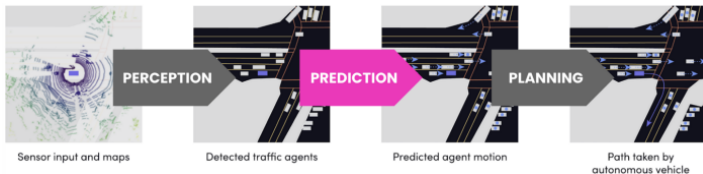


Interpretable Long Term Waypoint-Based Trajectory Prediction Model

Amina Ghoul, Itheri Yahiaoui and Fawzi Nashashibi

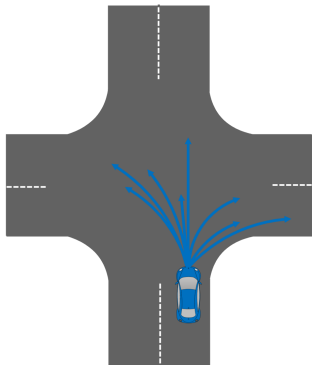
What is motion prediction ?

Predicting agent's future behavior in dynamic, multi agent system.



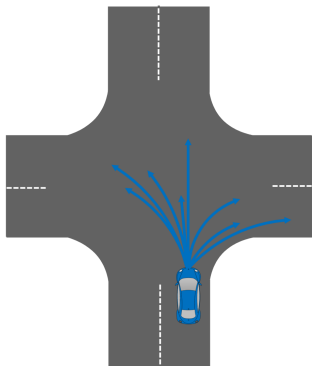
Autonomous vehicle pipeline

Why is motion prediction challenging ?

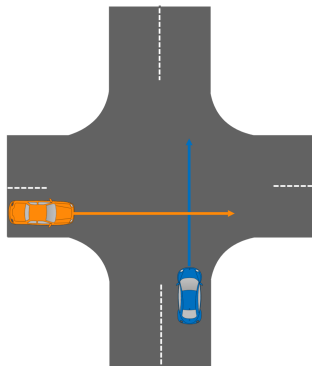


Future is uncertain.

Why is motion prediction challenging ?



Future is uncertain.



Interaction between agents.

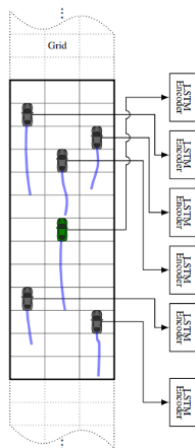
Related Work

Input Representation

Input cues :

- The agents' past states
(coordinates, velocity, heading...)

Encoder

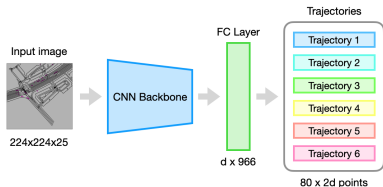
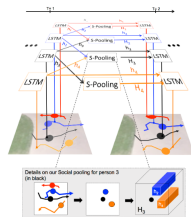


Related Work

Input Representation

Input cues :

- The target agent's past states.
- Interaction with its environment.
 - Other agents.
 - The static scene.

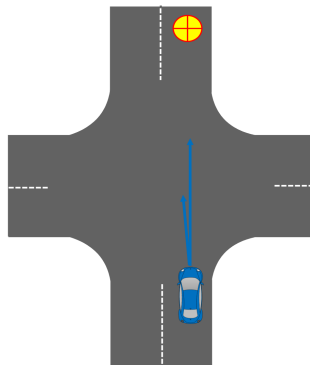



[2] MultiPath++ [Varadarajan et.al], [3] MotionCNN [Konev et.al]

Final destination

Input cues :

- The target agent's past states.
- Interaction with its environment.
- Its Long Term Waypoint (LTW) :
An agent future trajectory is most influenced by its final destination.

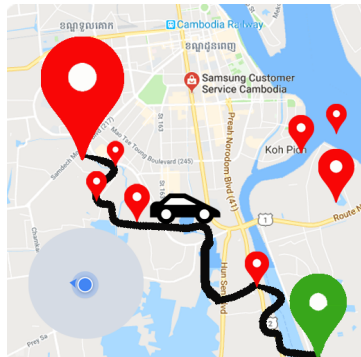


 Long Term Waypoint (LTW)

Final destination

Input cues :

- The target agent's past states.
- Interaction with its environment.
- Its Long Term Waypoint (LTW) :
An agent future trajectory is most influenced by its final destination.
- In practice : GPS provides waypoints to reach the final destination.



