

# Vehicle Interaction Learning

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# Additional Prior Information

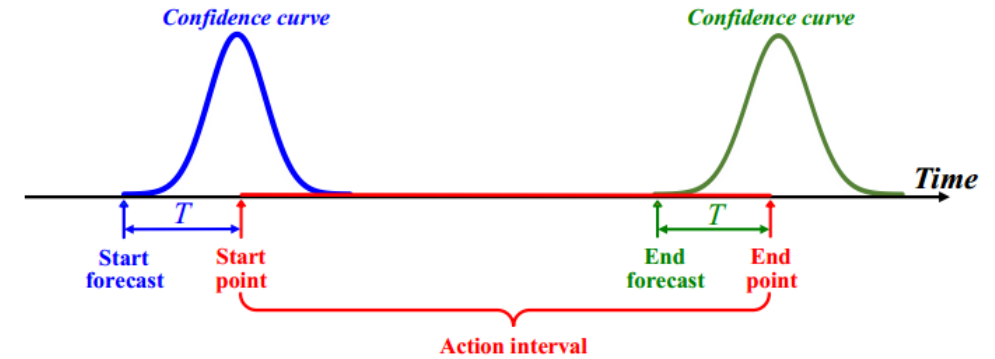
- Starting < Ending
- Naïve Algorithm: Confidence > threshold and Farthest Local Minima -> Changing\_Lane\_Time  
Acc > 90%
- Other Trying:

1.  $\sigma = 5$

```
[0.056 0.089 0.135 0.198 0.278 0.375 0.487 0.607 0.726 0.835 0.923 0.98  
1. 0.98 0.923 0.835 0.726 0.607 0.487 0.375 0.278 0.198 0.135 0.089  
0.056]
```

->  $\sigma = 1.2$  worse

```
[1.929e-22 5.670e-19 8.324e-16 6.102e-13 2.234e-10 4.083e-08 3.727e-06  
1.699e-04 3.866e-03 4.394e-02 2.494e-01 7.066e-01 1.000e+00 7.066e-01  
2.494e-01 4.394e-02 3.866e-03 1.699e-04 3.727e-06 4.083e-08 2.234e-10  
6.102e-13 8.324e-16 5.670e-19 1.929e-22]
```



# Additional Prior Information

- Other trying 2:
- Inverse Order Loss:

Start\_Confidence = [0.3, 0.6, 0.9, 0.3]

End\_Confidence = [0.7, 0.3, 1.0, 0.2]

Start\_Max\_After = [0.9, 0.9, 0.9, 0.3]

End\_Max\_Before = [0.7, 0.7, 1.0, 1.0]

Inver Order Loss =  $\max(0, \text{Start\_Confidence} + \text{End\_Max\_Before} - 1) + \max(0, \text{End\_Confidence} + \text{Start\_Max\_After} - 1)$

Results:

1. 6X slower
2. Inverse Order Loss  
~0 after average  
among batch and  
length

# Additional Prior Information

- Other possible way?

1. Tune  $\sigma$

2. Proposal Confidence Scorer -> Train by IoU

3. Loss Function

4. Model Structure

# About IoU

- $\text{IoU} = \frac{A \cap B}{A \cup B}$

- As for 15 frames (fps=5):

IoU = 0.6 -> 25 frames max-len or 3 frames offset ( $12/18 = 0.667$ )

IoU = 0.7 -> 21 frames max-len or 2 frames offset ( $13/17 = 0.765$ )

IoU = 0.8 -> 18 frames max-len or 1 frames offset ( $14/16 = 0.875$ )

IoU = 0.9 -> 16 frames max-len or 0 frames offset

# Best Results

- Train on NGSIM(50s):

Validation Set: Best\_IoU6\_Acc 0.848 Best\_IoU7\_Acc 0.736  
Best\_IoU8\_Acc 0.535 Best\_IoU9\_Acc 0.201, Best\_Change\_Lane\_Acc  
0.972, Best\_Traj\_Cls\_Acc 0.895

Transfer on HighD(50s):

Exchange xy: IOU6 Acc 0.741, IOU7 Acc 0.576, IOU8 Acc 0.316,  
IOU9 Acc 0.087, Change\_Lane\_Acc 0.919, Traj\_Cls\_Acc 0.762

no\_exchange xy: IOU6 Acc 0.630, IOU7 Acc 0.457, IOU8 Acc 0.244,  
IOU9 Acc 0.065, Change\_Lane\_Acc 0.870, Traj\_Cls\_Acc 0.750

# Best Results

- Train on HighD(50s):

Validation Set: Best\_IoU6\_Acc 0.966 Best\_IoU7\_Acc 0.899

Best\_IoU8\_Acc 0.711 Best\_IoU9\_Acc 0.355, Best\_Change\_Lane\_Acc 0.993, Best\_Traj\_Cls\_Acc 0.994

Transfer on NGSIM (10s):

Exchange xy: IOU6 Acc 0.635, IOU7 Acc 0.489, IOU8 Acc 0.304, IOU9 Acc 0.085, Change\_Lane\_Acc 0.873, Traj\_Cls\_Acc 0.842

no\_exchange xy: IOU6 Acc 0.647, IOU7 Acc 0.488, IOU8 Acc 0.268, IOU9 Acc 0.075, Change\_Lane\_Acc 0.865, Traj\_Cls\_Acc 0.792

# Some Visualization

- Train on NGSIM:

Training Set: Samples:5339, False Positive:414, False Negative:14

Validation Set: Samples:1310, False Positive:129, False Negative:27

Transfer on HighD: Samples:12514, False Positive:2958 False  
Negative:26



# About Trajectory Max Length

Train Set	Length	Validation Res			Transfer (exchange xy)			Transfer (not exchange xy)		
		IoU 6 Acc	Change Lane Acc	Traj Cls Acc	IoU 6 Acc	Change Lane Acc	Traj Cls Acc	IoU 6 Acc	Change Lane Acc	Traj Cls Acc
NGSIM	10s	91.7%	98.5%	90.8%	57.0%	85.7%	86.3%	57.0%	78.7%	89.0%
NGSIM	20s									
NGSIM	50s	84.8%	97.2%	89.5%	74.1%	91.9%	76.2%	63.0%	87.0%	75.0%
HighD	10s	96.3%	99.3%	99.3%	60.8%	90.3%	80.8%	64.6%	93.1%	78.7%
HighD	20s									
HighD	50s	96.6%	99.3%	99.4%	36.4%	75.3%	70.0%	37.1%	69.7%	75.2%

# About Scaling

- Scale locally by Trajectory (max) then scale globally by all frames(mean std) -> validation: worse, transfer: IoU (slightly better), traj\_cls (worse)
- Scale locally by Trajectory (max) -> worse
- Scale globally by maximum of all values then rotation -> not converge (deeper model does not help as well)
- Rotation then scale(mean-std) by batch -> not converge (deeper model does not help as well)
- Use same scale factor for car1 and car2 (still training)



# Data Augmentation

- xy augmentation:  $(x1, y1, x2, y2) \rightarrow (y1, x1, y2, x2)$
- vehicle augmentation:  $(x1, y1, x2, y2) \rightarrow (x2, y2, x1, y1)$
- Random rotation

4x samples + randomness  $\rightarrow$  Nearly No Overfitting (Maybe we could use deeper model)

Best Performance on Validation Set

- Slow (50s data)  $\rightarrow$  more than 25 hour to converge