1 Empirical Longitudinal Driving Behaviour in case of Authority Transitions between Adaptive

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1 ABSTRACT

Automated vehicles are expected to have a substantial impact on traffic flow efficiency, safety levels and levels of emissions. However, Fields Operational Tests suggest that drivers may prefer to disengage Adaptive Cruise Control (ACC) and resume manual control in dense traffic conditions and before performing manoeuvres such as lane changing. These so-called authority transitions can have substantial effects on traffic flow. To gain insight into these effects, a better understanding is needed of the relationships between these transitions, longitudinal dynamics of vehicles and behavioural adaptations of drivers.

8 In this context, a driving simulator experiment was set-up to gain insight into the effects of 9 authority transitions between ACC and manual driving on longitudinal dynamics of vehicles. Participants 10 were assigned to one of three conditions randomly. In the control condition, participants drove manually. In 11 the first experimental condition, a sensor failure was simulated at a specific location after which drivers 12 were expected to resume manual control. In the second experimental condition, drivers switched ACC off 13 and on pressing a button whenever they desired.

Statistical tests indicate that the distributions of speed, acceleration and time headway significantly differ between the three conditions. In the first experimental condition, the speed drops after the sensor failure and the time headway increases after the discretionary re-activation of ACC. These results seem to be consistent with previous findings and suggest that authority transitions between ACC and manual driving influence significantly the longitudinal dynamics of vehicles, potentially mitigating the expected

- 19 benefits of ACC on traffic flow efficiency.
- *Keywords*: Authority transitions, Adaptive Cruise Control, driving simulator experiment, longitudinal
 driving behaviour.

1 INTRODUCTION

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In recent years, interest in automated vehicles and systems supporting the drivers in their control task has increased. Automated vehicles are expected to have a significant impact on traffic flow efficiency, safety levels and the environment. These vehicles, in particular those that can show cooperative behaviour, are expected to reduce congestion levels because they will help to increase road capacity, anticipate traffic conditions downstream and increase the outflow from a queue [1].

The introduction of automated vehicles on public roads is likely to be gradual: the functionalities of
automated systems are introduced through intermediate steps. SAE International [2] defines the different
levels of automation as follows:

- Level 0: manual driving;
 - Level 1: driving assistance;
 - Level 2: partial automation;
 - Level 3: conditional automation;
 - Level 4: high automation;
- 15 Level 5: full automation.

16 At the driving assistance level, the system takes over either the longitudinal or the lateral control. For example, Adaptive Cruise Control (ACC) is a driver assistance system providing support in longitudinal 17 18 control through maintaining a desired speed and time headway. In partial automation, the system takes over 19 longitudinal and lateral control, while the driver permanently monitors the system and is expected to 20 resume control at any time. In conditional automation, the system takes over longitudinal and lateral 21 control, while the driver does not have to continuously monitor the system and is expected to resume 22 control in case of an emergency (e.g., sensor failure). In high automation, the system takes over longitudinal 23 and lateral control, even if the driver does not respond adequately to a request to intervene in case of certain 24 roadway and environmental conditions. In full automation, the system full-time takes over longitudinal and 25 lateral control under all roadway and environmental conditions. The driver is not required to monitor the 26 system.

Under certain traffic situations, however, drivers might disengage the automated system because they *prefer* to transfer to a lower level of automation (or manual driving) [3] or *are forced* to do so, for instance due to a sensor failure [4]. These transitions between different levels of automation are called *authority transitions*. These transitions can significantly affect the longitudinal and lateral dynamics of vehicles and are consequently expected to have a considerable impact on traffic flow efficiency (e.g., traffic flow stability).

To ex ante evaluate the impact of automated vehicles on traffic flow efficiency at varying penetration rates, mathematical models of driving behaviour of manually driven and automated vehicles can be implemented in microscopic simulation software packages [5, 6]. Currently, most mathematical models describing car-following and lane changing behaviour do not account for the possibility to switch the automated system on and off and are therefore not adequate in representing these transitions. Thus, an extension of these models is required. However, in order to do so, a better understanding is needed into how authority transitions affect the lateral and longitudinal dynamics of vehicles.

