Structuring Cooperative Behavior Planning Implementations for Automated Driving

Simon Ulbrich, Simon Grossjohann, Christian Appelt, Kai Homeier, Jens Rieken, and Markus Maurer

Abstract—Cooperative behavior planning for automated vehicles is getting more and more attention in the research community. This paper introduces two dimensions to structure cooperative driving tasks. The authors suggest to distinguish driving tasks by the used communication channels and by the hierarchical level of cooperative skills and abilities. In this manner, this paper presents the cooperative behavior skills of "Jack", our automated vehicle driving from Stanford to Las Vegas in January 2015.

I. INTRODUCTION

The issue of cooperative driving and behavior planning has recently become a trending topic in the automated driving research community. While it is already hard to come up with a definition of what cooperative behavior or cooperative driving actually is, it is even harder to contrast it against what was done in previous decades in the field of tactical driving behavior planning.

In this paper, the authors develop a structure to classify cooperative driving tasks. On the one hand, an essential difference for cooperative driving tasks is the communication or awareness channel being used. On the other hand, the level of cooperative behavior skills and abilities may differ. Based on this, the authors develop a matrix to organize and cluster cooperative driving behavior tasks.

We pinpoint the challenges for cooperative driving due to the limited availability of appropriate communication and awareness channels. Moreover, we derive scenarios for cooperative behavior planning. Based on this, we discuss the cooperative behavior skills implemented in "Jack", our automated vehicle driving from Stanford to Las Vegas for this year’s Consumer Electronics Show.

This paper is structured as follows: In section II, we review already existing definitions for cooperative behavior planning in automated driving and introduce the aspect of the employed communication channel. Based on this, the authors develop our so called cooperation matrix to cluster common cooperative behavior planning tasks in section III. Based on a survey in section IV, the authors present the cooperative behavior skills of Jack in section V. Last of all, section VI finalizes this paper with conclusions and a research outlook.

II. RELATED WORK AND TERMINOLOGY

In robotics, the idea of cooperative behavior planning has been addressed and reviewed in many publications already, compare e.g. [1], [2], [3], [4].

In psychology, Spieß [5] defines cooperation as a form of societal collaboration between persons, groups and institutions, or respectively as social interaction. Spieß stresses that cooperation entails conscious and planned acting, as well as processes of mutual coordination about specific objectives. Cooperation premises on fair conditions of collaboration and reciprocity.

Cao et al. [1] defines “[…] a multiple-robot system displays cooperative behavior if, due to some underlying mechanism (i.e., the mechanism of cooperation), there is an increase in the total utility of the system.” According to Franklin’s broad definition of cooperation in [6], a multi-agent system is independent if each agent pursues its own agenda independently from others. Agents are called discrete if the agendas of both agents do not bear any relationship with each other. Anyhow, from an observer’s point of view, agents can cooperate with no intention of doing so. If every agent is simply carrying out its individual behavior without the specific intention of cooperation, some emergent cooperation may still become visible for an outside observer.

Apart from independent multi-agent systems, Franklin [6] identifies cooperative systems as those in which the agendas of the agents explicitly include cooperation with other agents. He further differentiates between deliberative and negotiating communicative and non-communicative systems. Communicative multi-agent systems are intentionally sending and receiving signals from other agents and jointly plan their actions, or even compete for resources. Non-communicative agents coordinate their cooperative activity by observing and reacting to the behavior of others.

Norman uses a much sharper definition of cooperation in [6]. He defines that cooperating means “to act with another or others for a common purpose and for common benefit”. This definition embraces three aspects. The aspect of acting together, a common purpose and a common benefit.

