A DRIVING OPERATIONAL BEHAVIOR ANALYSIS BASED ON THE STATE TRANSITION MODEL FOR AUTONOMOUS VEHICLES

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Introduction

Shared control / Cooperative driving is a practical method for complicated traffic environment like urban areas.



Role of systems:

instruct the safe driving plan trough display, sound, haptic and force feedback HMI devices.



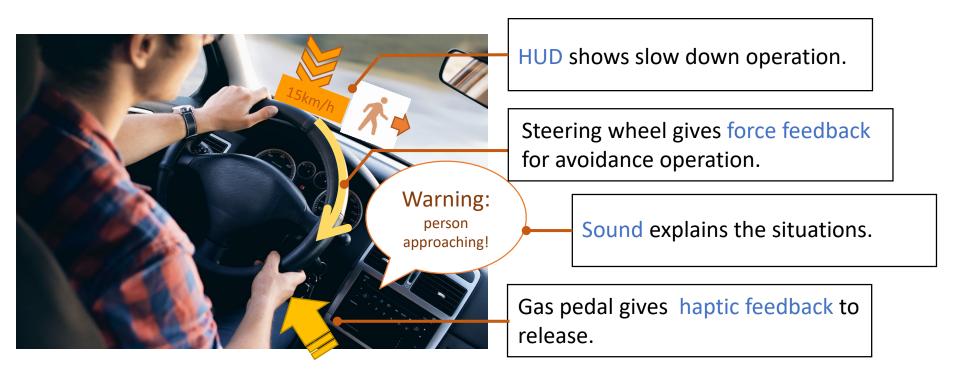
Role of human drivers:

follow (override) the system instructions to keep safety.

Complementing the weak points of both the human and systems realizes safe driving.

Problems

If systems present the multiple operational orders to the driver through multiple HMI devices at the same time.

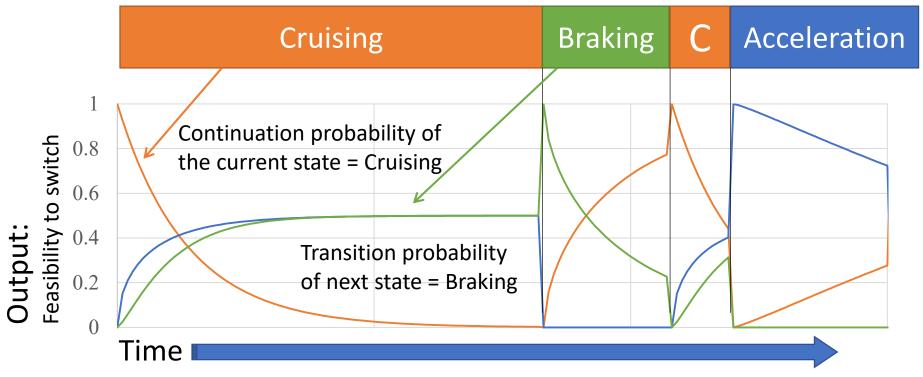


It is difficult for the driver to understand the multiple kind instructions in a short time.

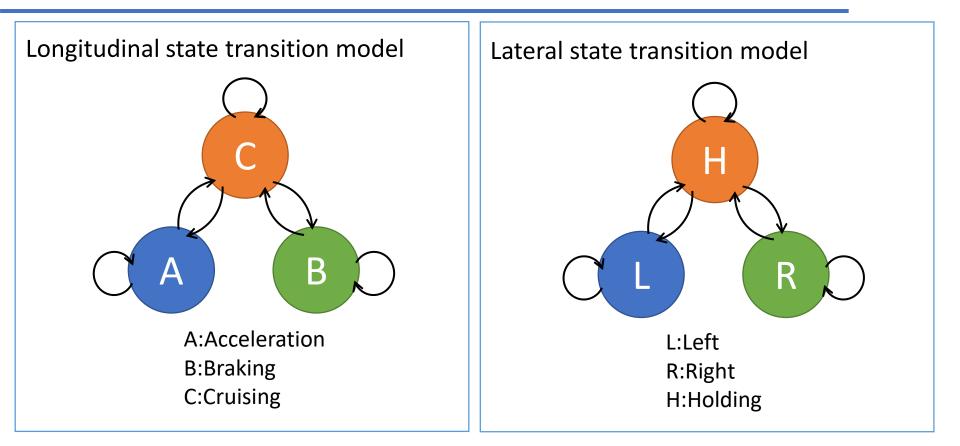
Goal

An evaluation model gives the feasibility of switching driving behaviors by using the state transition model.