40 The aim of this research is to provide insight into the theory and empirics of longitudinal driving 41 behaviour in case of authority transitions between ACC and manual driving. The main contribution of this 42 paper is to explore the effects of authority transitions on longitudinal dynamics through extensive statistical analyses of data obtained through a driving simulator experiment. Participants were asked to drive in a 43 44 vehicle equipped with ACC on a virtual two-lane highway using a medium fidelity fix-based driving 45 simulator at Delft University of Technology. Speed, acceleration, distance and time headway, lateral position and lane changes were measured through registered behaviour. In a control condition, participants 46 47 were required to drive manually. In the first experimental condition, a sensor failure was simulated at a 48 specific location after which the driver was required to resume manual control over the vehicle. In the 49 second experimental condition, the drivers were allowed to switch the system off and on voluntarily.

1 The paper is structured as follows. The next section reviews possible reasons for authority transitions. This

2 section is followed by a description of the driving simulator experiment. Next, the results of the experiment

3 are discussed in-depth through statistical analyses. The final section discusses the effects of authority

4 transitions on longitudinal driving behaviour and presents the limitations of the proposed approach, while

5 also suggesting recommendations for future research.

6 LITERATURE REVIEW

Before investigating the effects of authority transitions on longitudinal dynamics, it is essential to discuss the possible motivations that trigger the transitions. In this section, possible reasons for authority transitions between ACC and manual driving are proposed based on available literature. In order to explore the potential effects of the authority transitions on driving behaviour, we introduce an overview of the available research on behavioural adaptations and the changed role of the driver with ACC. After that, potentialities and limitations of data collection methods such as Field Operational Tests (FOTs) and driving simulator

13 experiments are discussed.

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14 Mandatory and discretionary authority transitions with ACC

The authority transitions appear to be strongly related to the characteristics of the driver support system. For example, FOTs [3] investigated driving behaviour with types of ACC systems that have limited decelerations capabilities and are inactive at speed below 30 km/h. Drivers prefer to disengage ACC and resume manual control during dense traffic conditions in order to have smaller distance headways. In medium – dense traffic conditions, drivers tend to deactivate the system to have full control of the vehicle (e.g., in case of overtaking manoeuver).

Discretionary authority transitions are defined as situations in which drivers disengage the automated system voluntarily. The most common motivations to initiate a discretionary authority transition with the above-mentioned types of ACC are presented below [5, 7, 8]:

- *Speed adaptation prior to a lane change manoeuver:* the driver plans to make a lane change and the current acceleration is not adequate;
- Overruling due to defensive or offensive behaviour: the driver brakes (or accelerates) to create a sufficient (or insufficient) gap for a vehicle in an adjacent lane for merging;
- *Left-lane speed adaptation*: the driver brakes to avoid illegal overtaking on the right and to adapt to the speed of the vehicle in the adjacent lane.

The authority transitions are defined as *mandatory* when drivers are forced to switch off by the system because of the functioning of the driver support system. Possible reasons for mandatory authority transitions with the above-mentioned types of ACC are presented below [5, 7, 8]:

- A sensor failure: the sensor cannot work properly (e.g., poor visibility due to adverse weather conditions) and the driver has to resume manual control;
 - *Reaching the system support constraints in a safety-critical situation:* the system support constraints in speed and acceleration are reached; however, the driver needs to exceed these limits in order to avoid collision or overtake.

38 Behavioural adaptations and changed role of the driver with ACC

Adaptations in driving behaviour are defined as the collection of behavioural aspects that arise following a change in road traffic [9]. For instance, the influence of ACC systems activated on the longitudinal driving behaviour of drivers has been extensively investigated since the 90s. Similarly, there has been an interest in Automated Highway System (AHS) which takes over the longitudinal control of vehicles driving in an

42 Automated Highw43 automated lane.

44 Driving simulator studies have found that ACC systems lead to more collisions than unsupported 45 driving, for instance when the drivers have to resume manual control because the deceleration is not 46 sufficient to avoid collision while approaching a stationary queue [10] and the system fails accelerating

unexpectedly towards the vehicle in front [11]. Recently, a driving simulator study pointed out that in case 1 2 of deceleration failures with ACC the mean reaction time of drivers varies between 1.60 s and 2.26 s, 3 depending on the magnitude of the deceleration failure, and concluded that humans are poor monitors of 4 automation [12]. Driving simulator studies have investigated the transfer of control between the AHS and 5 the driver of a vehicle entering and exiting an automated lane [13]. In the latter, drivers were warned 60 s in 6 advance before exiting the automated lane and resuming manual control. The authors concluded that these 7 transitions resulted in an unacceptable rate of incomplete lane changes and collisions. In addition, it appears 8 that ACC systems, which automatically regulate the speed when the vehicle gets too close to the leader, 9 result in higher speeds and shorter time headways when they are active [14, 15]. However, little attention 10 has been paid to the influence of mandatory and discretionary authority transitions as defined above on the 11 longitudinal dynamics of vehicles and the behavioural adaptations of drivers.

