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

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Agenda



- Introduction
- 2 Method Description
 - The Dataset Collection
 - Stage 1 Filtering process
 - Stage 2 Integration Model
- 3 Results
 - Filtering Process
 - Integration Method
- 4 Future Work



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





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 Preperation Model development

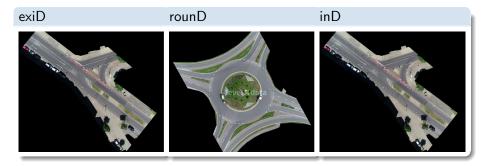
Baris



Literature Research Dataset Preperation Filtering module

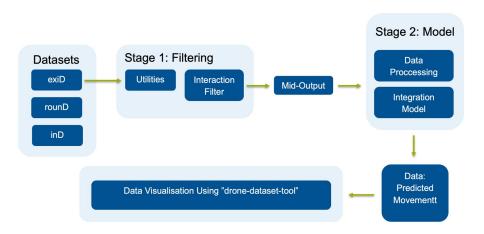
Dataset Collection





Method Description - Overview

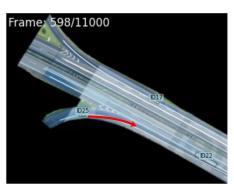












Merging Lane Entering Scenario



Merging Lane Exiting Scenario

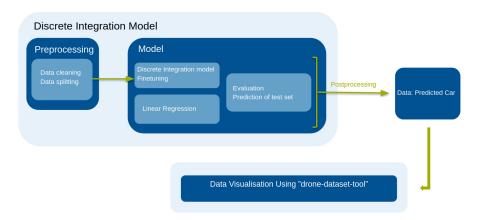


Filtering Stage: Identifying Vehicle Behaviors

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

```
interactions_filter
     💤 __init__.py
     antering behaviour.pv
     a exiting behaviour.pv
     interactions filter.pv
     lane_changing.py
     the merge_onto_exit_ramps.py
     a overtaking.pv
     a preprocessing.pv
     speed_adjustment.py
     Light vielding behaviour.pv
utilities
     💤 __init__.py
     data_loading.py
     track_import.py
```







Distance and Velocity Equations:



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$$s(k+1) = s(k) + dt \cdot v(k) + c_1 a(k) + c_2 a(k-1)$$

$$v(k+1) = v(k) + c_3 a(k) + c_4 a(k-1)$$

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Note:

- The acceleration resulting from both formulas should be equal
- Model can be solved using linear regression.

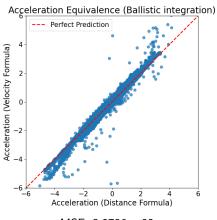
Scenario filtering



Video demo of the scenarios





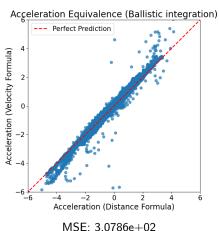


MSE: 3.0786e+02

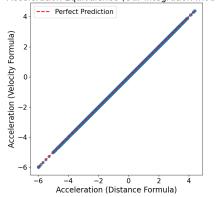
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Results: Comparison to the old acceleration model





Acceleration Equivalence (Our integration model)



MSE: 1.9220e-09

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Results: Integration Method



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

Video demo of predicted car



Scenario Filtering:

• Specify even more scenarios for a broader range of use cases.



Scenario Filtering:

- Specify even more scenarios for a broader range of use cases.
- Explore other datasets



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Integration Model:

• Finetune the integration model (adding other parameteres)



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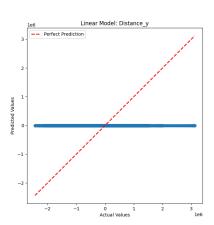
Integration Model:

- Finetune the integration model (adding other parameteres)
- Test the integration model with the neural network for performance (task for the next team)

Q&A

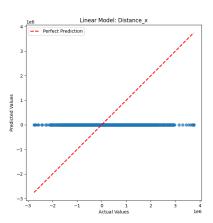
Acceleration Modification in the Y-axis





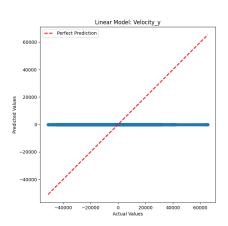






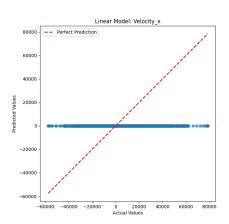






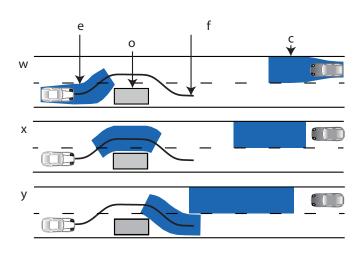






Motivation for Set-Based Prediction [1]



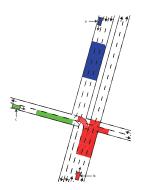


^[1] M. Althoff and S. Magdici, "Set-based prediction of traffic participants on arbitrary road networks," IEEE Transactions on Intelligent Vehicles, vol. 1, no. 2, pp. 187–202, 2016.



