Computational simulation of visual distraction effects on car drivers’ situation awareness

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Abstract

This paper presents a computational modeling approach for negative effects simulation of visual distraction while driving a car. In order to investigate these effects, an experiment was firstly implemented on a driving simulator. Twenty participants were invited to perform a car following task in different driving conditions (12 driving scenarios), with or without a secondary task of visual distraction. Empirical data collected through this experiment show that visual distraction negatively impacts the driving performance at both perceptual and behavioral levels, and then increase the risk of having a crash. Beyond these effects on the observable performance, the aim of this study is also to investigate and simulate these distinctive effects on mental models of the road environment. Indeed, driver’s decisions and behaviors are based on a temporal-spatial mental model, corresponding to the driver’s situation awareness (SA). This mental representation must be permanently updated by perceptive information extracted from the road scene to be efficient. In case of visual distraction requiring off-road scanning, mental model updating is imperfectly done and driver’s actions are thus based on a mental representation that can dramatically differ from the situational reality, in case of a critical change in the traffic conditions (e.g. sudden braking of the lead car). From these empirical results, a computational model (named COSMODRIVE for COgnitive Simulation MOdel of the DRIVER) was implemented for simulating visual distraction effects and human errors risks at perceptive (visual scanning changes) cognitive (erroneous Situation Awareness) and behavioral levels (late reaction time and crash risk increasing).

Keywords: Computational Model, car Driver, Visual Distraction, Situation Awareness, Temporal-Spatial Mental Representation.

1. Introduction: Visual distraction and research objective in terms of computational simulation

Driving requires visual attention in order to safely control the car and to respond to events happening on the road. Driver distraction occurs “when a driver is delayed in the recognition of information needed to safely accomplish the driving task because some event, activity, object, or person within or outside the vehicle compelled or tended to induce the driver’s shifting attention away from the driving task” (Treat, 1980). In the same way, distraction has been more recently defined by Lee, Regan and Young (2009) “as a diversion of attention away from activities critical for safe driving toward a competing activity”.

Two main forms of distraction are commonly described in the literature, namely visual and cognitive distraction. The former takes a driver’s “eyes-off-road”, while the latter takes their “minds-off-road” (Victor et al., 2005). The present study is focusing only on visual distraction due to a secondary visual task taking the driver’s eyes off the road.

This type of distraction can occurs when drivers look at in-vehicle displays. For example, in research conducted by Wierwille et al. (1988) under real traffic conditions where text was displayed on an on-board screen, the average length of a glance at the outside environment was 1.5 to 1.7 seconds, while the amount of time spent watching the road decreased to about 50 to 65% of total eye movement. Visual scanning toward on-board-devices varies with the nature of displayed information and the type of additional task to be performed while driving, but also according to the situational demand and driver’s adaptation strategies regarding both the driving situation and the demand and the reading task demands. However, focusing visual attention for some period of time on in-vehicle visual target creates an unsafe driving issue. Senders et al. (1967) argued that when drivers look away from the road, uncertainty about the roadway situation increases. When uncertainty reaches a certain threshold, drivers look back to the road. More recently, Wierwille (1993) quantified this threshold of off-road glance duration at 1.8 seconds on a straight road and 1.2 seconds on a curve on average for a normal driver. Such thresholds may also vary according to driver speed or traffic condition and may be also subject to individual differences.

Changes in driver behavior due to visual distraction have been identified in simulator or in vivo studies. Several studies have shown that visual distraction increases the dispersion of eye gaze pattern from the roadway (e.g. Donmez et al., 2007). In terms of driving performance, visual distraction has been also associated in the literature with large, discrete steering adjustments and increased lane deviations (e.g. Engström, Johansson and Östlund, 2005; Salvucci, 2001). However, as discussed by Zhang (2011), such inference has been mainly assessed for lower-level of driving control and less is known concerning the internal cognitive effect of visual distraction. Moreover, as explained this author, “a data-driven approach to identifying detrimental effects of distraction may not be sufficient to establish a causal link between driver performance and visual process interference. Modeling internal changes in driver Situation Awareness of the driving environment due to visual distraction are required to conclusively identify precise relationships between drivers’ performances and distraction and support effective mitigation strategies for distractions”. This is typically what this research would like to explore. Beyond the well-known impact of visual distractions on drivers’ visual strategies and driving performance at the operational level (e.g. Salvucci, 2001), the aim of this research is above all to
develop a computational model able to simulate these destructive effects on drivers’ Situation Awareness modeling as a dynamic visual-spatial model of the surrounding.