Input: state of driving operational behaviors



State transition model of driving behaviors

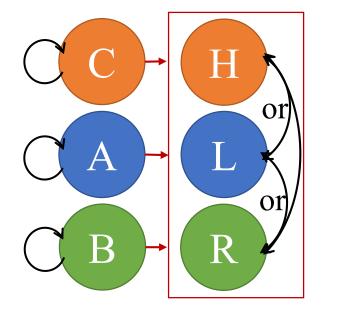


State transition probability is denoted by the conditional probability model.

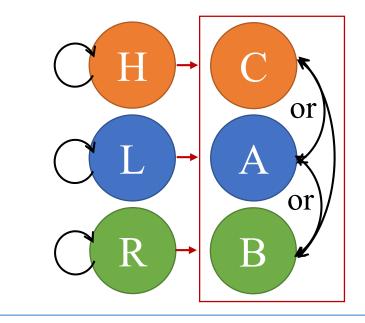
$$A \rightarrow C \qquad \Rightarrow P(C|A)$$

State transition model of driving behaviors

State transition model of the steering wheel operations when a foot pedal is operated.



State transition model of the foot pedal operations when a steering wheel is operated.



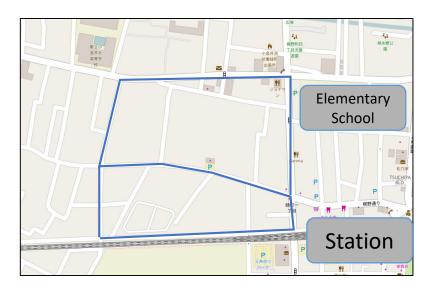
This model represents the co-occurrence probability of the state transition of a driving interface (foot pedal or steering wheel) while the driver is operating another interface.

Data collection

Subjects: 6 instructors of a driving school

- Time : 9 am, 4 pm
- Route : $4 \text{ km} \times 2 \text{ sessions}$
- Area : Residential area near the train station





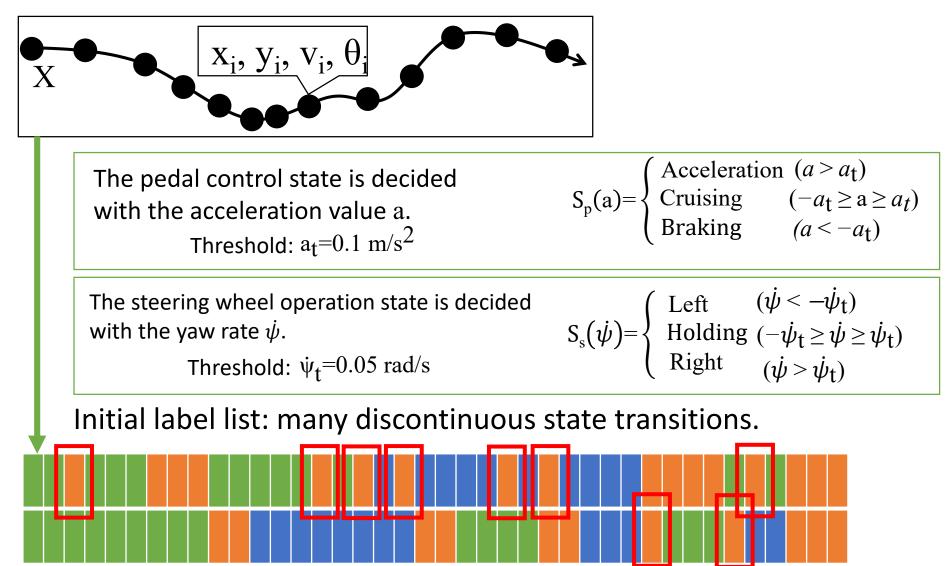
Data types:

