# **DeTra**: A Unified Model for Object **De**tection and **Tra**jectory Forecasting

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### Task: Object Detection and Trajectory Forecasting

- Input: past LiDAR sweeps and high-definition maps
- Output: Object detections and trajectory forecasts (multiple hypothesis)



#### Background: Previous approaches

- Split reasoning about the present and future into separate cascading modules
- Suffer from narrow interfaces and compounding errors



#### Background: Modular detection-tracking-forecasting

- A detection, tracker, and forecasting modules are cascaded
- These approaches suffer from narrow interfaces and compounding errors



#### Background: Prior end-to-end detection and forecasting

- Tracker is replaced by object features from the LiDAR backbone
- These approaches have wider interfaces, but are still cascading



### DeTra: Formulating the problem as trajectory refinement

- Object queries (features) and poses (coordinates) represent trajectories
  - $\circ$  t = 0 corresponds to the detection
  - $\circ$  t > 0 corresponds to the forecasts
- These are refined jointly over multiple refinement transformer blocks



#### DeTra: Model architecture





#### DeTra: Attention layers





#### DeTra: Refinement Results



#### DeTra: Refinement Results

DeTra i = 0 DeTra i = 1 DeTra i = 2 DeTra i = 3



### DeTra: Comparison results





#### DeTra: Comparison results

#### Scene Transformer

#### Multipath

#### GoRela

#### DeTra





### DeTra: Comparison results

## Multipath Scene Transformer GoRela DeTra -



#### Conclusion

We introduce DeTra, a unified model for object detection and trajectory forecasting

- Formulates detection and forecasting as a single trajectory refinement problem
- Flexible architecture that can handle heterogeneous inputs
- Performs strongly in Argoverse 2 Sensor and Waymo Open Dataset
- Design choices are validated through extensive ablations
  - Refinement is key
  - Leveraging geometric priors in cross-attention is important
  - Every proposed component has a positive contribution