

Scene-aware and Social-aware Motion Prediction for Autonomous Driving

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1 Introduction

2 Method Description

- The Dataset Collection
- Stage 1 - Filtering process
- Stage 2 - Integration Model

3 Results

- Scenario Filtering
- Integration Method

Introduction

Autonomous Driving Promise

- Efficiency and Safety

Challenges in Motion Prediction

- Multimodality
- Scene Dependence
- Social Acceptability

Crucial Understanding

- Human-Driven Behavior Key

Limitations of Current AI Tools

- Control Perspective Absent
- Intent Interpretation Challenge



Overview of Our Approach

Testing and Evaluating State-of-the-Art Tools

- Understanding the real-world applicability and limitations of these tools

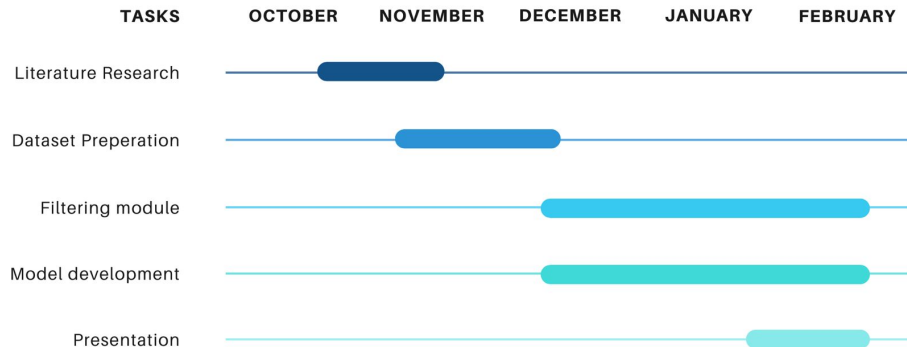
Developing Control-Oriented Tools

- Introduce virtual forces between vehicles to improve the accuracy of movement predictions

Specific Focus on Vehicle Interactions

- Formulate more accurate and socially-aware predictive models based on these analyses.

Timeline

**Alfred**

Literature Research
Dataset Preparation
Model development

Baris

Literature Research
Dataset Preparation
Filtering module

Agenda

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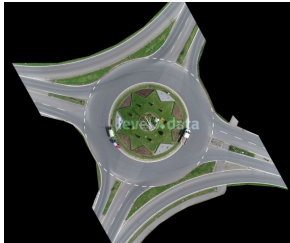
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- Integration Method