12 The effects of ACC on driving behaviour may be related to the changed role of the driver, who is 13 transformed from a manual controller to a supervisor of the system [1]. Indeed, automated vehicles require 14 drivers who are capable to resume control in case of authority transitions. Studies in the field of aviation 15 have suggested that monitoring the system for long periods of time might increase the workload of the 16 driver [16] which can result in a reduction in situation awareness and a failure in the detection of critical 17 changes in the state of the system [17]. In addition, indirect adaptation effects may be due to over-reliance 18 on the system, which is defined as the tendency of human supervisors to place too much trust in automated 19 systems [18]. In the road transport field, similarly, some driving simulator studies have found a reduction in 20 situation awareness [19] and very low levels of self-reported mental workload [20] while driving with 21 ACC.

22 Data collection methods

23 The validity of data collected in a FOT can be considered relatively high while the level of controllability is 24 limited [21]. Indeed, in a FOT it is not possible to present exactly the same conditions to all the participants 25 and therefore precisely control for potential confounding variables. Vice-versa, driving simulators possess a high degree of controllability. Presenting exactly the same traffic flow and environmental conditions to all 26 27 the participants, driving performances can be assessed objectively [22]. Since reality is represented 28 virtually, driving simulator experiments can result in a reduction in validity. However, recent findings [23] 29 suggest that driving simulator studies possess relative validity, which means that the observed behavioural 30 response converge in the same direction but not with the same magnitude as in real life.

31 The studies found [3, 5, 7, 8] point out that drivers may prefer to disengage ACC and resume 32 manual control for many traffic situations (e.g., dense traffic conditions, lane change manoeuver, gap 33 creation, left-lane speed adaptation). Moreover, the system can switch off due to sensor failure or system 34 support constraints reached. These studies were based on data collected in a FOT. In addition, ACC is 35 assumed to reduce driver's vigilance and situation awareness. Therefore, we may conclude that ACC can 36 compromise driver's ability to respond in case of emergency situations and sensor failure. Most of the 37 studies on the changed role of the driver in relation to automation were performed using driving simulator studies or were conducted in the field of aviation. Based on current literature, we miss a driving simulator 38 39 experiment analysing the influence of the above-defined mandatory and discretionary authority transitions 40 on the longitudinal dynamics of vehicles and the behavioural adaptations of drivers. Given the importance 41 of understanding this transitional process and its implications on driving behaviour, in this research we 42 focus on acquiring such data.

43 **RESEARCH METHOD**

In this paper, we aim at gaining in-depth insight into driving behaviour during authority transitions through a driving simulator experiment. The main objective is to analyse to what extent authority transitions

46 between ACC and manual driving affect the dynamics of vehicles. The behavioural assumption we would

47 like to test is that the authority transitions between ACC and manual driving cause significant changes in

48 speed, acceleration and time headway. In addition, we would like to explore the variations in the responses

49 of drivers during mandatory and discretionary authority transitions. To study this, we use an experiment

1 with a high degree of controllability.

In this section information is provided on the driving simulator as well as the driving environment designed for the purpose of this study. Next, the experimental design is discussed. In addition, the data collected to approximate the longitudinal driving behaviour and a description of the participants are provided.

6 The driving simulator and the driving environment

A medium-fidelity fixed-based driving simulator, which is displayed in Figure 1, was used in the experiments. This simulator was chosen because of availability reason. The simulator is composed of a steering wheel, pedals and gear stick which are obtained from a real car. Three screens which are placed at an angle of 120° show outside world images, the dash-board, the interior of the vehicle and the mirrors. The simulator provides a visual field of view of 180° horizontally and 45° vertically. The software was developed by StSoftwareTM [24]. The gearbox was set to automatic.

