-specific advices, especially when they are not in
485 the individual’s benefit. The system will consist of an in-vehicle part and a back-end that predicts traffic states
486 and approximates the optimal lane use situation.

487 The developed conceptual model described in this paper helped to direct our research in several ways. At the
488 ‘System Level’, a safety filter is required. Early in the design phase, this redirected the process from focusing
489 mostly on the effectiveness of the persuasive design, to an approach that considered potential effects on safety
490 and on the driver as well. As a result, we are developing an affordable driver monitoring system to estimate
491 driver state (Gent, Farah, Nes, & Arem, 2017; van Gent, Melman, Farah, van Nes, & van Arem, 2018). In
492 combination with environmental sensing systems built into the vehicle, this provides a safety filter that will
493 suppress messages to drivers that are estimated not to respond (safely) to the persuasion. The result of this

494 message filtering, we argue, is two-fold (see 4.2, 5.2, 5.3): apart from increasing the safety of the system, it
495 works to increase persuasive effectiveness and facilitate long-term usage of the system as well.

496 Persuasive strategies are outlined in the four support levels from the PSD model (Oinas-Kukkonen &
497 Harjumaa, 2008, see also 4.2). These support levels offer persuasive strategy elements from which a selection
498 can be made. We selected strategies mainly from primary task support and dialogue support, with some elements
499 from the other two support levels. The system will support the driver by breaking down a requested lane-change
500 into smaller steps, and guiding the driver through them (primary task support: 'reduction' and 'tunnelling'). This
501 will increase persuasive power and make the task less demanding, benefitting both safety and persuasion (Fuller,
502 2005; Wickens, 2002, see also 4.1). Second, the system will provide the user with transparent information
503 regarding obtained benefits in terms of travel time saved in relation to the performed behaviour through either an
504 app or a web-portal (primary task support: 'self-monitoring'). Providing a means of 'self-monitoring' of on-
505 going benefits increases immediate persuasive effects, but also works to increase 'trustworthiness' and
506 'verifiability' of the system (credibility support).

507 At the information transfer level, an advice is communicated to the driver, the effects of which are described
508 at the driver level (5.3). As described in 5.2, in the model the information transfer between system and driver is
509 operationalised as having content, modality and timing. The model shows how these factors mediate safety and
510 persuasive effectiveness through workload and perceived behavioural control (see also 4.2, 5.2). This means that,
511 in further development of our lane-specific advice system, our research will focus on how driver workload and
512 perceived behavioural control are influenced by content, modality and timing decisions with our lane-specific
513 advice system. Additionally, it simplifies the scope of our research: in order to estimate the effects on the
514 behavioural outcome, we only need to investigate how the three information transfer factors influence the
515 'attitude towards behaviour', the perceived 'social norms' and the PBC. How these three factors in turn influence
516 BI and Behaviour is known from several exhaustive meta analyses (Armitage & Conner, 2001; McEachan et al.,
517 2011; Notani, 1998). To assist in estimating how our persuasive system influences these factors, it is useful to
518 point out that guidelines have been formulated on how to operationalise these constructs (Ajzen, 2010; French &
519 Hankins, 2003).

520 In this section, we have applied the model to the design of our persuasive lane-specific advice system, and
521 have discussed how this helped shift the focus of our research away from one emphasizing persuasion, to one
522 that includes the driver's behaviour and traffic safety as well. We have shown how this shift will benefit not just
523 traffic safety but the persuasive effectiveness of the system as well.

524

525 **7. Conclusion**

526 In this paper, we have proposed a conceptual model to help guide the design of persuasive in-vehicle systems
527 with the aim of influencing driver behaviour at the tactical level. The model was designed with safety and
528 persuasion as core elements, and explains how a persuasive in-vehicle system is expected to affect driver
529 behaviour, workload, and safety. The model contains four 'support levels' from the PSD from Oinas-Kukkonen
530 (2009), that can be used as guidelines for implementing specific persuasive elements in persuasive in-vehicle
531 systems.

532 The proposed model is split into three levels explaining the different elements of the information chain: the
533 system level where the persuasive strategy is formed after a safety check, the information transfer level where
534 communication with the driver takes place, and the driver level where the act of presenting advice impacts driver
535 behaviour, workload and safety in several ways. The focus while designing the model was on safely attaining
536 effective driver persuasion. As a behavioural basis, the Theory of Planned Behaviour was selected. The
537 persuasive elements come from the PSD model. We have discussed how the PSD is built from elements in
538 Gamification, Behavioural Economics and Captology. We have also included elements from Wickens' MRT
539 Model and Fuller's TCI that help explain why the timing and modality of the information transfer are key factors
540 in both safety and persuasive effectiveness. Finally, we have applied the model to a persuasive system which
541 aims to reduce travel time delay and congestion by encouraging a better distribution of the cars over the

542 available motorway lanes, to illustrate how the application of the model guided our research efforts and helped
543 shape a safe and effective design.

544 Persuading drivers is a complex task, especially since the driving environment requires extra considerations
545 in terms of safety, and because the demands the environment places on drivers are highly dynamic. The
546 presented work and model in this paper, assist those working on driver persuasion by showing how to safely
547 achieve persuasive effectiveness.

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553

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