Dataset Collection

exiD



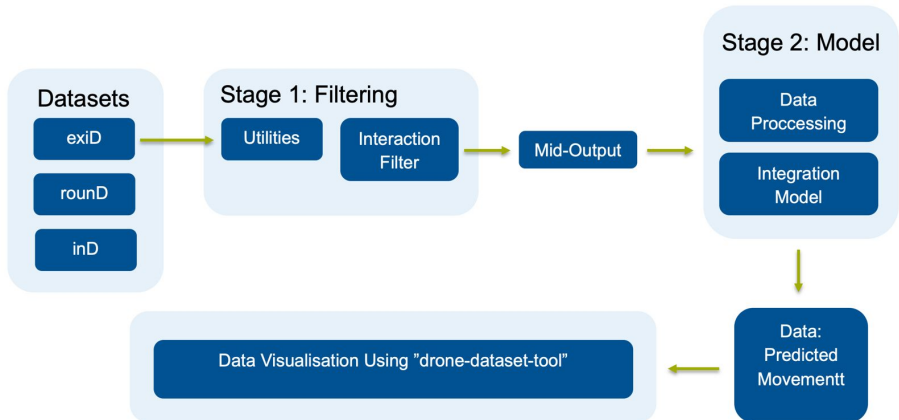
rounD



inD



Method Description - Overview



Agenda

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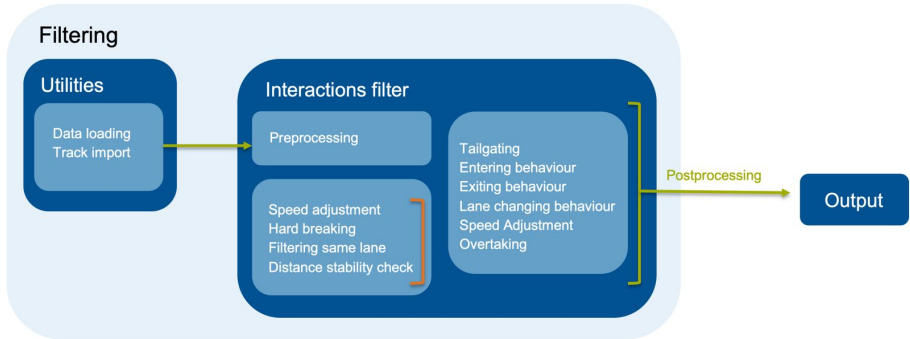
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- **Stage 1 - Filtering process**
- Stage 2 - Integration Model

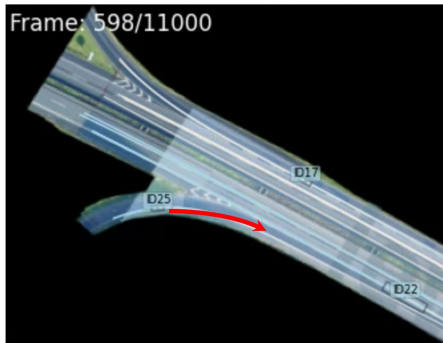
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Method Description - Stage 1



Method Description - Stage 1



Merging Lane Entering Scenario

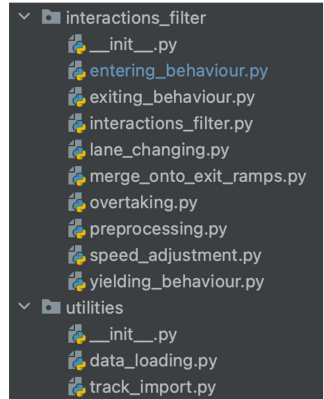


Merging Lane Exiting Scenario

Method Description - Stage 1

Filtering Stage: Identifying Vehicle Behaviors

- Preprocessing
- Behavior Detection
 - Entering/Exiting Behavior
- Interaction Analysis
- Lane Change Detection
- Thresholds and Conditions
- Data Grouping and Sorting



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Previous Integration Model

Distance and Velocity Equations (Ballistic Integration):

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$$s(k+1) = s(k) + dt \cdot v(k) + \frac{dt^2}{2} a(k)$$

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Our Integration Model

Distance and Velocity Equations:

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Our Integration Model

Distance and Velocity Equations:

$$s(k+1) = s(k) + dt \cdot v(k) + \textcolor{red}{c}_1 a(k) + \textcolor{red}{c}_2 a(k-1)$$

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Acceleration Equations:

$$a(k) = -\overline{\textcolor{red}{c}_1} a(k-1) + \overline{\textcolor{red}{c}_2} (s(k+1) - s(k) - dt \cdot v(k))$$

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⇒ This can be solved using linear regression.

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Scenario filtering

Video demo of the scenarios

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Reminder: Our Model - Matrix Form

Acceleration from Distance formula:

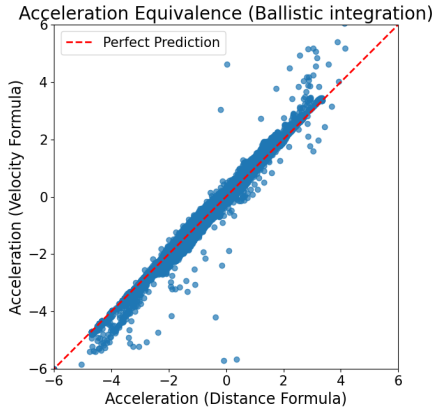
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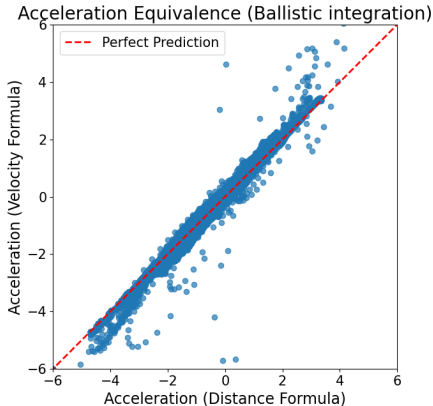
⇒ This can be solved using linear regression.