2. Empirical data collection among human drivers to study visual distraction effects

The methodological specificity of the driver modelling approach implemented in this research was to use the same virtual Platform (named SIVIC; Gruyer et al., 2006) as (i) a driving simulator for empirical data collection among human drivers, and then, as (ii) a virtual road environment to be interfaced with the driver model for virtual simulations (in charge to reproduce humans’ performances). According to this approach, human drivers’ behaviour and driver model performances were observed and simulated for the same driving scenarios, in the same virtual road environment.

2.1 Apparatus

The experiment used a fixed-base simulator integrating a real car seat, three PC monitors for presenting the driving scene (the back mirror view is computationally integrated in the central image), and a Logitec G 25 kit including the steering wheels, 3 pedals, a gear box, and indicators. Two web-cameras were used for recording drivers’ face and feet movement on the pedals. A third video camera was also added behind the car seat, in order to film the driving environment and the driver’s activity. A 12-inch tablet computer was placed in front of the main simulator screens. This display was used to present the visual distraction tasks to the drivers. This screen was positioned approximately 15 degrees down and 30 degree right of the natural line of sight of participants in viewing the driving scene.

2.2 Participants

Twenty experienced drivers of middle-age (from 23 to 56 years old) participated to this experiment. All the drivers have a minimum of 5 years of driving experience and they drive a minimum of 5,000 km per year. The recruitment of subjects was balanced for gender. Participants were instructed to perform the secondary task in accordance with the demands of the driving situation. The instruction emphasized that safe driving was of the highest priority.

2.3 Driving task

The full experiment followed a 3×2×2×2 factorial design with one primary driving task of car following to be performed in three different driving contexts (requiring different driving speeds: 130 km/h for Highway, 90 km/h for rural roads and 50 km/h for urban areas), from two required following distances (free versus imposed at a value of 0.6 second of Inter-Vehicular Time [IVT]), and two types of lead car behavior (having a steady versus irregular velocity) and then, two levels of visual distraction (with and without). In total, there were 12 driving scenarios to which each participant was exposed, once time without any secondary task, and then, on time with a secondary task. Each scenario was around 1 minute in duration and presented one experimental condition.

2.3 Visual secondary task

The Secondary Task of visual distraction to be performed by the participants was the following: a set of 3 visual pictograms, associated with an auditory beep, were displayed on an additional screen situated on the right side (near the usual position of the radio). Some seconds later (from 3 to 4 sec.), 1 of this 3 pictograms appeared under the first set, and the driver had to use a 3-buttons command for indicating which pictogram is replicated (Fig. 1).

![Figure 1: The visual Secondary Task to be performed](image)

2.4 Main results

2.4.1 Visual strategies for additional screen scanning

Visual strategies during secondary task have been extracted from the analysis of video film of participants’ faces collected during the experiment. The two main different visual scanning patterns of the additional-screen observed among human drivers are presented in figure 2 (others strategies are adaptations of one of the two main patterns).

![Figure 2: Visual scanning patterns observed among human drivers](image)
answer. By contrast with the preceding one, this second strategy requires a several glances, but the advantage is to process visual information in two times (corresponding to question, and then answer), requiring a shortest last glance.

2.4.2 Visual distraction effect on driving performance

Two main negative impacts of a visual distraction on the drivers’ performances were observed during this experiment. The first one occurs in normal conditions, and the second one occurs for critical scenarios (i.e. when the lead car brakes), increasing the accident risk.

In normal driving conditions, two main differences due to visual distraction were observed among the participants: (i) a significant reduction (T-test, p<0.001) of the safety margins in free following conditions (without ST, mean value of IVT is of 3 s. without ST, vs 2.65 s. with ST) and (ii) a significant degradation (p< 0.05) of the following performance in constrained following conditions (in these scenarios, drivers have to follow the lead car at an imposed IVT of 0.6 s., and the percentage of time when this value is performed is of 57% without ST, vs 44 % with ST). These results show a negative effect of visual distraction for short following distance keeping.