13 For this experiment, we developed a driving environment composed of two main parts (7 km in 14 total). The first part consisted of a test run (2 km) in an urban environment. In this phase, all the participants 15 drove manually and the use of ACC was not possible. The aim was to accustom the participants to driving 16 in the driving simulator as well as to check for simulator sickness. The second part, which is displayed in Figure 1, consisted of two segments (2 km each) of a virtual freeway with two lanes in each direction, 17 18 connected by a one lane stretch (1 km). In this research, only the data collected in the two highway 19 segments were analysed. The speed of the surrounding vehicles was programmed to vary randomly in the 20 intervals (80, 85) and (110, 115) km/h in the left lane and (120, 125) km/h in the right lane. These vehicles 21 changed lane when the speed of their leader was lower than their own speed. When ACC was switched on, 22 the speed was set to 120 km/h (i.e., the speed limit) and the time headway was set to 1.5 s, without any 23 possibilities to regulate the system settings.





FIGURE 1 The medium-fidelity fixed-based driving simulator (a) and the driving simulation environment in the two-lane highway (b).

28 Experimental design

29 The experiment consisted of a control condition and two experimental conditions, making up a *complete* 30 three group independent samples randomized experimental design. The driving environment and the characteristics of the surrounding vehicles were exactly the same for each condition. In the control 31 condition (CC), authority transitions were not possible by definition and the participants had to drive 32 33 manually. In the first experimental condition (EC1), ACC was switched on automatically after merging into 34 the highway and the drivers were informed by a message on the screen ('ACC is switched on'). On the 35 second stretch of the highway, a sensor failure was simulated at a predefined location and the system automatically disengaged by decelerating the vehicle. The driver was warned by a message on the screen 36 37 ('Sensor failure!') and was expected to resume manual control. At the next location, another message 38 appeared on the screen ('Sensors are ok!') after which it was possible to switch on ACC again. In the second

experimental condition (EC2), the drivers were allowed to switch ACC off and on by using a button on the
 dashboard discretionary.

3 Participants and data collection

4 The participants were assigned randomly to one of the above-mentioned groups. Seventy-five participants 5 were recruited among the male and female inhabitants of Delft between the ages of 20 and 72 years old. A 6 valid driving license and more than one year of driving experience were considered as a prerequisite.

7 Before the experiment, participants received written instructions on the general scope of the 8 research, the features of the driving simulator and the potential risks related to simulator sickness. 9 Participants were asked to drive as in real life and allowed to overtake. In addition, they were informed on 10 the characteristics of the ACC available and warned to monitor the system and be able to resume manual control at any time. However, the precise scope of the experiment (i.e., analysing driving behaviour during 11 authority transitions) was not communicated. After that, a written informed consent was signed. The whole 12 13 procedure was executed following the regulations of the ethics committee of Delft University of 14 Technology.

15 The duration of the experiment varied between 8 and 20 minutes, depending on the participants. 16 After that, we asked to complete a questionnaire in which they reported demographic characteristics, driving experience, previous experience with cruise control and ACC in real life, information related to 17 18 driving styles and mental workload experienced. Eight participants were not able to complete the experiment due to simulator sickness. Statistics regarding the characteristics of the participants who 19 20 successfully completed the experiment are reported in Table 1. The analysis of the full questionnaire is not 21 included due to space limitations. The two-sample Kolmogorov-Smirnov test is performed in order to 22 determine whether the three groups come from the same population. The null hypothesis that the 23 distributions of the variables gender, age and driving experience in the three groups come from the same 24 distribution cannot be rejected at the 5% significance level. This means that the distributions of gender, age 25 and driving experience do not differ significantly between the three groups. The test statistics are presented 26 in Table 1.

Longitudinal driving behaviour was measured through registered data in the driving simulator.
 Speed, acceleration, distance and time headway, lateral position and lane changes were measured at a
 sampling rate of 10 Hz. Sixty-seven complete observations were collected and analysed in this paper.

TABLE 1 Statistics on the participants' characteristics in the Control Condition (CC), the Experimental Condition 1 (EC1) and the Experimental Condition 2 (EC2). *Female* is a variable which is equal to 1 when the participant is a female and 0 otherwise.

