

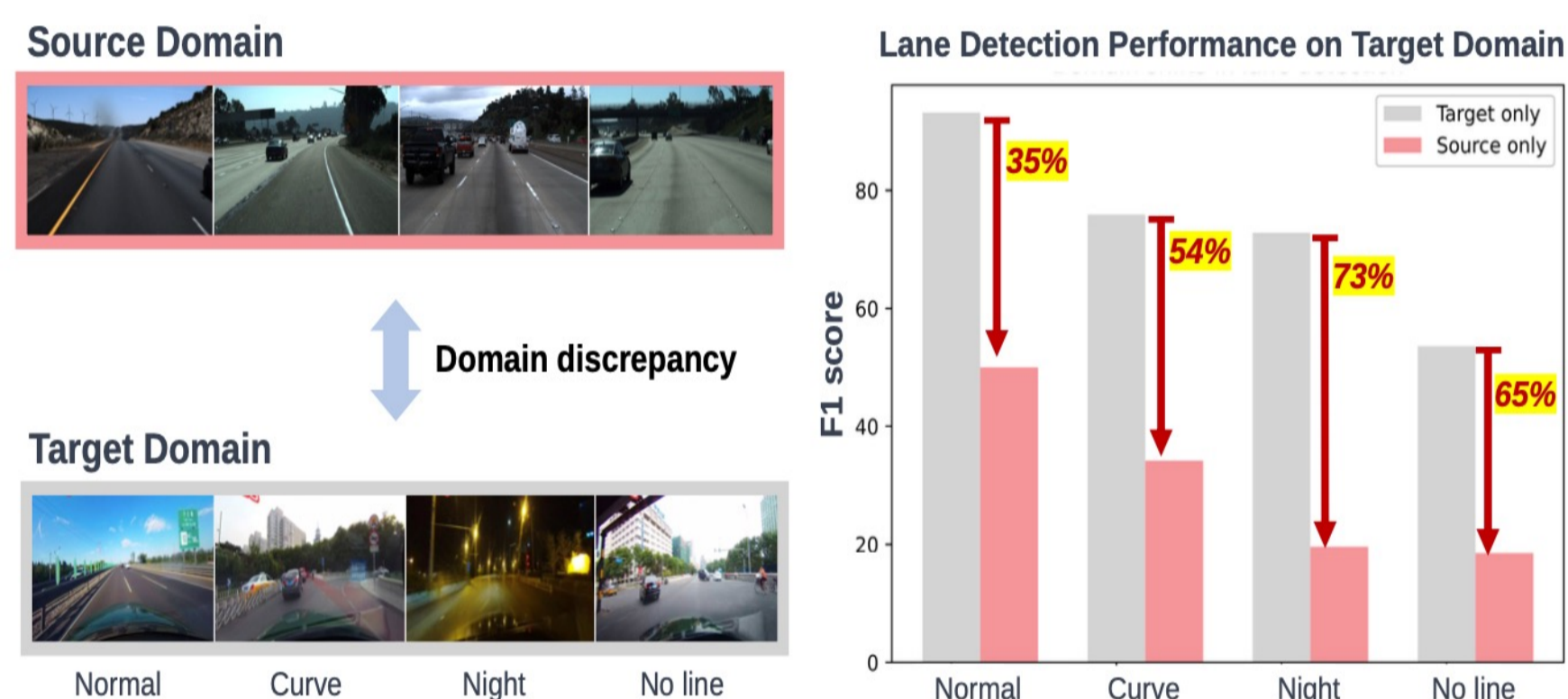
# HD Maps are Lane Detection Generalizers: A Novel Generative Framework for Single-Source Domain Generalization

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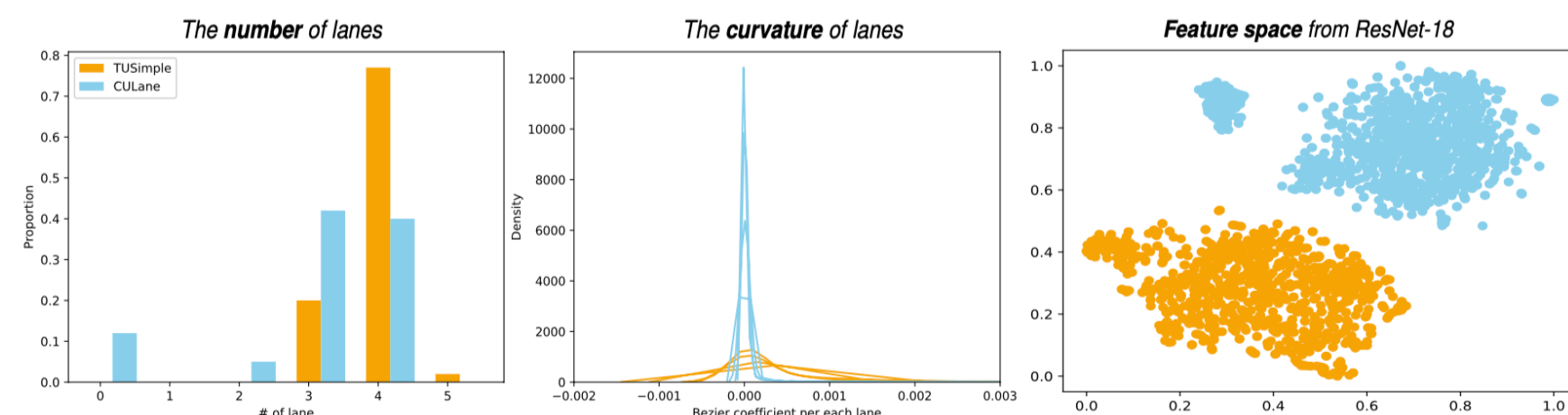
## Motivation