In critical driving conditions, the two main negative impacts of the visual ST on drivers’ performances are (i) an increasing of reaction time for braking (the differences are only significant for the constrained following task : 0.89 s. vs 1.1 s.; p<0.05), and (ii) a risk of crash increasing. The Table 1 presents the percentages of collision occurring with the lead car for the total number of required emergency braking, by respectively considering the different driving scenarios investigated. It appears that the risk of collision due to a visual distraction is here significantly increased for 4 of the 10 driving scenarios requiring an emergency braking (i.e. bold values). The highest negative impacts of visual ST were observed for the constrained unsteady car following scenarios, in both urban and rural environments.

<p>| Table 1: Percentages of collision with the lead car |
|---------------------------------|-----------------|-----------------|-----------------|</p>
<table>
<thead>
<tr>
<th>Context</th>
<th>Driving scenario</th>
<th>No ST</th>
<th>With ST</th>
</tr>
</thead>
<tbody>
<tr>
<td>Highway</td>
<td>Free steady lead car following</td>
<td>55%</td>
<td>50%</td>
</tr>
<tr>
<td></td>
<td>Free unsteady lead car following*</td>
<td>35%</td>
<td>50%</td>
</tr>
<tr>
<td></td>
<td>Constrained steady lead car following</td>
<td>65%</td>
<td>70%</td>
</tr>
<tr>
<td></td>
<td>Constrained unsteady lead car following</td>
<td>70%</td>
<td>70%</td>
</tr>
<tr>
<td>Rural</td>
<td>Free unsteady lead car following</td>
<td>60%</td>
<td>60%</td>
</tr>
<tr>
<td></td>
<td>Constrained unsteady lead car following*</td>
<td>55%</td>
<td>80%</td>
</tr>
<tr>
<td>Urban</td>
<td>Free steady lead car following*</td>
<td>20%</td>
<td>30%</td>
</tr>
<tr>
<td></td>
<td>Free unsteady lead car following</td>
<td>30%</td>
<td>30%</td>
</tr>
<tr>
<td></td>
<td>Constrained steady lead car following</td>
<td>30%</td>
<td>30%</td>
</tr>
<tr>
<td></td>
<td>Constrained unsteady lead car following*</td>
<td>25%</td>
<td>90%</td>
</tr>
</tbody>
</table>

(*Bold Values indicate the main observed differences in driving performance due to visual distraction)

2.4.3 Example of crash due to visual distraction

The following figure presents a typical case of driving accident due to visual distraction, as observed during this experiment (in free following conditions, view a). In this example, the lead car brakes when the driver is looking for the additional screen (view b), via a long glance of 2 seconds. When she repays attention to the road (view c), she however discovers a critical gap between the expected position of the lead car as mentally assessed during the off-road glance (by assuming a steady speed of the lead car during this period) and the objective reality where the lead car is actually very close. Therefore, she immediately carried out an emergency braking (0.78 second of reaction time). Unfortunately, the collision cannot be avoided, and the crash with the lead car occurs in view d.