The DFG priority program proposal [7, p. 5] differentiates between using implicit cooperation only and using implicit together with explicit cooperation. The program proposal avoids a clear definition of explicit and implicit cooperation. Rather more, it illustrates the differentiation by two examples: According to [7, p. 1] explicit maneuver coordination allows to plan driving trajectories within safety critical margins, for which human drivers are not able to do the same because of their limited communication and reaction
abilities. Hence, explicit cooperation seems to necessitate explicit communication. For implicit cooperation a merge situation is discussed, where the drivers’ head poses or even the lateral offsets within a lane are used to negotiate the merging process. In [8], our team pointed out that V2X-communication is just a particular communication channel among other channels that have been used long before like indicators, flashing headlights or a signal horn. No conclusive decision has yet been made, e.g., if those signaling devices should be considered as explicit or implicit communication. Similarly, Sawade and Radusch [9] differentiate between implicit, explicit, and collaborative cooperative driver assistance systems. The latter are assumed to “actively interfere in two or more vehicles at the same time” and to “negotiate”.

Düring and Pascheka [10] clarified that agents are not cooperative per se, but cooperative is rather an attribute to an agent’s behavior. They define “cooperative behavior with respect to [another] agent [...] and with respect to a total utility function [...] if by choosing this behavior [the first] agent [...] knowingly and willingly increases the total utility [...] in a coupled situation, compared to a reference utility.” They assume that knowingly and willingly shall imply more cooperative behavior than just what the agent is forced to do “by legislation or physical laws”.

III. COOPERATION MATRIX

Figure 1 illustrates different levels of skills and abilities and existing communication channels to address these. It lists common examples for these situations in daily driving. It differentiates disparate channels of communication and awareness and different levels of cooperative skills and abilities.

### A. COMMUNICATION AND AWARENESS CHANNELS

As pointed out before, different communication and awareness channels exist. Apart from technical channels like a V2X-communication interface, a standard vehicle already has by traffic laws mandatory signaling devices for communication. Moreover, intentional gestures like longitudinal driving maneuvers, hand gestures or even a driver’s hand gestures might be used. Technically challenging but still a viable mode of communication are unintentional gestures. All communication channels can be used to receive or transmit information. E.g. we can intentionally use indicator flashing to communicate an intention and/or be able to perceive the indicator flashing of other vehicles. If the receiving part dominates the usage of the channel it may be more intuitive to call it an awareness channel. All channels necessitate a detection and tracking of objects to cooperate with.

### B. COOPERATIVE SKILLS AND ABILITIES

Any kind of cooperative multi-agent systems in the scheme of Franklin in [6] require certain cooperative skills and abilities, e.g. regarding perception, reasoning or communication. These skills and abilities enable cooperation on different hierarchical levels and allow different scopes of optimization and altruism as pointed out above.

According to Häcker [11] an ability is defined as the entirety of conditions which are necessary to deliver a performance. In contrast, Heuer [12] defines a skill as performance in a particular task, which results for humans from the background of task-unspecific abilities by training.

Cooperative skills and abilities are to a certain extend hierarchically distinguished.

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**Fig. 1. Examples for cooperative behavior planning tasks classified by communication channel and necessary cooperative skills and abilities**

<table>
<thead>
<tr>
<th>Communication and awareness channels</th>
<th>Cooperative skills and abilities</th>
</tr>
</thead>
</table>
| By unintentional gestures/communication | • Head poses  
• Face expressions  
• Focus of view |
| By intentional gestures/communication | • Head gestures  
• Long. driving gest.  
• Lateral offset  
• ... |
| By mandatory signaling devices | • Horn  
• Indicator  
• Headlight flashing  
• ... |
| By V2X communication | • Vehicle-to-Infrast.  
• Vehicle-to-Vehicle  
• ... |
C. EXAMPLES OF COOPERATIVE BEHAVIOR