### □ The overall framework using HD Maps



Current data-driven lane detection models still suffer from **performance degradation** when tested on unseen domains.

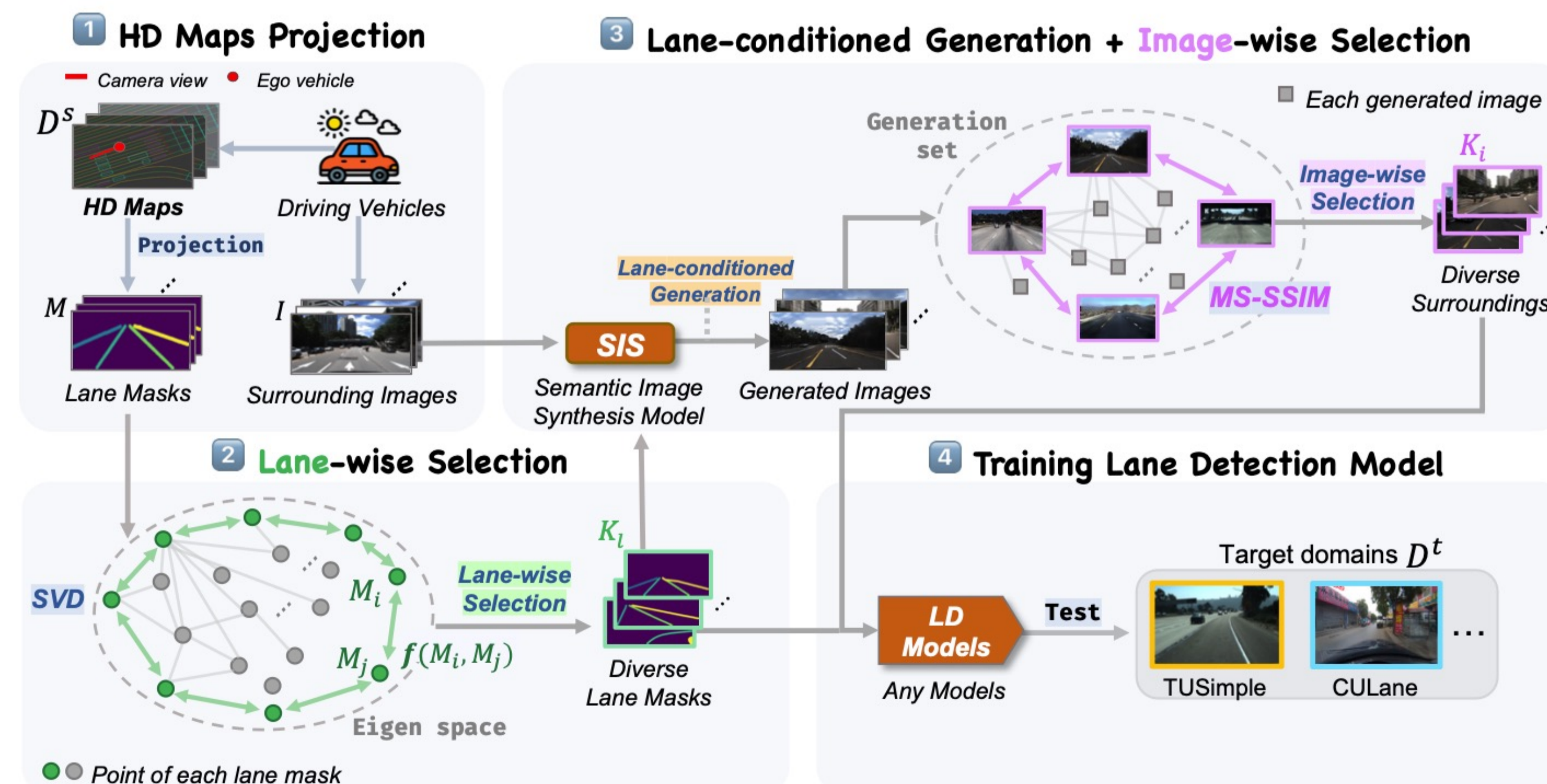
### □ Domain Discrepancy in Lane Detection



We categorized the cause of domain discrepancy in lane detection to **(i) Lane structures** and **(ii) Surrounding images**.