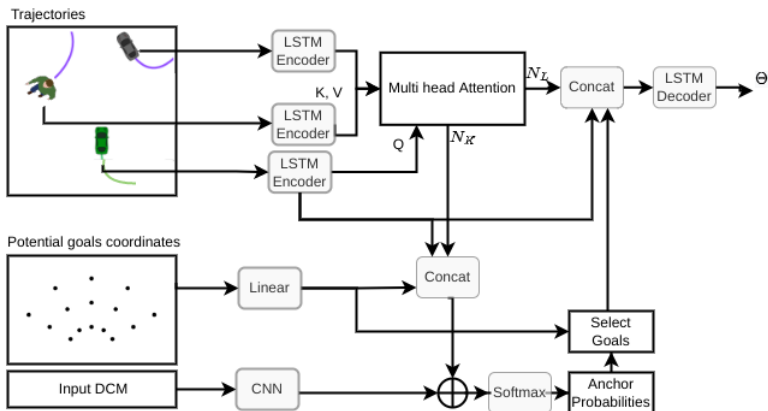
WayDCM

- Follow-up to our previous work : TrajDCM
- Interpretable.
- Lightweight.
- Supervised training.
- Goal based.
- A new type of input : Long Term Waypoints (LTW).

Key insight: The long term waypoint capture most uncertainty of a trajectory.

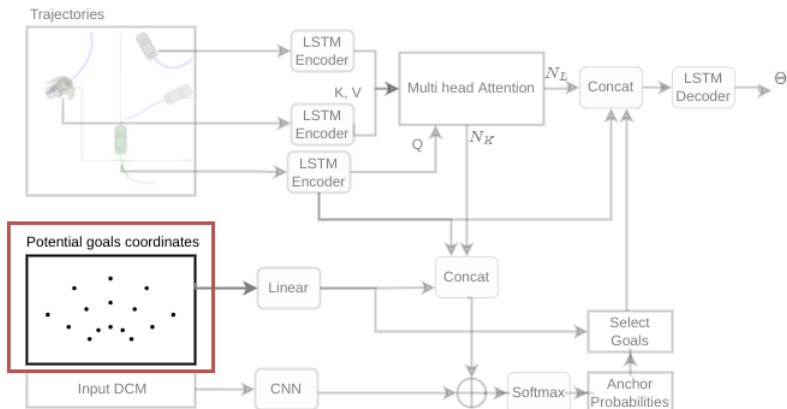
WayDCM

General Architecture



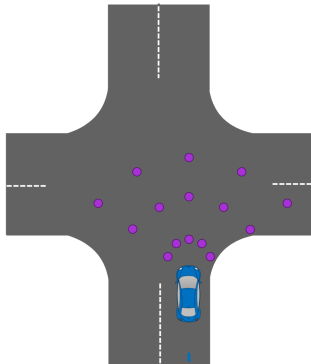
WayDCM

1. Extract potential goals from a dynamic radial grid.

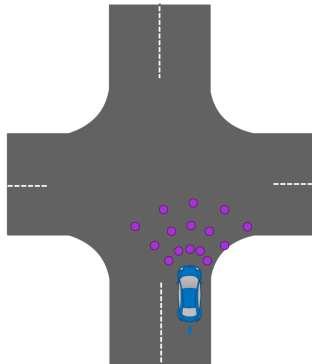


WayDCM

1. Extract potential goals from a dynamic radial grid.



Higher speed.

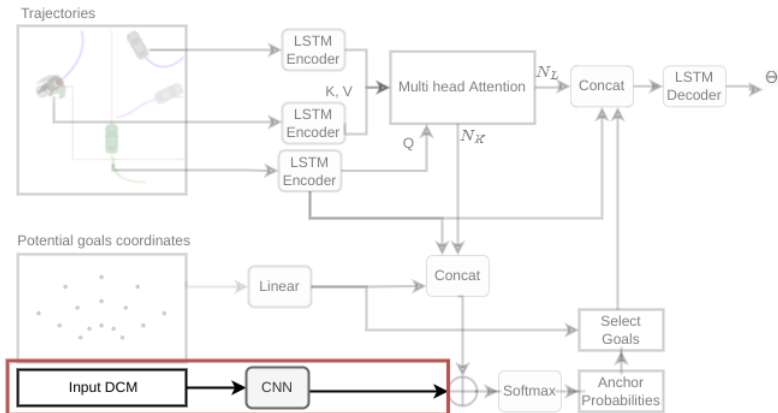


Lower speed.