			CC			EC1			EC2
Participants (N)			25			21			21
Gender (n_{male}, n_{female})	1	4	11	1	5	6	12	2	9
Age (M _{years} , SD _{years})	47.0	8	14.75	38.1	0	11.52	39.19)	12.86
Driving experience (M_{years}, SD_{years})	24.7	2	12.79	19.38		11.22	21.55	5	13.86
		(Gender			Age	Drivir	ig expe	rience
		(f	emale)		((years)		(:	years)
Two-sample Kolmogorov-Smirnov	CC	CC	EC1	CC	CC	EC1	CC	CC	EC1
test	EC1	EC2	EC2	EC1	EC2	EC2	EC1	EC2	EC2
<i>p</i> -value	0.93	1.00	1.00	0.07	0.19	0.19	0.11	0.57	0.57
Test statistic	0.15	0.01	0.17	0.37	0.31	0.19	0.34	0.22	0.17

33 DATA ANALYSIS METHOD

34 The distributions of speed, acceleration and time headway are analysed in order to study the longitudinal

35 dynamics of vehicles in case of authority transitions between ACC and manual driving. The behavioural

hypothesis tested is that authority transitions between ACC and manual driving cause significant changes in speed, acceleration and time headway. After that, the characteristics of the mandatory and discretionary authority transitions in EC1 are investigated in terms of time needed to resume control and the consequent speed variation. Finally, a detailed analysis of the driving behaviour of two single drivers in EC1 is presented and discussed in-depth. Here, authority transitions are investigated by using a relative speed-spacing plane.

7 Analysis of speed, acceleration and time headway distributions

8 The outputs of the driving simulator are processed for each participant and the values of the variables speed, 9 acceleration and time headway (rear bumper of the leader – front bumper of the follower) are calculated 10 every two meters. For each location, the mean and the standard deviation of these variables are calculated 11 between the participants in each condition. The distributions are plotted in Figure 2.

12 A one-sample Kolmogorov-Smirnov test is performed to check if the mean and standard deviation 13 of the variables calculated as a function of distances in each condition are normally distributed. The null 14 hypothesis that the distributions of the mean and standard deviation of speed, acceleration and time 15 headway are normally distributed is rejected at the 5% significance level. After that, the two-sample 16 Kolmogorov-Smirnov test is performed in order to understand if the mean and standard deviation of the variables are homogenous between the three groups. The null hypothesis that the mean and standard 17 18 deviation of speed, acceleration and time headway in the three groups come from the same distribution is 19 rejected at the 5% significance level. The p-value and the test statistic are reported in Table 2. The results 20 indicate that the longitudinal dynamics of the vehicles differ significantly between the three conditions. The 21 largest difference in speed and time headway can be found between the CC and the EC1.

22 Comparing the CC and the EC2 in Figure 2, the speed distributions seem to be similar in terms of 23 mean and standard deviation. This result appears to be consistent with findings by Klunder et al. [5]. 24 However, analysing the first segment of highway, in the CC the mean and the standard deviation are 25 generally constant over the distances, while in the EC2 the mean speed increases progressively and the 26 standard deviation decreases. These results seem to be consistent with the fact that more drivers switched 27 on ACC over time. The distributions of the mean accelerations are similar in both segments. However, the 28 possibility to switch ACC on and off discretionary can lead to higher variability between the drivers and 29 therefore to a higher standard deviation. In the EC2, the mean and standard deviation of the time headway 30 distribution are generally smaller and clearly decrease over distance in the first segment. This can be 31 interpreted as an adaptation effect related to switching ACC on and off.

32 Comparing the EC1 to the CC and the EC2 in Figure 2, the use of ACC results in higher mean 33 speeds and a lower standard deviation in the first segment where authority transitions are not possible. After 34 the sensor failure, it is possible to note a significant drop in speed and increase in the standard deviation of 35 speed, as a result of the different responses of drivers. A second drop in speed can be recognized after the message informing the drivers that ACC could be switched on again. Significant changes in mean values of 36 37 acceleration can be noted corresponding to the aforementioned authority transitions. Small mean time headway can be observed in the first segment of the highway, while higher mean values can be found in the 38 39 second segment. After sensor failure, indeed, the time headways increase, reaching values similar to these 40 observed during manual driving. However, after the sensors are functioning again and thus it is possible to 41 switch ACC on voluntary, higher time headways can be recognized. The latter result appears to be 42 consistent with findings by Pauwelussen and Minderhoud [7] and Pauwelussen and Feenstra [8]. Higher 43 time headways can potentially have a negative effect on traffic flow efficiency.



