

PlanT: Explainable Planning Transformers via Object-Level Representations

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Figure 1: Driving Pipeline (© University of Tübingen)





- We consider the task of planning in an autonomous driving stack
- Planning is often done as a rule-based system

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• We propose a learned planner

Motivation

- Rule-based planning
- Hard to scale
- + Interpretable

End-to-End models

- Scales with data +
- Not interpretable

Learned planner: PlanT Best of both worlds \rightarrow

Results



- We add a perception module to the driving stack
- With the full model we obtain state of the art on the longest 6 benchmark

Explainability



Figure 5: PlanT attention (© University of Tübingen)

- Visualization of **attention weights** to show the **most important object**
- **Temporarily** more **consistent** than the

 Scaling dataset and model improves performance

• Expert level performance

Architecture



Figure 4: PlanT Architecture (© University of Tübingen)

- We train a standard transformer encoder from scratch
- The model is trained with a loss on **future positions** of the ego vehicle and the other vehicles

CNN-based method + also takes geometrically distant objects into account



Code Check out our paper and code



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