# A Prioritized Trajectory Planning Algorithm for Connected and Automated Vehicle Mandatory Lane Changes 

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## Setting

-Macroscopic routing decision given for each CAV
-Mandatory lane changes (MLCs) guided by routing decisions
-Geometry
-An innermost CAVH dedicated lane with CAVs only
-Other HDV lanes with mixed traffic
-A diverging zone (Zone $B$ ) allowing only lane changes of CAVs from dedicated lane into HDV lane
-CAVH intelligence level 3
-Partial Automation
-Driver Assistance System
-Hands off in dedicated lane
-V2V and V2I communication
-Roadside units detect instantaneous information of all vehicles
-System communicates with and controls CAVs in dedicated lane


## Problem Statement

## Customers: CAVs executing MLCs from diverging zone into HDV lane

Service: Trajectory Planning, Preparation for MLCs

## Our Algorithm

- Prioritized: CAVs near the end of the diverging zone given priority in planning
- System-Optimal: Minimizes the total travel time for CAVs in the diverging zone for each decision making
- MLC-Aware: Gives time and space for consecutive MLCs into the off-ramp
- Safety-Guaranteed: Ensures collision avoidance and gap acceptance


## Motivation: MLC

## Importance:

-Guided by routing decisions
-A major reason for congestion at bottlenecks

## Challenges:

-Limited length of the the diverging zone
-Urgency posed by routing decision
-Unpredictability of HDVs on the adjacent lane
-Lacks data set
-Few studies considering MLC for CAVs on a system level

## Assumptions

## -Passenger cars only in the network

-Offers planning for only processes in the diverging zone


## Definitions: Kinematic Parameter

Kinematic Parameter: A parameter chosen by a vehicle that affects its motion
Sufficient Tuple of Kinematic Parameters: A tuple of kinematic parameters given which the position is a function of time


## Definitions: Space Time Slot (STS)

Space Time Slot (STS): An ordered pair of time and position

| Category\Factors to be <br> satisfied | Central <br> Vehicle <br> Kinematic <br> Parameters | Longitudinal <br> Collision <br> Avoidance | Gap <br> Acceptance <br> on HDV Lane |
| :--- | :--- | :--- | :--- |
| Reachable STS | Exists at least 1 <br> sufficient tuple | N/A | N/A |
| Attainable STS | N/A | Satisfied | N/A |
| Joinable STS | N/A | N/A | Satisfied |
| Candidate STS | Exists at least 1 <br> sufficient tuple | Satisfied | Satisfied |

## Definitions: Trajectory

Feasible Trajectory: A trajectory whose STS's are all reachable, attainable under the same sufficient tuple of kinematic parameters, and ends with a candidate STS.


## Algorithm: Framework

Sorting and routing Classification:

- Sorting: Sort the position of CAVs in the diverging zone with descending order
- Classification: Classify the routing decisions of CAVs
- Extract HDVs: Based on HDVs' location extract those that could influence MLC C'AVs' decisions
- Predict HDVs: Predict or Interpolate the future positions of HDVs

Start iteration:

- $\quad$ Start iterating over each MLC CAV in the diverging zone





## Algorithm: Framework

Example:

- 3 sufficient tuples of kinematic parameters available
- Dashed Line: HDV trajectories
- Black Arrow: Leading CAV in the Diverging Zone
- Shaded Area: Joinable STS's
- Red Line: Reachable, but not Attainable
- Yellow Line: Reachable, Attainable but not Joinable
- Green Line: Candidate STS's
- OB: A Feasible Trajectory

Feasible Trajectory: A trajectory whose STS's are all reachable, attainable under the same sufficient tuple of kinematic parameters, and ends with a candidate STS.


## Case Study

Car Following Model: Spring Mass Damper Model with no Leader

- Only 1 element in each sufficient tuple of kinematic parameter


## Cost Function:

1) Total delay in the diverging zone

+ 

2) Expected detour time if a CAV fails to exit at the target off-ramp

HDV Speed Prediction Model: Uniform Speed Prediction

- Advantage: Do not need to perform training
- Disadvantage: Lacks Accuracy


## Simulation

## Geometry:

- An inner CAVH dedicated lane
- An outer HDV lane
- 1500 m of diverging zone

Initialization:


- 5 CAVs with initial speed of $100 \mathrm{~km} / \mathrm{h}$ and position randomized in the first 500 m of diverging zone
- 8 HDVs with initial speed between $60 \mathrm{~km} / \mathrm{h}$ and $100 \mathrm{~km} / \mathrm{h}$, desired speed between $80 \mathrm{~km} / \mathrm{h}$ and $100 \mathrm{~km} / \mathrm{h}$, and random initial position in the HDV lane
- Vehicles in HDV lane use Newell car-following model

Comparison:

- Prioritized System-Optimal Algorithm vs. Gap Acceptance Model


## Results: Position-Time Diagram

Prioritized System-Optimal Algorithm


Gap Acceptance Model (Theoretical)


## Results: Speed-Time Diagram (Diverging Zone)

Prioritized System-Optimal Algorithm


Gap Acceptance Model (Theoretical)


## Results: Metrics

| Metric/Model | Gap Acceptance Model | Prioritized <br> System- Optimal <br> Algorithm |
| :--- | :--- | :--- |
| Average Speed in Diverging Zone | $63 \mathrm{~km} / \mathrm{h}$ | $85 \mathrm{~km} / \mathrm{h}$ |
| Average Distance Driven before a <br> MLC | 257.82 m | 162.1 m |
| Average Time Driven before a MLC | 15.60 s | 7.16 s |
| Calculation Time per CAV | N/A | 0.03 s |

## Results: Discussions

Advantages of Prioritized System-Optimal Algorithm:

- Produced relatively smooth speed change
- Earlier time taken before MLC executions
- Earlier distance driven before MLC executions
- Higher average speed in the diverging zone
- Efficient utilization of spacing in HDV lane
- Relatively low run-time


## Future Research

## Sources of Improvements:

- Examine the application of other car-following models
- Test on other cost functions
- Apply machine learning to forecast trajectories in the HDV lane
- Carry out a larger-scaled simulation to further examine the efficiency
- Consider semi-trucks and buses in the algorithm


## Thanks for Listening!

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