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2 FIGURE 2 Mean (continuous line) and standard deviation (dashed line) of speed (a), acceleration (b) 3 and time headway (c) distributions calculated as a function of the distance travelled since the 4 beginning of the simulation for the Control Condition (blue), the Experimental Condition 1 (green) 5 and the Experimental Condition 2 (red). The curve lines separate the first and the second segment of 6 the highway. For each segment, drivers enter and exit the highway through on and off-ramps. The 7 first dashed black line (distance= 5480 m) indicates the location where sensor failure is simulated. 8 After sensor failure, drivers are expected to resume manual control. The second dashed black line 9 (distance= 5981 m) indicates the location after which it was possible to switch on ACC again.

1 TABLE 2 Statistics on the speed, acceleration and time headway distributions calculated as a 2 function of the distance travelled in the first and the second segment of highway for the Control 3 Condition (CC), the Experimental Condition 1 (EC1) and the Experimental Condition 2 (EC2). In 4 EC1, the sensor failure is simulated in the second segment of highway.

	Speed (km/h)			Acceleration (m/s^2)			Time headway (s)		
Mean of mean values over distances	CC	EC1	EC2	СС	EC1	EC2	CC	EC1	EC2
First segment	99.86	113.13	105.09	-0.02	0.16	0.03	3.07	1.30	2.36
Second segment	104.22	107.25	108.10	-0.06	-0.03	-0.01	2.21	2.10	1.66
Overall	102.15	110.05	106.67	-0.04	0.06	0.01	2.62	1.72	1.99
One-sample Kolmogorov-Smirnov test	CC	EC1	EC2	CC	EC1	EC2	CC	EC1	EC2
<i>p</i> -value	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Test statistic	1.00	1.00	1.00	0.35	0.32	0.30	0.88	0.80	0.85
Critical Value	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03
Two-sample	CC	CC	EC1	CC	CC	EC1	CC	CC	EC1
Kolmogorov-Smirnov test	EC1	EC2	EC2	EC1	EC2	EC2	EC1	EC2	EC2
<i>p</i> -value	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Test statistic	0.73	0.51	0.73	0.17	0.19	0.11	0.53	0.51	0.39
Mean of std. dev. values over distances	CC	EC1	EC2	СС	EC1	EC2	CC	EC1	EC2
First segment	18.59	5.28	13.65	0.50	0.31	0.62	2.43	0.60	1.38
Second segment	14.28	10.29	12.71	0.38	0.58	0.51	1.27	1.15	0.88
Overall	16.33	7.91	13.16	0.44	0.45	0.56	1.81	0.89	1.12
One-sample Kolmogorov-Smirnov test	CC	EC1	EC2	CC	EC1	EC2	CC	EC1	EC2
<i>p</i> -value	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Test statistic	1.00	0.98	1.00	0.55	0.51	0.54	0.76	0.57	0.61
Critical Value	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03
Two-sample	CC	CC	EC1	CC	CC	EC1	CC	CC	EC1
Kolmogorov-Smirnov test	EC1	EC2	EC2	EC1	EC2	EC2	EC1	EC2	EC2
<i>p</i> -value	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Test statistic	0.78	0.41	0.52	0.27	0.26	0.27	0.48	0.41	0.22

1 Analysis of authority transition in case of sensor failure (EC1)

2 In this section, the "time to resume control" and the "resulting speed variation" in case of authority

- transitions are analysed for each participant n in the EC1. The sensor failure triggers a mandatory authority
- transition between ACC and manual driving (i.e., mandatory switching off). After that, the possibility to
 reactivate ACC can lead to a discretionary authority transition between manual driving and ACC.
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6 Mandatory switching off action

7 The time necessary to resume manual control $T_{RMC,n}$ in case of mandatory authority transitions is defined as

- 8 the interval between the instant of sensor failure $T_{SFL,n}$ and the instant when the gas pedal is pressed again 9 $T_{GPP,n}$. The distribution of $T_{RMC,n}$ is presented in Figure 3.
- 10 Assuming that T^* is the median value of $T_{RMC,n}$, the median of the speed variation distribution ΔV_n 11 which occurs during the authority transition is calculated as follows in equation (1):

$$\Delta V = median(V_n^* - V_{SFL,n}), \tag{1}$$

12 Where

13 V_n^* is the speed at the instant T^* for each participant *n*;

14 $V_{SFL,n}$ is the speed at the instant of the sensor failure for each participant *n*.