Results: Comparison to the old acceleration model

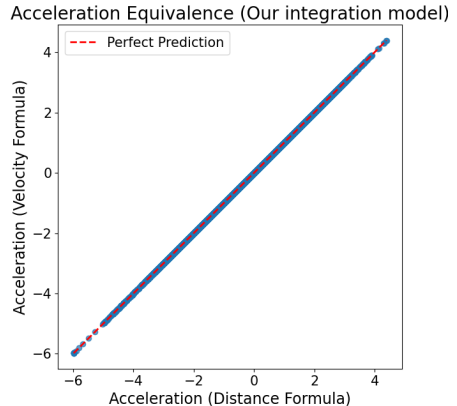


MSE: 3.0786×10^2

Results: Comparison to the old acceleration model



MSE: 3.0786×10^2



MSE: 1.9220×10^{-9}

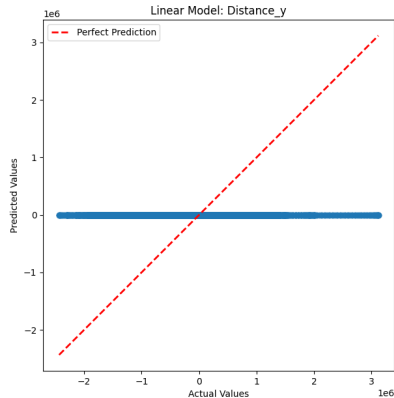
Results: Integration Method

Rearranging the formula to the distance and velocity gives us these results:

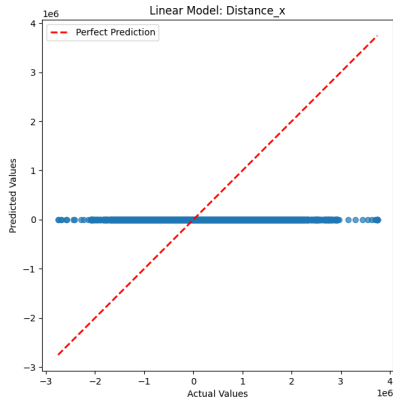
Video demo of predicted car

Q&A

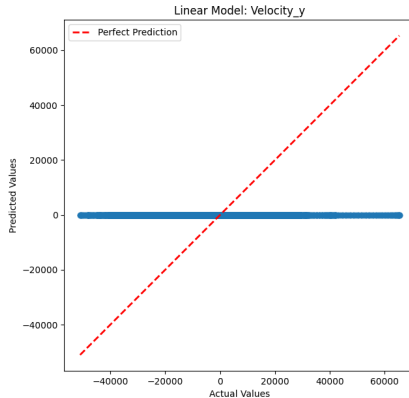
Acceleration Modification in the Y-axis



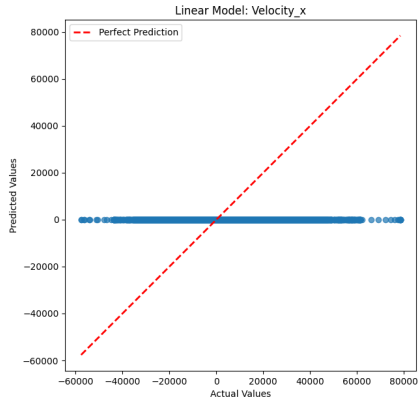
Acceleration Modification in the Y-axis



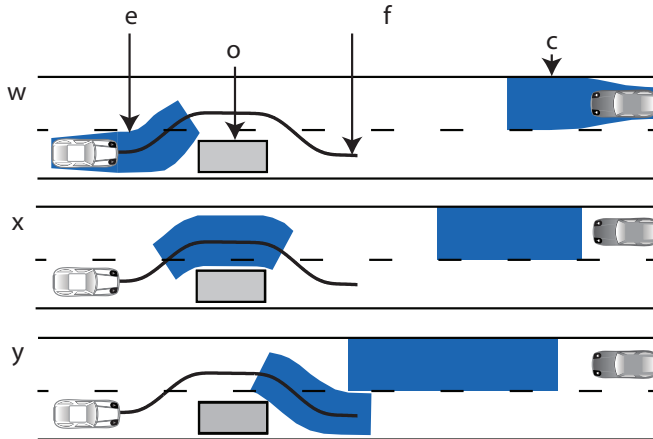
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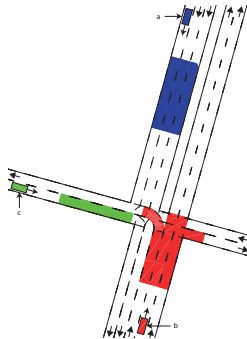
Motivation for Set-Based Prediction [1]



[1] M. Althoff and S. Magdici, "Set-based prediction of traffic participants on arbitrary road networks," IEEE Transactions on Intelligent Vehicles, vol. 1, no. 2, pp. 187–202, 2016.

SPOT

SPOT: A tool for set-based prediction of traffic participants [2]



Initial configuration and $\mathcal{O}(t)$ for $t \in [1.5\text{ s}, 2.0\text{ s}]$

[2] M. Koschi and M. Althoff, "SPOT: A tool for set-based prediction of traffic participants," in Proc. of the IEEE Intelligent Vehicles Symposium, pp. 1679–1686, 2017.

Conclusions

- Item

- Item

- Item

beginframe

Distance and Velocity Equations:

$$s(k+1) = s(k) + dt \cdot v(k) + \frac{dt^2}{2} a(k)$$

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onslidej2j

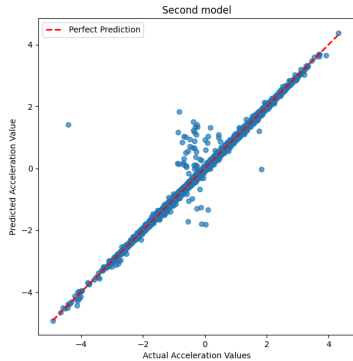
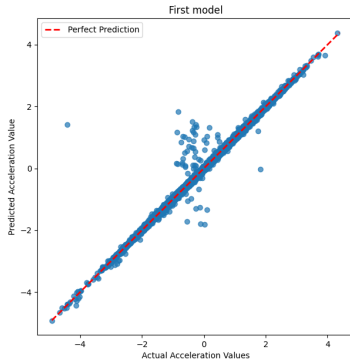
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$$a(k) = \frac{2}{dt^2} \left(s(k+1) - s(k) - dt \cdot v(k) \right)$$

endframe

$$a(k) = \frac{1}{dt} \left(v(k+1) - v(k) \right)$$

Results: Integration Method



Results: Integration Method

Video demo of predicted car

Our Integration Model

Our Distance and Velocity Equations:

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Model in matrix form:

$$\begin{bmatrix} a(k) \\ a(k) \end{bmatrix} = \begin{bmatrix} -a(k-1) & s(k+1) - s(k) - dt \cdot v(k) & 0 & 0 \\ 0 & 0 & -a(k-1) & v(k+1) - v(k) \end{bmatrix} \begin{bmatrix} \overline{c_1} \\ \overline{c_2} \\ \overline{c_3} \\ \overline{c_4} \end{bmatrix}$$

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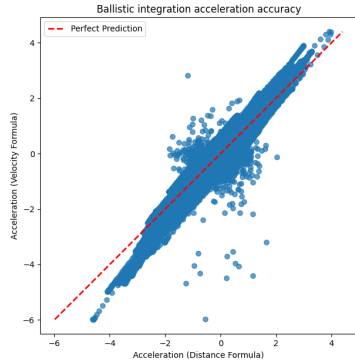
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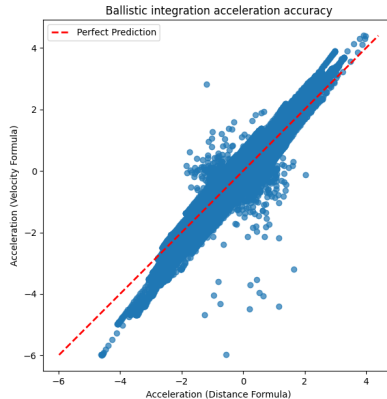
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Accuracy of the prediction for the acceleration using the Ballistic Integration method (MSE): 4.3249×10^{-2}