On a very basic abstraction level, cooperation can be achieved by the communication and consideration of states and actions for cooperative driving behavior. A possible state to be communicated could be an intervention of the Electronic Stability Control (ESC) system of a vehicle. A possible action to be communicated could be a braking action. The next level does not address single actions or states, but rather tactical maneuvers. A maneuver entails a sequence of actions and states. Maneuvers, which are to be executed are motivated by intentions and goals. Thus, an even higher abstraction level is to communicate and consider intentions of other vehicles even before they are substantiated into executable maneuvers. An example for such an intention could be a bus at the roadside communicating its merge intention and an altruistic automated vehicle which brakes a bit for the bus and communicates that it will let the bus merge back into the traffic. If the skills and abilities are limited to the communication and consideration of intentions and goals, decisions need to be made locally by each deliberative agent (cf. [6]). Vice versa, for negotiating multi-agent systems it is necessary to communicate (inquire and answer) and consider options for behavior alternatives. It is necessary to negotiate by exchanging information about costs of potential behavior options (intentions, goals, maneuvers, actions). This allows to achieve solutions closer to the true overall system optimum with still independent agents. As an example for this, it could be beneficial for a bus to communicate via V2X high costs for the behavior option of not letting it merge in front of an automated vehicle because the bus is behind it’s schedule. It is to be noted, that this communication about options does rarely happen in today’s traffic. On the one hand, because typical traffic participants do not have a communication channel with a suitable bandwidth (shouting between fast moving vehicles, that one is behind its schedule is rarely seen). On the other hand, this level of cooperation requires honest behavior. An individual agent gains personal benefit from communicating wrong behavior option costs. Without a mechanism (e.g., a system of trust) to penalize such selfishness, cooperation on such a high abstraction level might not work at all.

D. THE CHALLENGE FOR COOPERATIVE BEHAVIOR PLANNING IN TODAY'S AUTOMATED VEHICLES

Figure 1 provides a scheme to substantiate the vague term of "cooperation". Yet, figure 1 also illustrates one of today's main challenges for cooperative behavior planning in automated driving. In the very right it lists possible cooperation scenarios using V2X-communication. But, since V2X-communication as of 2015 is not widely available, it is at best an additional channel to improve comfort. Anyhow, cooperative automated driving needs to work safely also without this channel.

On the other end of the axis of communication and awareness channels, is the communication by intended and unintended gestures. Using these channels imposes high perception requirements, which are currently not fully met. As of 2015, it is already challenging to perceive lanes and decently sized objects like cars or trucks reliably. Identifying the driver in a tracked car is at the forefront of research. Estimating his head pose or hand signals seems far out of reach for reliable use for today's cooperative, automated vehicles. The same holds true -although to a smaller extend-for detecting indicators, brake lights or a headlight flashing.

In essence, cooperative automated driving is limited to very few situations, for which communication and awareness channels exist in today's automated vehicles. For many of those, it is rather a matter of taste to call them cooperative behavior or just regular automated driving. E.g., aborting an already initiated lane change maneuver due to suddenly perceiving a vehicle on the neighbor lane could be considered a cooperative maneuver to prevent discomfort or stress for the other vehicle, but it could also just be called a necessary basic feature of a lane change planning module.

IV. SURVEY OF PUBLISHED IMPLEMENTATIONS

Nakamura et al. [13] propose a general concept for vehicle-to-vehicle and vehicle-to-infrastructure communication in automated driving. They see applications in warnings for dangers ahead, emergency notifications, driver information about surrounding vehicles and collision prevention systems.

Longitudinal behavior planning for automated vehicles has been the main focus of the Grand Cooperative Driving Challenge held in 2011 [14]. It stimulated a lot of research around longitudinal behavior planning and vehicle control with V2X-communication, e.g., regarding string stability, fuel consumption minimization and platooning.

However, as todays availability of V2X-communication partners on a regular stretch of highway or in urban areas is close to zero, the focus for the remainder of this survey is directed to cooperative behavior without V2X-communication.

Milanes et al. [15] demonstrate a similar system as above, where they compare a series production ACC-system and a cooperative ACC-system optimized for vehicles cutting in and out closely in front of the test vehicle (cf. SC 6 in table I; "SC" = Scenario) without V2X-communication. Similarly, Freyer et al. [16] try to improve the ACC behavior in scenarios as in SC 6 and SC 9 in table I, by incorporating a situation analysis and a behavior prediction of the lane changing behavior of other traffic participants.

