

Introduction

- Traffic accidents cause about 37,000 fatalities and over 230 billion dollars in economic loss every year
- 75% of accidents happen during lane changes
- Advance Driver assistance Systems (ADAS) have the potential to reduce the risk of accidents as the system will take over control of the vehicle in dangerous scenarios and prevent potential crashes
- Detecting dangerous lane change events and modifying them to avoid accidents is essential in increasing traffic safety



Challenges

- \succ Clean and pre-process the naturalistic driving data which contains over 50,000 observations of lanechange events
- \succ Transform the large-scaled raw data to useful, organized information
- \geq Develop a model to capture the real life factors which have the potential to cause the accidents while changing the lane
- > Design the control system to react according to the risk estimation result, in order to decrease the level of danger

Proposed Solution

Scenario: Lane-change maneuver

- **Risk Assessment**: Estimate the risk of ego vehicles and surrounding vehicles that provide a reasonable danger-level estimation
- Collision Avoidance: Implement the model to adjusting the driving behavior to avoid potential collision and reduce risk of current driving behavior

A Collision Avoidance System with Fuzzy Danger Level Detection

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Methodology

Fuzzy Danger Level Detection	E
Stimate on Subject vehicle	
Inputs: Velocity, Yaw-rate,	
Accelerations in Longitudinal & Lateral	
 Model based on Mamdani's fuzzy inference 	
method	> Fig
 Output: Danger-level percentages 	sul adj
Stimate on Surrounding vehicle	and
• <i>Inputs</i> : Relative velocity of surrounding	> Fig
vehicle, relative distance in longitudinal &	tra
lateral	> Th
 Model Based on Mamdani's fuzzy inference 	hig
method	be
 Output: Danger-level percentages 	acc
	2 ~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~
Collision Avoidance System	Acceleratio
Model predictive control(MPC)	
Trajectory planning	400 E 300 -
• Optimize the accelerations in longitudinal	
• Optimize the distance in lateral	0 2 34
• Danger level affect on safe sap constrain	[5] 32 -] Ĕ
Fault Determination Classifier	Γ
• Extreme Gradient Boosting trained with	0.5
naturalistic driving data	tiou [m/s ²]
Driving Driving	-1- 4CCCE ACCE
Euzzy Danger Level Model	400
	 ප 5j200 -
Danger Level Maneuver Modification	
	0 2 27
Avoidance Modify the safe gap SG(new)=SG(default)*(1+DL) Modify the velocity constrains and other for safer trajectory	26.5 [s/L] 26 Ation 26
	25.5 25 24.5
	υ 2

Figure 2: Schematic diagram of the proposed system

Simulation and Results

Example Lane-change Driving Scenario:



Figure3: Example of a near crash event

gure 3 is one near-crash event from dataset as bject vehicle trying to execute lane-change to jacent lane, but become too close to other vehicle d have to turn back to prevent the accident

gures 4-7 present the longitudinal and lateral jectory planning of an event before and after the stem is activated

e proposed system can detect that maneuver with ghest 75% danger level and the fault is most likely to subject vehicle, the trajectory will be modified cordingly



4, 5: Longitudinal trajectory, Lateral trajectory of a near crash event



Figure 6, 7: Longitudinal trajectory, Lateral trajectory of the modified crash event



Significance

✓ Estimate the driving behavior • Ability to evaluate the driving behavior of subject vehicle and provide a reasonable estimation of the danger of the maneuver

• Ability to plan the trajectory according to the percentage of the danger-level to ensure the safety of the subject vehicle

Future Work

 \blacktriangleright Developing a risk assessment that considers the interactions between road users and intention of surrounding vehicles to help the lane-change execution more smooth and efficient

Target Market

• Vehicle manufacturers developing advance driving assistance system

• Insurance companies who wish to monitor the abnormal behavior of customers' vehicle

industry

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Conclusion/ Future Work

✓ <u>Prevent/decrease the potential danger</u>

Relevance

- *ADAS is a popular technic in the Automotive
- Situation Assessment & Collision Avoidance is a essential part for ADAS
- Naturalistic driving data analyze method could applied in similar dataset such as: UAV datasets

References

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