15 Discretionary switching on action

16 The time necessary to resume automatic control $T_{RAC,n}$ in case of discretionary authority transition is defined

17 as the interval between the instant when the sensors are functioning again $T_{SFC,n}$ and the instant when ACC

18 is switched on again by pressing the button $T_{ACC ON,n}$. The distribution of $T_{RAC,n}$ is presented in Figure 3. Two

19 participants did not switch ACC on after the sensors were functioning again.



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FIGURE 3 Time to resume manual control $T_{RMC,n}$ after sensor failure (blue) and time to resume automatic control $T_{RAC,n}$ after sensors are functioning again (light blue).

The speed variation distribution ΔV_n is calculated similarly as described in equation (1). Statistics on the speed variation ΔV_n , the time to resume manual control $T_{RMC,n}$ and automatic control $T_{RAC,n}$ are reported in Table 3.

TABLE 3 Statistics on the distributions of time to resume control and speed variation in case of authority transition

	Tim	e to resume	e control (s)	S	peed varia	tion (km/h)
EC1	min	max	median	min	max	median
Mandatory switching off	1.70	14.50	3.85	-20.37	-8.34	-18.18
Discretionary switching on	1.40	31.40	5.80	-26.00	5.42	-4.22

1 When observing the time to resume control, it can be pointed out that the minimum value is lower in case of

2 discretionary switching on. However, the discretionary switching on results in a higher median value of

3 time to resume control due to the larger variability in the response of drivers. It is interesting to note that in

4 both cases the authority transitions result in a negative median speed variation, which can have considerable

5 effects on traffic flow. The mandatory switching off always implies a negative speed variation while the

6 discretionary switching on can lead to positive or negative speed variation, depending on the response of

7 the drivers.

8 Analysis of longitudinal dynamics of single drivers (EC1)

9 In this section, the longitudinal dynamics of two individual drivers (driver-1 and driver-2) in EC1 are 10 analysed in detail. The scope is to confirm and examine in-depth the general results found for the whole 11 sample.

In Figure 4 (a) – (b) speed, acceleration and time headway distributions are calculated as a function of distance travelled since the beginning of the simulation. In addition, the relative speed $dv = v_{i,1} - v_i$ to the leader *i*-1 and the distance headway $s = x_{i,1} - x_i$ (rear bumper of the leader – front bumper of the follower) are calculated and plotted in a (dv, s) plane in Figure 4 (c) – (d). When no leader was present, the data were discarded. In these (dv, s) planes, four different phases are distinguished following the definitions proposed

- 17 in the previous section:
- 18 1. ACC before sensor failure (driver-1 and driver-2);
- 19 2. Authority transition after mandatory switching off (driver-1 and driver-2);
- 20 3. Manual driving after resuming control (driver-1 and driver-2);
- 21 4. ACC after discretionary switching on (driver-2).

Constant acceleration periods could be clearly recognized. The duration of these periods is not fixed but is related to the state of the follower in relation to the leader. It can be assumed that the transitions between the above-mentioned different phases correspond to an action of the follower who wants to increase or decrease the acceleration.

When driving with ACC, periods of constant relative distance can be identified. The system tends to reduce the relative speed to zero. Here, discontinuities in the plots correspond to changes in the leader and consequently rapid variations in the acceleration. After the sensor failure, the vehicles decelerate uniformly and the relative speeds increase, until the drivers resume control and start to press the gas pedal again.

When the vehicle is driven manually, an oscillation of the vehicle motion around states with a relative velocity equal to zero can be recognized [25]. It is interesting to note the same phenomenon cannot be identified during authority transitions and with ACC, which reacts to small speed differences.

Driver-1 did not switch on again ACC after resuming control. When driver-2 decided to switch on again ACC, the relative speed and distance headway increased compared to ACC before sensor failure, meaning that the gap to the leader increased in space and speed.