WayDCM

2. Target Prediction

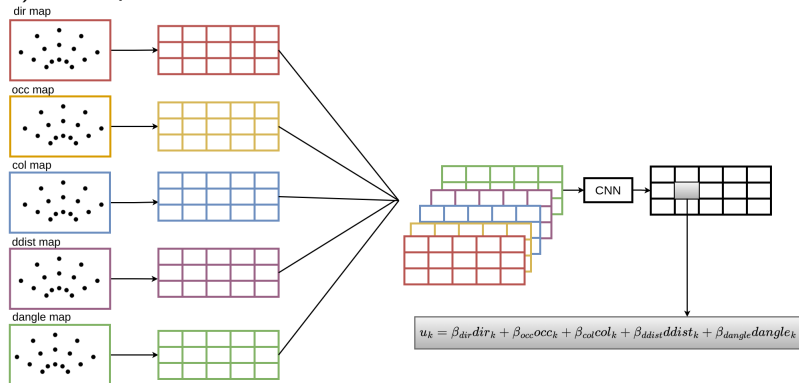
1) Discrete Choice Model (DCM) prediction :



WayDCM

2. Target Prediction

1) DCM prediction :

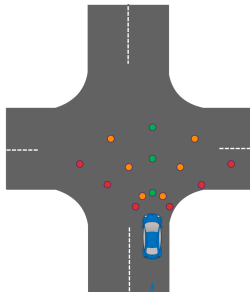


WayDCM

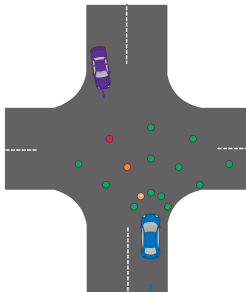
2. Target Prediction

1) DCM prediction :

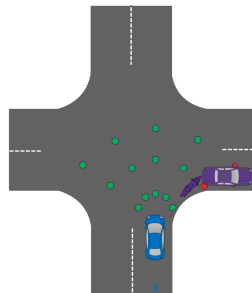
a) Behavioral functions



Keep Direction (dir).



Collision Avoidance (col).

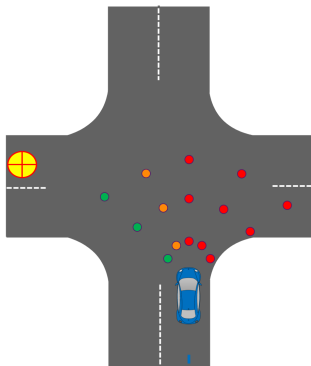


Occupancy/ Overtaking
(occ).

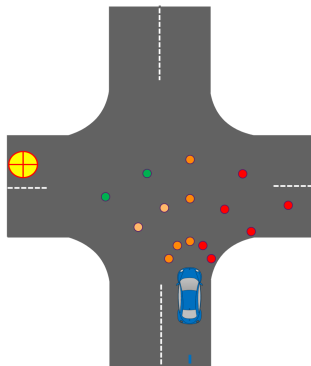
WayDCM

2. Target Prediction

- 1) DCM prediction :
- b) Introduction of the LTW



Minimizing the angle ($dangle$).

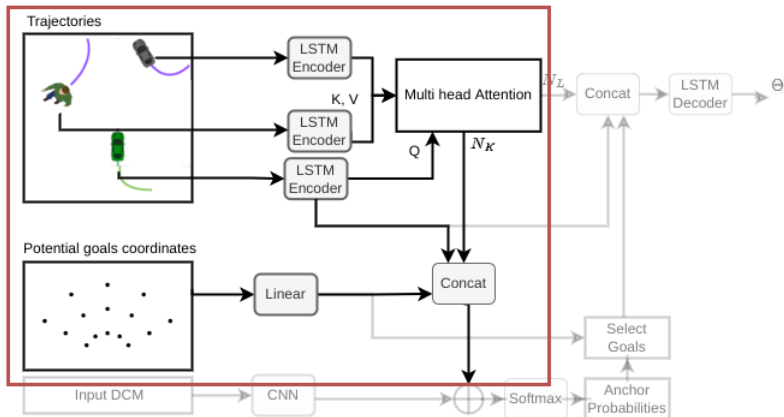


Minimizing the distance ($ddist$).