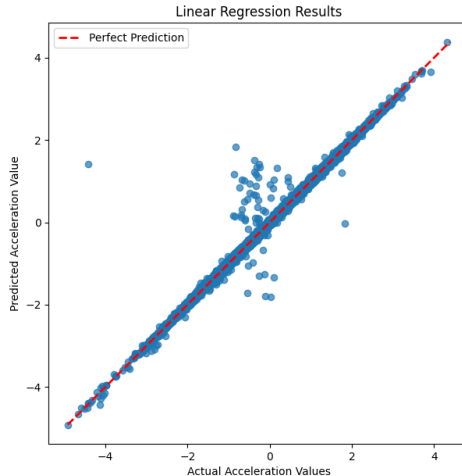
Previous Integration Model - Accuracy

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Results: Integration Method

Accuracy of the prediction for the acceleration (MSE): $3.0955e-03$



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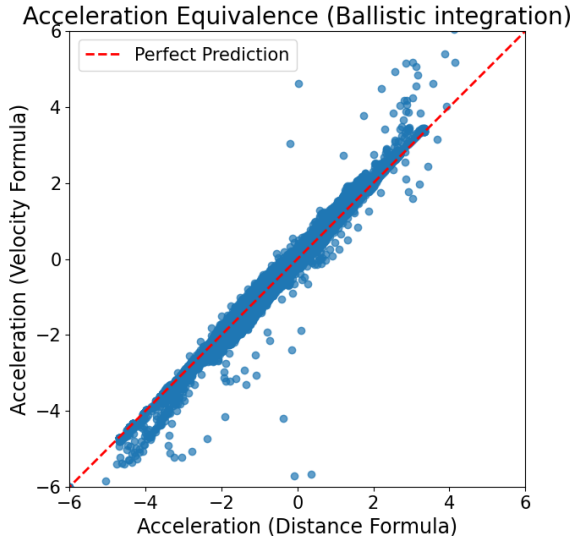
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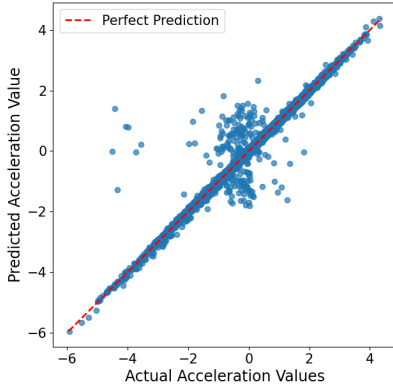
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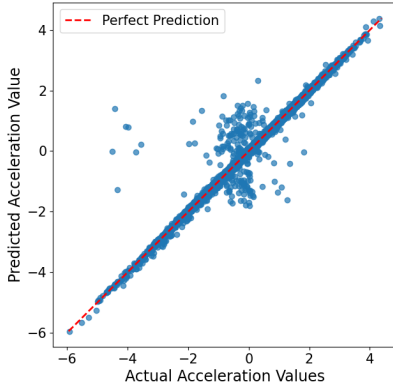
Results: Integration Method

Prediction of the Acceleration: Distance formula

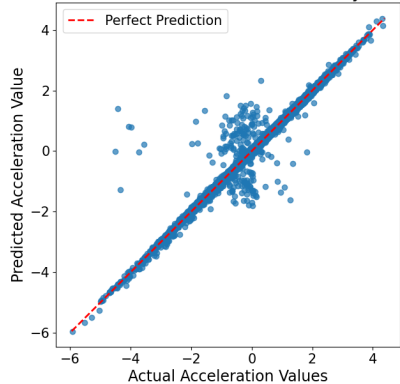


Results: Integration Method

Prediction of the Acceleration: Distance formula



Prediction of the Acceleration: Velocity formula



Results

Summary:

- Successfully implemented the filtering mechanism

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Summary:

- Successfully implemented the filtering mechanism
- Able to filter out X different scenarios in Y datasets
- Found a better integration method where the accelerations match
- Able to visualize the integration method and modulate the movement of a car