3. Computational modeling and simulation of visual distraction effects on drivers’ SA

By using the empirical data collected in this experiment, a computational model, based on the COSMODRIVE (COgnitive Simulation MÖdel of the DRIVEr) theoretical approach (Bellet et al., 2007), has been implemented into the SIVIC virtual plate-form (Gruyer et al., 2006). By contrast with other driver models available in the literature, the core specificity of COSMODRIVE is to simulate drivers’ mental representation as a visual-spatial (i.e. 3 Dimensional) and dynamic model of the road environment. Indeed, from their interaction with the road environment, drivers build mental model of events and objects surrounding them. This mental model corresponds to the driver’s Situation Awareness (Endsley, 1995). They are dynamically formulated in working memory through a matching process between perceived information and pre-existing operative knowledge (Ochanine, 1977). At the tactical level (Michon, 1985), such a mental representations provides an ego-centred and a goal-oriented understanding of the traffic situation. They take the form of a dynamic 3D model of the road environment, liable to be mentally explored by the driver in order to anticipate events or action effects through cognitive simulations of mental deployment (Bellet et al, 2010; 2011), and thus providing expectations on future situational states. This cognitive process of anticipation, based on both implicit and explicit mental
simulations (Bellet et al., 2009), is a core function of the human cognitive system in dynamic contexts. The central structure supporting to the driver’s SA in COSMODRIVE cognitive architecture is *working memory*. From this point of view, this architecture is inspired by the ACT-R theory (Anderson et al., 2006). However, the working memory of COSMODRIVE merges both *procedural* and *declarative* memories, and comes more from the *operational memory* concept of Zinchenko (1966) than from the Baddeley’s *working memory* model (1986). With COSMODRIVE, car driving is modeling as a dynamic regulation loop of interaction between drivers and the road environment.

Figure 4 provides a synthetic overview of this model as implemented on the SIVIC virtual platform. synthetically, the functional architecture of the model is based on 3 main modules (i.e. Perception, Cognition, and Action modules), in order to drive a virtual car into a virtual environment, through two synchronized “Perception-Cognition-Action” regulation loops: an automatic and implicit mode versus an attentional and explicit mode (Bellet et al, 2009). This dichotomy is well established in scientific literature, for example, with the distinction put forward by Schneider and Schiffrin (1977) between controlled processes, which require cognitive resources and which can only be performed sequentially, and automatic processes, which can be performed in parallel without any attentional effort. In the same way, Rasmussen (1986) distinguishes different levels of activity control according to whether the behaviours implemented rely on (i) highly integrated sensorial-motor reflexes (Skill-based behaviors), (ii) well mastered decision rules for managing familiar situations (Rule-based behaviors), or (iii) more generic knowledge that is activated in new situations, for which the driver have not any prior experience (Knowledge-based behaviors).

### 3.1 The Perception module

The Perception Module acts as an “interface” between the external road environment (as simulated with SiVIC) and the driver model. It simulates human information processing of sensorial data before their integration in the Cognition module for traffic conditions analysis, situational change anticipation, decision making, and then action planning and implementation through the Action Module. The *Perception module* is based on a *virtual eye* (Figure 5). This virtual eye includes three visual field zones: the central zone corresponding to *foveal vision* (solid angle of 2.5° centred on the fixation point) with a high visual acuity, *para-foveal vision* (from 2.5° to 9°), and *peripheral vision* (from 9° to 150°), allowing only the perception of dynamic events.

From this virtual eye, COSMODRIVE is able to integrate information perceived in the road environment through two main processes (Bornard et al, 2011). The first one, named *perceptive integration*, is a data-driven process (i.e. bottom-up) and allows the cognitive integration of environmental information in the driver’s mental representations. The second one, named *perceptive exploration* and based on Neisser’s *perceptual cycle theory* (1976), is a “knowledge-driven” process (i.e. top-down) in charge to actively explore the road scene, according to the tactical goal to be reached and to the event expectations included in the driving schemas. In the frame of a car following task, the main point of interest of the driver’s visual attention is the lead car. However, in case of a visual secondary task to be performed while driving, the virtual eye must sometimes leave the road in order to observe the additional screen, according to the 2 different visual scanning patterns observed among human drivers during our experiment (presented in fig. 2).