3 FIGURE 4 Speed, acceleration and time headway distributions calculated as a function of the 4 distance travelled since the beginning of the simulation in the Experimental Condition 1 (EC1) for 5 driver-1 (a) and driver-2 (b). The curve lines separate the first and the second segment of the 6 highway. For each segment, drivers enter and exit the highway through on and off-ramps. The first 7 dashed black line (distance= 5480 m) indicate the location where sensor failure is simulated. After 8 sensor failure, drivers are expected to resume manual control. The second dashed black line 9 (distance= 5981 m) indicates the location after which it was possible to switch on Adaptive Cruise 10 Control (ACC) again. The (dv, s) planes in EC1 are reported for driver-1 (c) and driver-2 (d). Four phases are distinguished: ACC before sensor failure (red); authority transition after mandatory 11 12 switching off (black); manual driving after resuming control (green); ACC after discretionary 13 switching on and off (magenta). Each dot corresponds to a time step.

1 CONCLUSION AND FUTURE RESEARCH

The available literature indicates that drivers may prefer to disengage ACC and resume manual control in dense traffic conditions and to perform manoeuvres such as lane changing. Authority transitions can have significant effects on traffic flow. However, these studies rely on data collected in FOTs and thus little insight is available on the relationships between the mandatory and discretionary authority transitions dentified, longitudinal dynamics of vehicles and behavioural adaptations of drivers.

In this paper, an in-depth insight is gained into the influence of these transitions between ACC and manual driving on longitudinal dynamics of vehicles. For this purpose, a driving simulator experiment was set-up. Participants were asked to drive a vehicle equipped with ACC on a virtual two-lane highway. In a control condition (CC), participants were required to drive manually. In the first experimental condition (EC1), a sensor failure was simulated at a specific location after which the driver was required to resume manual control. In the second experimental condition (EC2), the drivers were allowed to switch the system off and on voluntarily.

14 The distributions of speed, acceleration and time headway are analysed for each group. Statistical 15 tests indicate that these variables significantly differ between the three conditions. Comparing the CC and 16 the EC2, the speed distributions seem to be similar in terms of mean and standard deviation. Looking at the 17 EC1, the use of ACC results in higher mean speeds and lower standard deviation in the first segment of the 18 highway where authority transitions are not possible. After the sensor failure, it is important to note a 19 significant drop in speed (ΔV =-18.18 km/h) and increase in the standard deviation of speed, following from 20 the different responses of drivers. The median time to resume control after sensor failure is equal to 3.85 s. 21 Notably, a similar speed drop is recognizable when the system can be voluntary switched on again 22 $(\Delta V = -4.22 \text{ km/h})$. The median time before voluntary switching ACC on after the message is equal to 5.80 s. 23 Small mean time headways (1.30 s) can be observed in the first segment of the highway, while higher mean 24 values (2.10 s) can be found in the second segment where the sensor failure is simulated. Authority 25 transitions seem to result in higher time headways than these observed when ACC is activated permanently.

The results suggest that authority transitions between ACC and manual driving have significant effects on the longitudinal dynamics of ACC vehicles that can lead to negative effects on traffic flow efficiency in mixed traffic condition, such as traffic flow instability, an increase in congestion levels and a slower clearance of congestion. In addition, these outcomes seem to be consistent with studies found in literature where data from FOT were analysed [3, 5, 7, 8]. Therefore, the assumed relative validity of driving simulator experiments [23] seems to be confirmed.

32 The driving simulator appears to be a useful instrument to do an in-depth investigation of the effects 33 of authority transitions on longitudinal dynamics with a high level of controllability. However, further 34 analysis is necessary to better understand the role of discretionary authority transitions and to validate the results obtained in the driving simulator experiment by using data from FOTs. A limitation of this study is 35 36 that participants drove for a very short period of time and because of this, little insight is gained on the variations within drivers. In addition, these results are related to light traffic flow condition and cannot be 37 38 directly generalized to dense traffic flow. Further research directions might be as follows. First, the analysis 39 of driving behaviour could be extended to lateral dynamics. Second, more work is needed in order to assess 40 the performances of current mathematical models during authority transitions. Third, new mathematical 41 models which account for authority transitions should be developed and the effects on traffic flow should be 42 investigated by using microscopic simulations. Fourth, the research could be extended to investigate 43 authority transitions in case of partial and high automation. 44

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