- GNSS positioning
- Steering wheel angle
- Gas pedal operation
- Brake pedal operation
- Velocity
- Acceleration
- Yaw rate

https://www.openstreetmap.org/#map=17/35.70374/139.51991

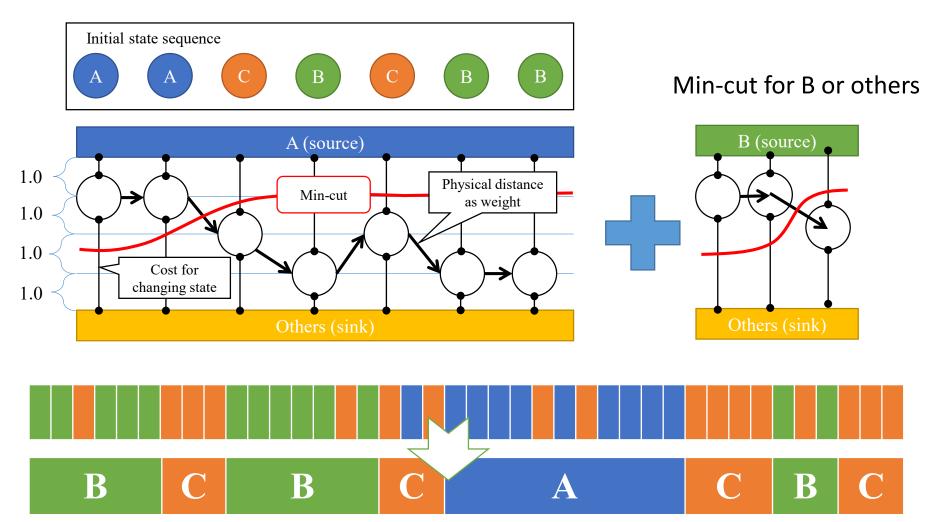
Initial labeling of driving behaviors

The input motion data is divided every 0.1 second to classify.

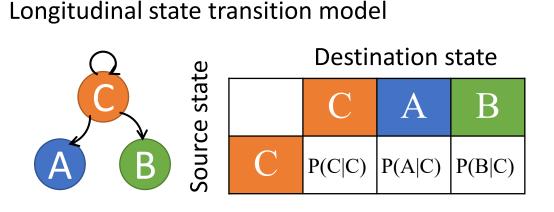


Integration of the driver states based on minimum cut algorithm

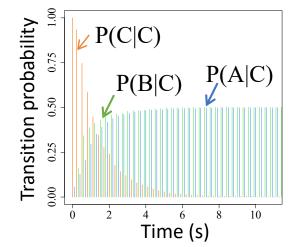
By applying the minimum cut algorithm, the driver states are classified in two states.



Probabilistic models of the driver state transition





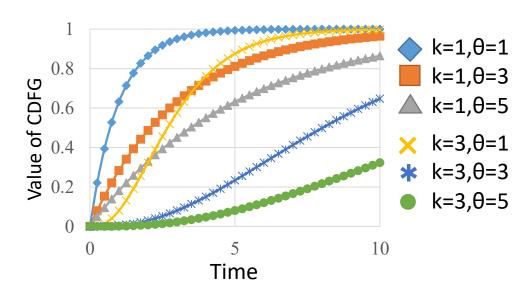


Approximation function:

Cumulative distribution function of Gamma distribution (CDFG)

$$CDFG(t,k,\theta) = \frac{\gamma(k,t/\theta)}{\Gamma(k)}$$

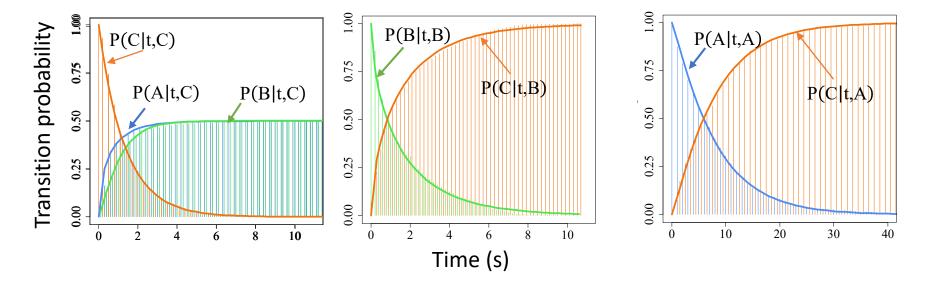
- t: Continues time of the current state
- $\Gamma:$ Gamma function
- γ : Incomplete Gamma function



		С	A	В
Source state	С	P(C t,C)	P(A t,C)	P(B t,C)
	Α	P(C t,A)	P(A t,A)	0
	В	P(C t,B)	0	P(B t,B)

Destination state

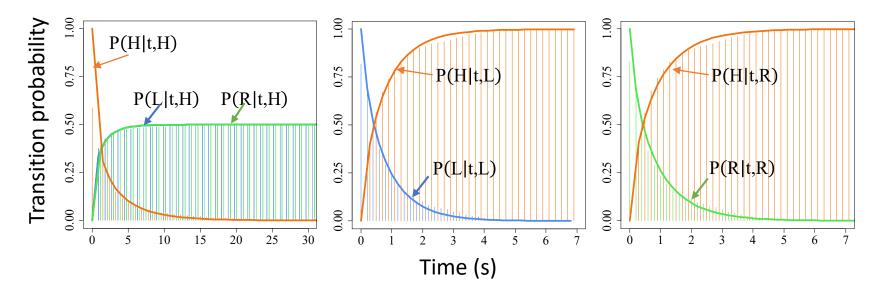
$$\begin{split} P(A|t,C) &= CDFG(t,k_{ca},\theta_{ca})/2\\ P(C|t,C) &= 1 - P(A|t,C) - P(B|t,C)\\ P(B|t,C) &= CDFG(t,k_{cb},\theta_{cb})/2\\ P(C|t,A) &= CDFG(t,k_{ac},\theta_{ac})\\ P(A|t,A) &= 1 - P(C|t,A)\\ P(C|t,B) &= CDFG(t,k_{bc},\theta_{bc})\\ P(B|t,B) &= 1 - P(C|t,B) \end{split}$$

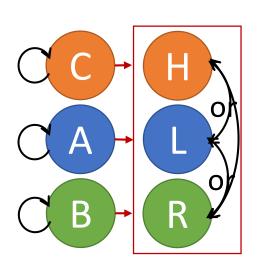


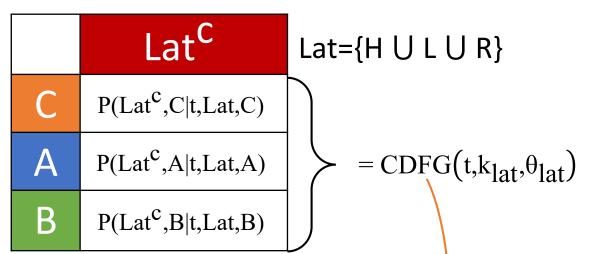
		Destination state				
		Η	Ц	R		
state	Н	P(H t,H)	P(L t,H)	P(R t,H)		
Source s	L	P(H t,L)	P(L t,L)	0		
Sol	R	P(H t,R)	0	P(R t,R)		

Destination state

$$\begin{split} &P(L|t,H) = CDFG(t,k_{hl},\theta_{hl})/2 \\ &P(H|t,H) = 1 - P(L|t,H) - P(R|t,H) \\ &P(R|t,H) = CDFG(t,k_{hr},\theta_{hr})/2 \\ &P(H|t,L) = CDFG(t,k_{lh},\theta_{lh}) \\ &P(L|t,L) = 1 - P(H|t,L) \\ &P(H|t,R) = CDFG(t,k_{rh},\theta_{rh}) \\ &P(R|t,R) = 1 - P(H|t,R) \end{split}$$