In simulated environments, the development of car-following models for behavior simulation has been a focus of research for several decades. Literature is vast and has extensively been reviewed by, e.g., [17], [18], [19]. Extensions of these models exist to address lane changing and gap selection aspects, in which cooperative behavior is a central topic. Gipps [20] proposes a framework for lane change decision making in sub-urban driving situations (cf. SC 1). Hidas [21, p. 366] and [22] analyzes that merging situations are just a special case of lane changing, because a (ramp) lane ends or is blocked. Particularly but not limited to these situations, he proposes an improved lane change model. If a lane change is necessary but not feasible the merging vehicle
will determine a best possible merge acceleration to make the traffic situation more favorable for lane changes. If no gap seems suitable for a lane change it will pick the most behind gap and thus decelerate (cf. SC 5). Moreover, he implements some kind of cooperative behavior for other vehicles to clear a lane for a merging vehicle (cf. SC 7).

Ruf et al. [23] develop the SPARC framework for behavior planning and addressed scenario SC 10 and SC 13. The prediction model is separated from the reward model. So far it is applied on perfect, simulated data but seem to scale well with real, uncertain sensor data.

Among the core issues for cooperative behavior planning is an appropriate situation assessment. Schubert et al. [24] use a Bayesian network for situation assessment and decision making for lane changes (cf. SC 1). Deceleration to safety time (DST) is used as a central criterion for lane change situation assessment.

Reichel et al. [25] present an approach for situation aspect modeling and situation assessment for merging situations as in scenario SC 9. Frese [26] develops a framework for the planning of cooperative driving maneuvers for automated vehicles. He addresses how to determine cooperative groups and how to modify the trajectory planning to cooperate with other vehicles. His evaluations are based on a simulation environment. He assumes a communication channel to communicate vehicle state variables and maneuvers. Sivaraman et al. [27] demonstrate an assistance system for lane changes to address the scenarios SC 1 and SC 5. Schwarting et al. [28] propose a system to consider costs of other vehicles in a cooperative group and tests the approach in a simulation environment and offline with recorded data form an automated vehicle driving on a highway.

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Ardelt et al. [31], [32] present BMW’s approaches for the scenario SC 1 in its ConnectedDrive-project, focused on highly automated driving on highways.

Team Autonomous from FU Berlin [33] demonstrates the handling of traffic lights/intersections (cf. SC 12, SC 13), obstacle avoidance/passing (cf. SC 10) and according to [33, p. 72] some briefly mentioned lane change implementation (cf. SC 1). Ulbrich et al. [34], [35] demonstrate the handling of lane changes in automated driving in urban areas (cf. SC 1). In their PROUD demonstration, Broggi et al. [36] have addressed cooperative behavior at crosswalks (cf. SC 11), traffic lights/intersections (cf. SC 12) and merging/yielding at intersections and roundabouts (cf. SC 13). Daimler and the KIT [37] demonstrated the handling of yielding and merging at roundabouts (cf. SC 13), traffic lights/intersections (cf. SC 12), pedestrians (cf. SC 11), obstructing objects (cf. SC 10) [38] in their Bertha-Benz automated driving tour.

In the remainder of this section, we present some papers not addressing cooperative behavior as a whole, but which are focusing on certain key aspects that will help to implement cooperative behavior in future. Fröhlich et al. [39] present an approach for turn signal recognition in the frequency domain. Klöden [40], Schmidt et al. [41] and Keller et al. [42] address the perception of head poses and crossing intentions for pedestrians crossing a street. They use unintended gestures for this.

Awal et al. [43] use an V2X-communication based approach to coordinate highway merge situations (cf. SC 6, SC 7), where one vehicle is selected as a group leader for a cooperative group. The group leader receives state information and calculates/communicates a cost-optimal choreography for the cooperative group to cross the intersection. Makarem et al. [44] simulate a V2X-Communication based approach for vehicles to cross intersections without traffic lights by a decentralized model-predictive control approach (cf. SC 13). They assume the communication of states and intentions. Similarly, de Campos et al. [45] present a decentralized approach for unsignalized intersection crossing for automated vehicles with V2X-communication. They assume the exchange of attraction sets for behavior options, thus allowing a negotiation of policies between vehicles.

