

Lane Detection Datasets Report

Anubhav Jha, Ashmit Gupta, Saahil Kapoor

September 11, 2025

Abstract

This report provides a comprehensive overview of lane detection datasets, including TuSimple, CULane, LLAMAS, CurveLanes, and VIL-100. We summarize dataset statistics, annotation formats, evaluation metrics, limitations, and their importance in advancing autonomous driving research.

1 Introduction

Lane detection is a vital perception task for autonomous driving systems. Benchmark datasets play a central role in developing and testing lane detection algorithms. The TuSimple dataset tus [2017] was one of the earliest benchmarks but is limited to daytime U.S. highway driving. To address its gaps, additional datasets such as CULane Pan et al. [2018], LLAMAS Behrendt and Mielenz [2019], CurveLanes Xu et al. [2021], and VIL-100 Xu et al. [2022] were introduced to cover diverse driving conditions.

2 Dataset Description

2.1 Dataset Statistics

Table 1 provides statistics of major lane detection datasets.

2.2 Annotation Format

Most datasets use polyline-based lane annotations. TuSimple employs row-anchor based polylines in JSON format. CULane and CurveLanes provide lane segmentation masks. LLAMAS focuses on highway lane polylines, while VIL-100 provides dense annotations for video sequences.

3 Evaluation Metrics

TuSimple evaluates models using lane accuracy (percentage of correctly predicted lane points). CULane employs Intersection-over-Union (IoU) with precision/recall across categories like

Table 1: Comparison of Lane Detection Datasets

Dataset	Year	Images/Clips	Resolution	Scenes	Conditions
TuSimple	2017	6,408 images	1280×720	Highway (US)	Daytime, clear
CULane	2018	133,235 images	1640×590	Urban, Highway	Day/Night, occlusion, curves
LLAMAS	2019	100k images	1276×717	Highways (Germany)	Daytime
CurveLanes	2021	150k images	1640×590	Challenging curved roads	Day/Night, adverse conditions
VIL-100	2022	12k video clips (100 frames)	1920×1080	Urban (China)	Diverse lighting, traffic

night, shadow, and occlusion. VIL-100 introduces video-based evaluation, enabling temporal consistency assessment.

4 Applications in Research

Lane detection datasets have been used to develop diverse models:

- Instance Segmentation Approaches (e.g., LaneNet Neven et al. [2018]) achieved real-time lane segmentation.
- Parametric Curve Models (e.g., BezierLaneNet Zheng et al. [2022]) fit lanes with Bezier curves.
- Transformer-based Models (e.g., PriorLane Zhou et al. [2022]) leverage prior knowledge for robust detection.
- Video-based models trained on VIL-100 learn temporal consistency across frames.

5 Limitations

- TuSimple is restricted to daytime U.S. highways.
- CULane, though diverse, is limited to Beijing traffic.
- LLAMAS lacks urban driving conditions.
- CurveLanes emphasizes curves but not general traffic diversity.
- VIL-100 is relatively small.

6 Conclusion

TuSimple pioneered lane detection benchmarks, but newer datasets have introduced diverse environments, weather, and temporal aspects. Future research should combine highway, urban, and video-based datasets for comprehensive model evaluation.

References

- Tusimple lane detection challenge. <https://github.com/TuSimple/tusimple-benchmark>, 2017.
- Karsten Behrendt and Holger Mielenz. Unsupervised llamas dataset. <https://unsupervised-llamas.com>, 2019.
- Davy Neven, Bert De Brabandere, Stamatios Georgoulis, Marc Proesmans, and Luc Van Gool. Towards end-to-end lane detection: An instance segmentation approach. In *IV*, 2018.
- Xingang Pan, Jianping Shi, Ping Luo, Xiaogang Wang, and Xiaoou Tang. Spatial cnn for traffic scene understanding. In *CVPR*, 2018.
- Qi Xu et al. Curvelanes: A new benchmark for curve lane detection. *arXiv preprint arXiv:2108.11544*, 2021.
- Yujun Xu et al. Vil-100: A new video lane detection dataset and benchmark. *arXiv preprint arXiv:2204.01361*, 2022.
- Lianqing Zheng, Jian Wu, et al. Bezierlanenet: Lane detection with bezier curve modeling. *arXiv preprint arXiv:2203.02431*, 2022.
- Yan Zhou et al. Priorlane: Incorporating prior knowledge into lane detection. *arXiv preprint arXiv:2209.06994*, 2022.