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



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

^[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:

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

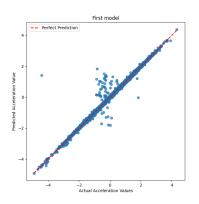
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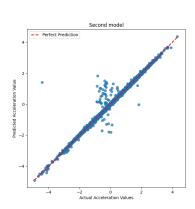
Acceleration Equations:

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

Results: Integration Method







Results: Integration Method



Video demo of predicted car

Our Integration Model



Our Distance and Velocity Equations:

Our Integration Model



Our Distance and Velocity Equations:

$$s(t+1) = s(t) + dt \cdot v(t) + c_1 a(t) + c_2 a(t-1)$$

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Our Acceleration Equations:

$$\begin{aligned} a(k) &= -\overline{c}_1 a(k-1) + \overline{c}_2 \big(s(k+1) - s(k) - dt \cdot v(k) \big) \\ a(k) &= -\overline{c}_3 a(k-1) + \overline{c}_4 \big(v(k+1) - v(k) \big) \end{aligned}$$



Model in matrix form:

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

Our Integration Model



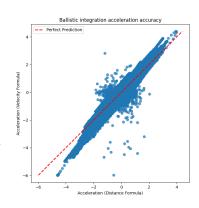
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⇒ This can be solved using linear regression.



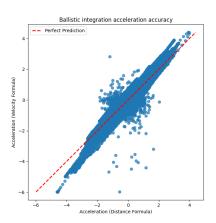
Accuracy of the prediction for the acceleration using the Ballistic Integration method (MSE): 4.3249e-02





Previous Integration Model - Accuracy

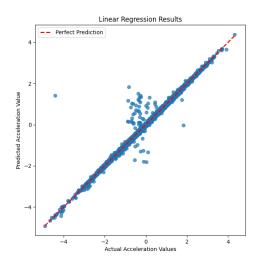
Accuracy of the prediction for the acceleration using the Ballistic Integration method (MSE): 4.3249e-02







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



Previous Integration Model



Distance and Velocity Equations (Ballistic Integration):

Previous Integration Model



Distance and Velocity Equations (Ballistic Integration):

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

$$a(k) = \frac{2}{dt^2} \Big(s(k+1) - s(k) - dt \cdot v(k) \Big)$$
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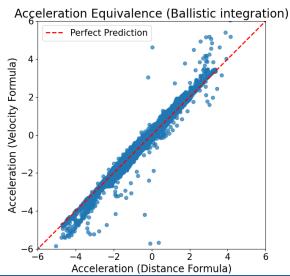
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Problem: Accelerations are not equal!

Previous Integration Model - Accuracy









Acceleration from Distance formula:

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Our Integration Model - Matrix Form



Acceleration from Distance formula:

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Acceleration from Velocity formula:

$$[a(k)] = [-a(k-1) \quad s(k+1) - s(k) - dt \quad v(k)] \begin{bmatrix} \overline{c_3} \\ \overline{c_4} \end{bmatrix}$$

 \Rightarrow This can be solved using linear regression.

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Efficiency and Safety

Challenges in Motion Prediction

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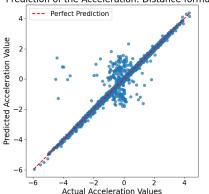
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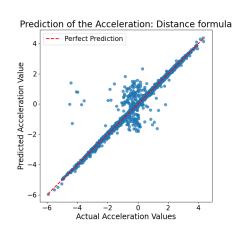


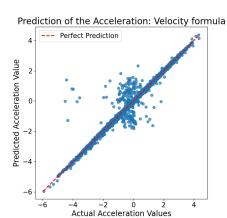












Results



Summary:

• Successfully implemented the filtering mechanism

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- Able to filter out X different scenarios in Y datasets



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- Found a better integration method where the accelerations match



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

Previous Integration Model



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