### 3.2 The Cognition module

The Cognition Module is in charge to support drivers’ Situation Awareness, Decision-Making and Action Planning. The specificity of COSMODRIVE architecture, by contrast with other driver models developed with ACT-R (e.g. Salvucci, 2006), is to be able to support dynamic reasoning based on visual-spatial mental models. This type of 3-Dimensional (3D) mental representation is a key component of real drivers’ cognition (Bellet et al, 2009), but they are not easy to implement, and then to process, with the ACT-R architecture (except as a *chunk*, i.e. a set of facts or logical units, stored in *declarative memory*). In order to simulate human drivers’ visual-spatial knowledge and dynamic reasoning in a more realistic way, we therefore defined a specific computational formalism named *driving schemas* (Bellet et al, 2007). Coming from both Piaget’s
The concept of *operative scheme* and the Minsky (1975) *frames theory*, a *driving schema* is a functional 3D-model of the road *Infrastructure* associated with a *Tactical Goal* to be reached in this infrastructure. It is made of a *Driving Path*, defined as a sequence of *Driving Zones*, and integrates a sequence of *Actions* to be progressively implemented when progressing on the path. The decision to implement or not an action depends of *Conditions* to be checked by the driver regarding the occurrence of *Events* in particular *Perceptive Zones* of the road infrastructure. An event is an *Object* with specific *Characteristics* (its aspect, behaviour, or status). Once activated in working memory and instantiated with the characteristics of the current road environment, the active driving schema becomes the *tactical mental representation* of the driver, that is continuously updated as and when s/he progresses on the road. It corresponds to the driver’s situation awareness of the situation. In the frame of a car-following task on straight line as investigated in this paper, the driving schema is focused on the tactical goal of *progressing along the same road lane* (no overtaking), at a given speed, and keeping a safe distance with the lead car.

At the operational level, corresponding to an automatic control loop, COSMODRIVE regulation strategy is jointly based on *envelope zones* and *pure pursuit point* approaches. From a theoretical point of view, the concept of envelope zones comes from two classical theories in psychology: the notion of *body image* of Schilder (1950), and the theory of *proxemics* defined by Hall (1966), relating to the distance keeping in social interactions with other humans. Regarding car-driving activity, envelope zones also refer to *safety margins*. At this last level, COSMODRIVE model (Fig.6) is based on Kontaratos’ work (1974) distinguishing a *safety zone*, a *threat zone*, and a *danger zone*. Envelope zones correspond to the portion of the path of driving schema to be occupied by the vehicle in the near future. As an “hidden dimension” of the social cognition, as suggested by Hall’s theory (1966), these proxemics zones are also mentally projected to other road users, and are then used to dynamically interact with them, as well as to anticipate and manage the collision risks. This “virtual skin” is permanently active while driving, as an implicit awareness of our expected allocated space for moving. As with the Schilder’s body schema, it belongs to a highly integrated cognitive level (i.e. implicit regulation loop), but at the same time, it favors the emergence of critical events in the driver’s explicit awareness. Therefore, the envelope zones play a central role in the regulation of “social” as well as “physical” interactions with other road users under normal driving conditions (e.g., inter-vehicle distance keeping), and in the risk assessment of path conflicts and their management, if a critical situation occurs (commitment of emergency reactions).

Moreover, two Decision-Making processes are implemented in COSMODRIVE model, one for each regulation loops presented in fig. 4. At the attentional level, corresponding to *explicit* decisions, this process is modelling through State-Transition automats intimately linked with the *driving path* and *conditions* integrated in tactical *driving schemas*. In real driving conditions, this tactical level is typically used for overtaking decision-making. However, in the frame of the empirical data collected in our experiment, primarily involving automatic driving abilities, the tactical level is mainly active when the lead car suddenly brakes and when the situation becomes critical. At the automatic level, an *implicit* decision-making is implemented through envelope zones, in order to keep a safety distance with the lead car (i.e. keep it in the green zone).

### 3.3 The Action module

The Action Module is in charge to perform vehicle-control skills, according to the driving actions decided and planned at the representational level by the Cognition module. The two core regulation mechanisms effectively implemented by the Action Module are based on (i) the *Pure-Pursuit Point method* and (ii) safety margin keeping by using *Envelope Zones*. The *Pure Pursuit Point* method is used by COSMODRIVE for the lateral and the longitudinal controls of the car along the driving path of a tactical schema (Mayenobe, 2004). Mathematically, the pure-pursuit point is defined as the intersection of the desired vehicle path and a circle of radius centered at the vehicle’s rear axle midpoint (assuming front wheel steer). Intuitively, this point describes the steering curvature that would bring the vehicle to the desired lateral offset after traveling a distance of approximately 1. Thus the position of the pure-pursuit point maps directly onto a recommended steering curvature: \( k = -\frac{2x}{l} \), where \( k \) is the curvature (reciprocal of steering radius), \( x \) is the relative lateral offset to the pure-pursuit point in vehicle coordinates, and \( l \) is a parameter known as the look-ahead distance. According to this definition, the operational control of the car by COSMODRIVE is a monitoring loop in charge to permanently keep the Pursuit Point in the driving path, to a given speed assigned with each segment of the tactical schema, as instantiated in working memory.