WayDCM

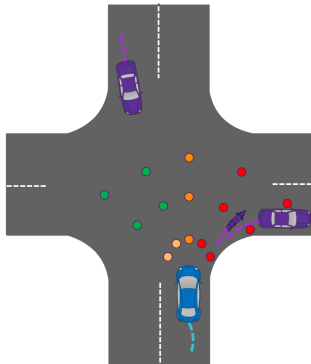
2. Target Prediction

2) Neural Network Prediction :



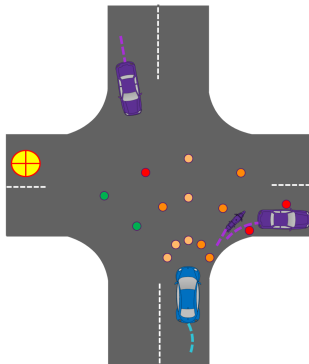
WayDCM

2. Target Prediction



Neural Network Prediction.

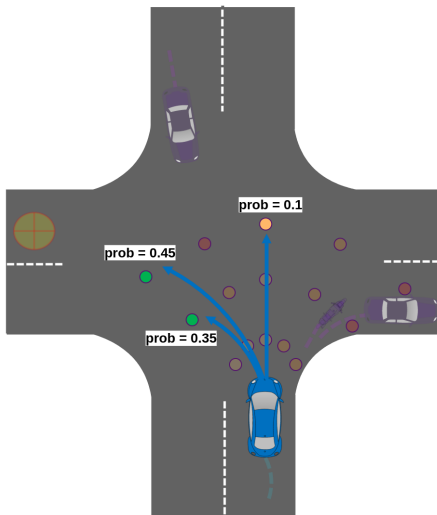
+



DCM Prediction.

WayDCM

3. Motion Estimation : Conditioned on goals.



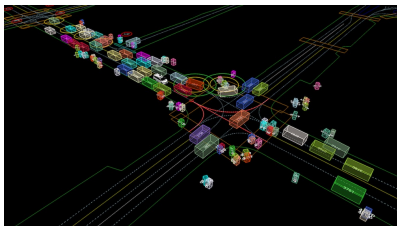
Waymo Dataset

Table: Past, future and prediction horizons on different motion prediction datasets

Dataset	Past (s)	Future (s)	Prediction (s)
nuScenes	3	5	5
INTERACTION	1	3	3
Argoverse 1	2	3	3
Argoverse 2	5	6	6
Waymo	1	8	3, 5 and 8

Waymo Dataset

- Provides 103,354, 20s 10Hz segments (over 20 million frames)
- Collected from different cities in the United States of America (San Francisco, Mountain View, Los Angeles, Detroit, Seattle and Phoenix)



Past time horizon : 1 second

Prediction horizon : 3 seconds

Long term Waypoint : the position of agent at 8 seconds

Results

- $MinADE_6$: Average of pointwise L2 distances between the predicted trajectory and ground truth.
- $MinFDE_6$: Average of pointwise L2 distances between the final points of the prediction and ground truth.

Table: Comparison of different methods on the Waymo validation set

Model	$MinADE_6$	$MinFDE_6$
MotionCNN	0.3365	0.6145
MultiPath++	0.2692	0.4951
LSTM	0.4018	0.8029
MHA-LSTM	0.3141	0.7577
TrajDCM	0.3060	0.7201
WayDCM (without ddist)	0.2779	0.6261
WayDCM	0.2721	0.6037

Results

$$u_k = \beta_{dir}dir_k + \beta_{occ}occ_k + \beta_{col}col_k + \beta_{ddir}ddir_k + \beta_{ddist}ddist_k$$

Table: Estimated parameters β

Model	β_{dir}	β_{col}	β_{occ}	β_{dangle}	β_{ddist}
TrajDCM	-2.70	-0.07	-0.06	-	-
WayDCM (without ddist)	-3.48	-0.09	-0.04	-15.23	-
WayDCM	-2.64	-0.05	-0.06	-10.83	-20.86

Conclusion and future work

- First approach using Long Term Waypoints.
- Long Term Waypoints should be included in motion prediction datasets.
- For future work : use this approach with other goal set representations.

Thank you for your attention !