## Methodology

### □ The overall framework using HD Maps



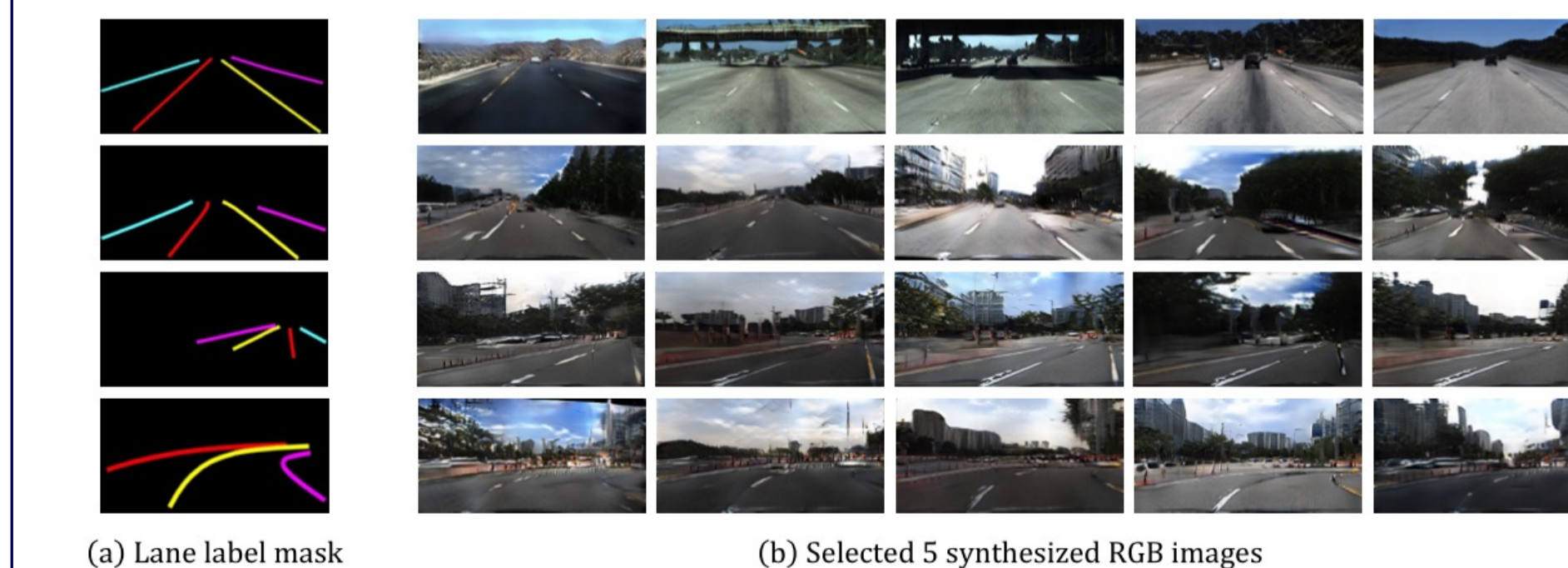
We first **extract** numerous lane masks from **HD Maps** with surrounding images. Then, we train SIS (Semantic Image Synthesis) models to synthesize **lane-conditioned images** with **diverse** surroundings. We select a core set among synthesized images with two criteria to maximize its diversity: **(i) lane structure** and **(ii) surroundings**. Then, we finally train the lane detection model and obtain enhanced **generalization** performance.

## Experiments

### □ Quantitative Results on CULane

Experiment setting	Trained data			Total	Normal	Crowded	Night
	TUSimple	CULane	LabsLane				
Source only - ERFNet	✓			24.2	41	19.6	9.1
Source only - GANet	✓			30.5	50	25.5	19.6
Advent [24]	✓	✓		30.4	49.3	24.7	20.5
PyCDA [13]	✓	✓		25.1	41.8	19.9	13.6
Maximum Squares [1]	✓	✓		31	50.5	27.2	20.8
MLDA [9]†	✓	✓		<u>38.4</u>	<u>61.4</u>	<u>36.3</u>	<u>27.4</u>
Source only - GANet			✓	34.9	52.4	25.9	27.1
Ours - ERFNet			✓	38.2	<u>56.8</u>	32.5	24.4
Ours - GANet			✓	<u>39.6</u>	56.7	34	<u>30.1</u>

### □ Qualitative Results from Generative Model



## Conclusion

We propose a novel training framework that utilizes HD Maps for robust lane detection