COSMODRIVE abilities for vehicle-control are thus supported in the Action module by the *pure-pursuit point method* (for monitoring the lateral and longitudinal position...
of the car), and by the envelope zones strategies (for managing interactions with the other road users). Figure 6 illustrates this regulation strategy in the frame of a car-following task: the pursuit point determines the cap to be followed by the virtual ego-car, and the envelope zones are used for keeping a safe IVT distance with the lead car.

### 3.4 Simulation of visual distraction effects

By considering the empirical data presented in section 2, the visual scanning patterns of the additional screen collected during this experiment among human drivers (cf. fig 2) were implemented in the Perception module of COSMODRIVE, in order to simulate visual distraction effects on drivers’ behaviors (visual strategies and vehicle control) and to investigate human errors liable to occur when drivers perform a visual secondary task while driving. Indeed, beyond the observable effects of visual distraction on drivers’ performance, the aim of the COSMODRIVE computational modeling approach was also to simulate such distractive impacts on car drivers Situation Awareness.

When driving, drivers must continually update their mental model of the driving situation as and when they dynamically progress on the road. In case of additional task requiring off-road scanning, mental model updating is imperfectly done and driver’s actions are thus based on a mental representation that may dramatically differ of the situational reality, in case of a critical change in the traffic conditions.

This is typically what occurred in the example of crash initially presented in fig. 3, and then analyzed in Figure 8 and 9 from COSMODRIVE simulations. These 2 Figures correspond to a simulation case for a similar driving scenario presented in fig. 3 (free following task). Like 58 % of the observed human drivers, COSMODRIVE implemented here the first visual strategy for scanning of the additional screen (cf. fig. 2), requiring a long glance of 2 seconds. During these 2 seconds, the model manages the IVT with the lead car by using its mental representation of the driving situation (see stages 2 on fig. 9). Unfortunately, the lead car brakes when the virtual eye is off-road and COSMODRIVE Situation Awareness progressively becomes very different of the situational reality (stage 3 on Fig. 9). When the driver/model repays attention to the road scene (view c on Fig.8 and stage 4 on Fig. 9), they suddenly become aware of the critical gap between the expected lead car position (as mentally assessed during the off-road glance by assuming a steady speed of the lead car) and the critical nature of the objective reality (as illustrated at stage 4 on fig. 9). Therefore, like the human driver presented in fig. 3, the model immediately carried out an emergency braking (reaction time of 0.8 sec. on Fig 8), but the crash cannot be avoided.

![Fig. 8: driving performance simulation of a distracted driver](image)

![Stage 1: car following situation (left) and driver’s SA (right)](image)

![Stage 2: beginning of driver’s on-board screen visual scanning](image)

![Stage 3: Lead car braking while the driver is visually distracted](image)

![Stage 4: driver’s eyes go back on road and she becomes aware of erroneous updating of her mental model while distracted](image)

![Stage 5: Crash (due to a too late reaction time)](image)

Figure 9: simulations of visual distraction effect on driver’s SA

### 3.4 Conclusion and perspectives

As illustrated in Fig. 9, this type of simulation based on COSMODRIVE allow us to in-depth investigate and understand what happens in the driver’s mind when visually distracted: incomplete or incorrect perception of roadway cues, due to off-road glances required by the secondary task, directly impacts the formulation of an adequate mental model (i.e. Situation Awareness) that will affect, in a second times, the decision making and the driving performance.

From these cognitive simulation abilities, it is expected in the future to explore visual distraction effects for a large set of driving scenarios, more particularly in terms of
inadequate mental model, that is of a crucial interest for analysing human errors at both behavioural and cognitive levels, or for explaining some involuntary risk taking of distracted drivers.

Acknowledgments
The research was supported by the European Commission 7th Framework Program, in the frame of ISI-PADAS Project.

References