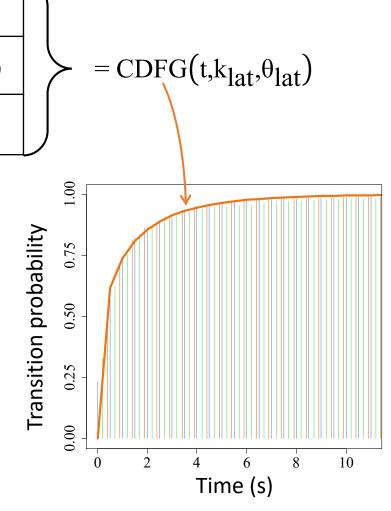


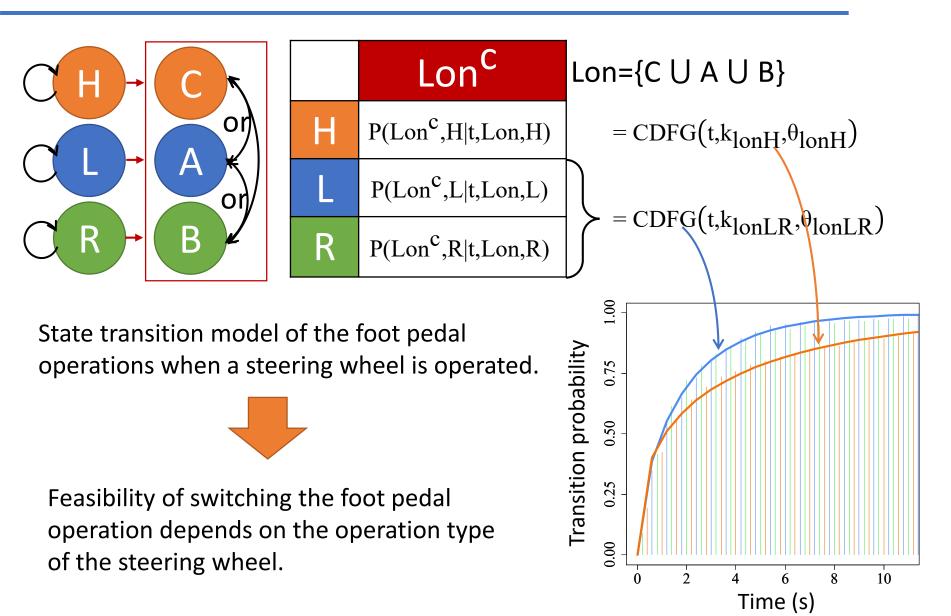




State transition model of the steering wheel operations when a foot pedal is operated.

Feasibility of switching the steering wheel operation is not related to the operation type of the pedals.