Figure 2 tries to group some of the references in this paper in the matrix from figure 1. Many publications span over several aspects. Anyhow, for the sake of readability, publications are shown as one point rather than a spanning area. Focus of today’s not-V2X-communication research is the communication and consideration of states, actions and maneuvers by intentional gestures. Many publications use signaling devices to broadcast information. Only few to receive information. Higher levels of cooperative behavior are currently mainly tested in simulations which assume a V2X-like communication channel.

V. IMPLEMENTATIONS IN “JACK”

In January 2015, our team presented “Jack”, the Audi A7 piloted driving concept vehicle for automated driving, at the Consumer Electronics Show 2015 in the US1 and in February on a German highway to the media2. For this, the vehicle drove a stretch of 550 miles on a highway from Stanford to

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1http://www.audi.com/content/com/brand/en/vorsprung_durch_technik/ content/2014/10/piloted-driving.html

<table>
<thead>
<tr>
<th>Nr.</th>
<th>Scenario</th>
<th>Illustration</th>
<th>Addressed by publication</th>
<th>Addressed by &quot;Jack&quot;</th>
</tr>
</thead>
<tbody>
<tr>
<td>SC1</td>
<td>Considering (dis-)comfort costs for rear vehicles</td>
<td><img src="image1.png" alt="Image" /></td>
<td>[20], [24], [31], [32], [34], [35], [27], [29], [30]</td>
<td>✓</td>
</tr>
<tr>
<td>SC2</td>
<td>Giving way to pressing rear vehicles</td>
<td><img src="image2.png" alt="Image" /></td>
<td></td>
<td>✓</td>
</tr>
<tr>
<td>SC3</td>
<td>Requesting cooperation of slow front vehicle (&quot;tailgating&quot;). May not be legal</td>
<td><img src="image3.png" alt="Image" /></td>
<td></td>
<td>x</td>
</tr>
<tr>
<td>SC4</td>
<td>Squeezing into gaps by lateral offsets to the lane center</td>
<td><img src="image4.png" alt="Image" /></td>
<td></td>
<td>✓</td>
</tr>
<tr>
<td>SC5</td>
<td>Squeezing into gaps by longitudinal adjustment to gaps and usage of the indicator</td>
<td><img src="image5.png" alt="Image" /></td>
<td>[21], [22], [27]</td>
<td>✓</td>
</tr>
<tr>
<td>SC6</td>
<td>Letting vehicles merge in front if their lane or on-ramp ends soon, or if they enter the traffic from a bus stop or parking space</td>
<td><img src="image6.png" alt="Image" /></td>
<td>[43], [15], [16]</td>
<td>✓</td>
</tr>
<tr>
<td>SC7</td>
<td>Clearing a lane for vehicles if their lane or on-ramp ends soon, or if they enter the traffic from a bus stop or parking space</td>
<td><img src="image7.png" alt="Image" /></td>
<td>[22], [43]</td>
<td>✓</td>
</tr>
<tr>
<td>SC8</td>
<td>Not changing to a lane on which another vehicle is about to merge to</td>
<td><img src="image8.png" alt="Image" /></td>
<td></td>
<td>✓</td>
</tr>
<tr>
<td>SC9</td>
<td>Dedicated handling of zipper method merging where the automated vehicle merges or let merge</td>
<td><img src="image9.png" alt="Image" /></td>
<td>[25], [16]</td>
<td>✓/x</td>
</tr>
<tr>
<td>SC10</td>
<td>Off-centered driving to leave space for obstructing objects</td>
<td><img src="image10.png" alt="Image" /></td>
<td>[23], [37], [29], [30], [33], [38]</td>
<td>✓</td>
</tr>
<tr>
<td>SC11</td>
<td>Letting a pedestrian pass, who has just entered the road. Possibly -but not limited to crosswalks</td>
<td><img src="image11.png" alt="Image" /></td>
<td>[36], [37]</td>
<td>x</td>
</tr>
<tr>
<td>SC12</td>
<td>Not entering an intersection/conflict area, if it cannot be cleared and cross-traffic will be obstructed</td>
<td><img src="image12.png" alt="Image" /></td>
<td>[33], [36], [37]</td>
<td>x</td>
</tr>
<tr>
<td>SC13</td>
<td>Yielding and merging to another lane. This subsumes 4-way stops, roundabouts, pedestrian and bike lanes, etc. Conflict areas highlighted in red</td>
<td><img src="image13.png" alt="Image" /></td>
<td>[23], [33], [36], [37], [44], [45], [29], [30]</td>
<td>x</td>
</tr>
</tbody>
</table>
Las Vegas and around Braunschweig, Germany. To achieve this, it was necessary to implement several cooperative behavior skills.