Parameters of the state transition probabilities

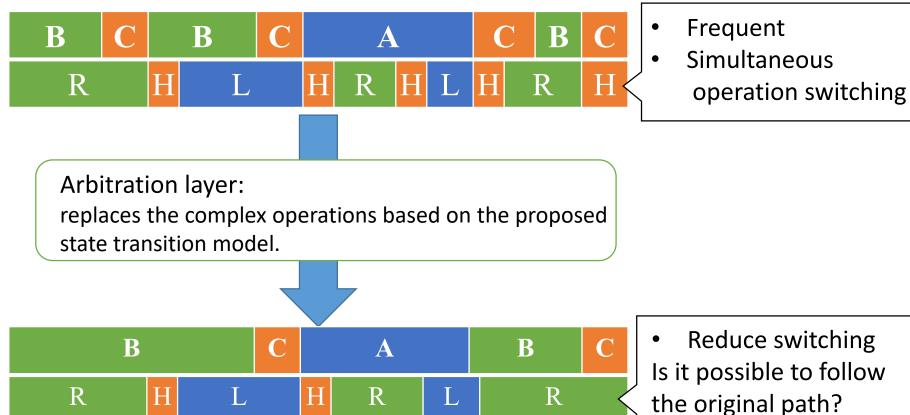
Longitudinal control state			Lateral control state		
transition	К	θ	transition	K	θ
P(A t,C)	0.5	1.31	P(L t,H)	0.43	2.27
P(B t,C)	1.31	0.82	P(R t,H)	0.43	2.27
P(C t,A)	1.18	6.81	P(H t,L)	0.75	0.95
P(C t,B)	0.58	2.8	P(H t,R)	0.68	1.12
Longitudinal co-occurrence state			Lateral co-occurrence state		
$P(Lat^{C},C t,Lat,C)$			$P(Lon^{C},H t,Lon,H)$	0.37	9.2
$P\left(Lat^{C},A \middle t,Lat,A\right)$ $P\left(Lat^{C},B \middle t,Lat,B\right)$	0.31	2.96	$\begin{array}{c c} P\left(Lon^{C},L \mid t,Lon,L\right) \\ P\left(Lon^{C},R \mid t,Lon,R\right) \end{array}$	0.63	2.84



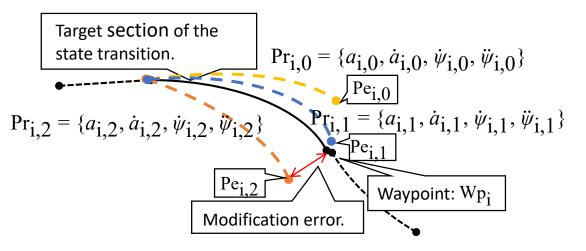
Behavior Planner for Autonomous Vehicles

which translates the motion plan can be easier to control by the driver.

Input: motion plan generated by a driving support system



Application

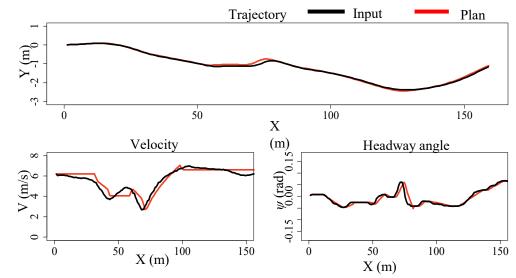


The control parameters are optimized to minimize the sum of the difference from the input path plan.

Results:

Simplification level and path enol				
Feasibility	Mean	Standard		
threshold (%)	error (m)	deviation (m)		
0	0.11	0.20		
5	0.13	0.30		
10	0.17	0.41		
15	0.59	1.90		
20	1.22	3.87		
25	1.97	5.46		

Simplification level and nath error

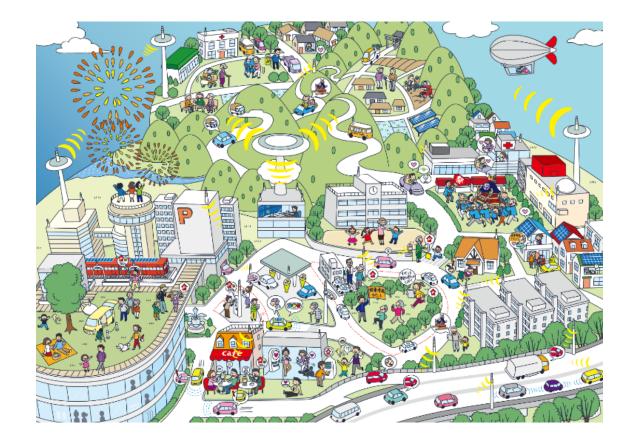


Conclusion

- The state transition models of the driving operations are proposed.
- The feasibility of the operation switching of drivers can be approximated by using Cumulative distribution function of Gamma distribution.
- The driving behavior planner for the arbitration layer between the motion planner and HMI devices is shown as the application.

 As a future work, the comparative experiments of the acceptability of HMI devices equipped with/without the arbitration layer will be performed.

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