Table I shows a non-conclusive list of scenarios and conjoined cooperative behavior. Most of the scenarios need to be considered from both the blue vehicle’s, but also the orange vehicle’s, point of view.

Many of the maneuvers are by some extend linked to lane changes. Lane changes are of particular importance as they often necessitate cooperative behavior or are at least vastly simplified by this. Moreover, one should note, that traffic regulations explicitly dictate cooperation in some of these scenarios.

The automated vehicle is able to perform lane changes while considering the costs of discomfort for other vehicles (cf. SC1). Moreover, it is able to give way to other vehicles tailgating the automated vehicle (cf. SC2). This is particularly relevant for automated driving on German highways, where overtaking on the right is not allowed. The opposite scenario, to tailgate a slow vehicle to request its cooperation to clear the lane (cf. SC3) is technically feasible to implement but will conflict with traffic regulations and has therefore not been implemented.

To enable lane changes in dense traffic, it is necessary not just to wait until a decently sized gap appears, but also to cooperatively “open up” gaps for lane changes. To achieve this, the automated vehicle is able to perform off-centered driving (cf. SC4) and a longitudinal adjustment to the center of a gap and activating the indicator (cf. SC5).

At highway entrance ramps, some degree of cooperation is often expected by other traffic participants. In particular if a truck is merging onto the highway it will expect cooperative behavior of regular vehicles. The scenarios SC6 to SC8 illustrate three aspects of these situations. First, the automated vehicle will open up a gap in front of it in order to let vehicles merge in front (cf. SC6). Secondly, it will avoid changing onto the rightmost lane of a highway if a merging vehicle may also be changing to that lane (cf. SC8). Thirdly, depending on the accuracy of perception and a-priori map data, it will clear a lane for merging traffic by a lane change (cf. SC9).

A zipper-method merging as in scenario SC9 has not dedicatedly been implemented, but is partly covered by the cooperative behavior in the earlier mentioned scenarios already. However, no dedicated counting of other vehicles to determine their order of merging has been demonstrated. If a vehicle insists to move before the automated vehicle it will be tolerated no matter if it is its turn or not.

Scenario SC10 illustrates cooperative, off-centered driving for maintaining comfortable lateral distances towards other objects. This has initially been implemented as a part of the trajectory planning for collision avoidance but also serves the purpose of cooperation on the tactical behavior level.

Cooperative behavior as in scenarios SC11 to SC13 have not yet been demonstrated to the public and are not relevant for the domain of highways that has been demonstrated for the CES 2015 show case.

Implementation details of the cooperative, tactical behavior planning go beyond the scope of this paper. Parts of them are presented in Ulbrich & Maurer [46], [47].

VI. CONCLUSIONS

In this paper, we reviewed definitions and concepts for cooperation in tactical behavior planning for automated vehicles. We pinpointed a possible differentiation by communication and awareness channels and the orthogonal dimension of cooperative behavior skills and abilities. Based on this, we identified the challenge for cooperative tactical behavior planning for today’s automated vehicles imposed by the limitations of available communication channels. Nonetheless, we identified a list of scenarios for cooperative behavior planning being addressable with today’s automated vehicles already. Here we presented what has been implemented in the Audi A7 piloted driving concept vehicle driving automated from Stanford to Las Vegas.

So far, the list of scenarios is not yet complete. The authors expect more scenarios especially in the urban domain. The authors surveyed several publications. However, the matrix in figure 2 as well as the list of scenarios demonstrate several gaps, which haven’t been addressed yet. This publication is limited to just naming these scenarios and fields. It opens a big field of application development to actually implement them facilitating different channels of communication.

REFERENCES
