



Exorcising “Wraith”: Protecting LiDAR-based Object Detector in Automated Driving System from Appearing Attacks

32ND USENIX
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Qifan Xiao, Xudong Pan, Yifan Lu, Mi Zhang*, Jiarun Dai, Min Yang*

System and Software Security Lab

School of Computer Science

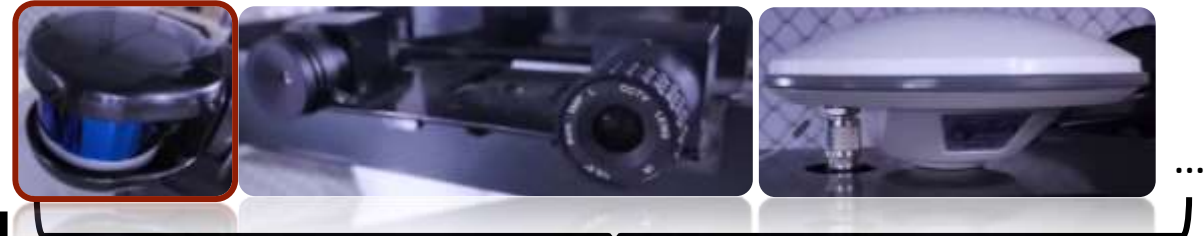
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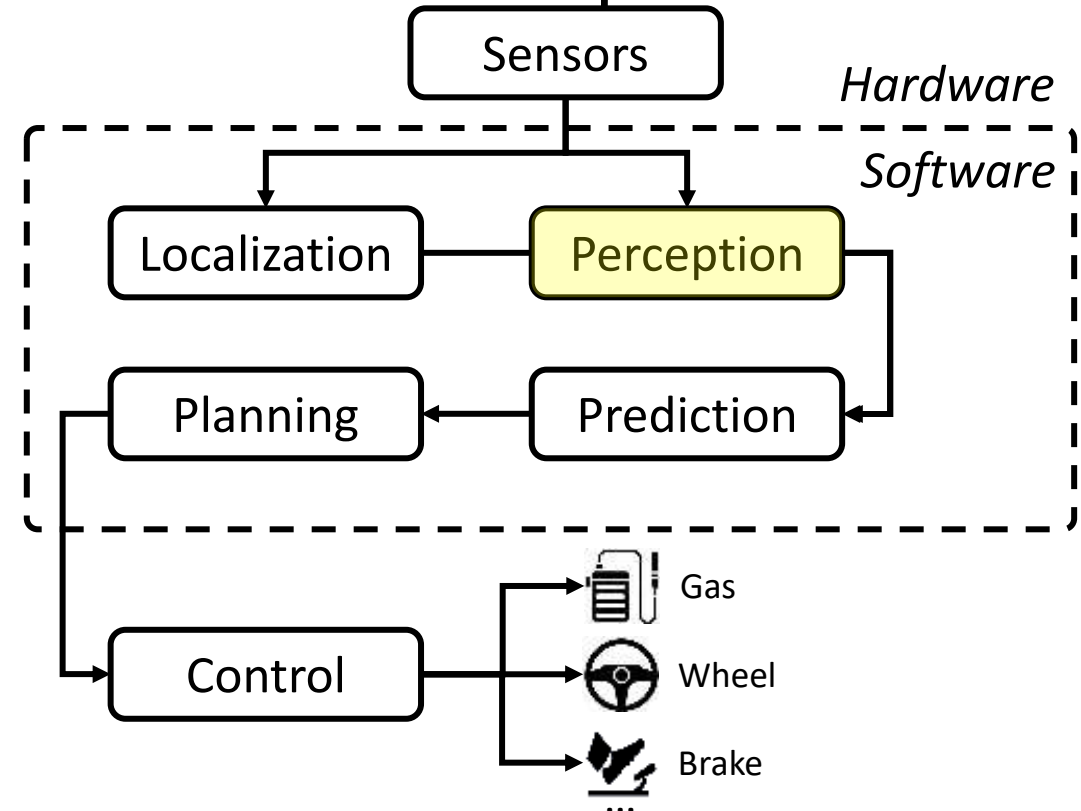


LiDARs in Automated Driving System

- Most ADS companies take LiDARs as main sensors



ADS Company	LiDAR Type	LiDAR as main sensor?	Open-Source?
	Velodyne	Y	Y
	unknown	Y	N
	/	N	N
	FirstLight	Y	N
	IRIS	Y	N

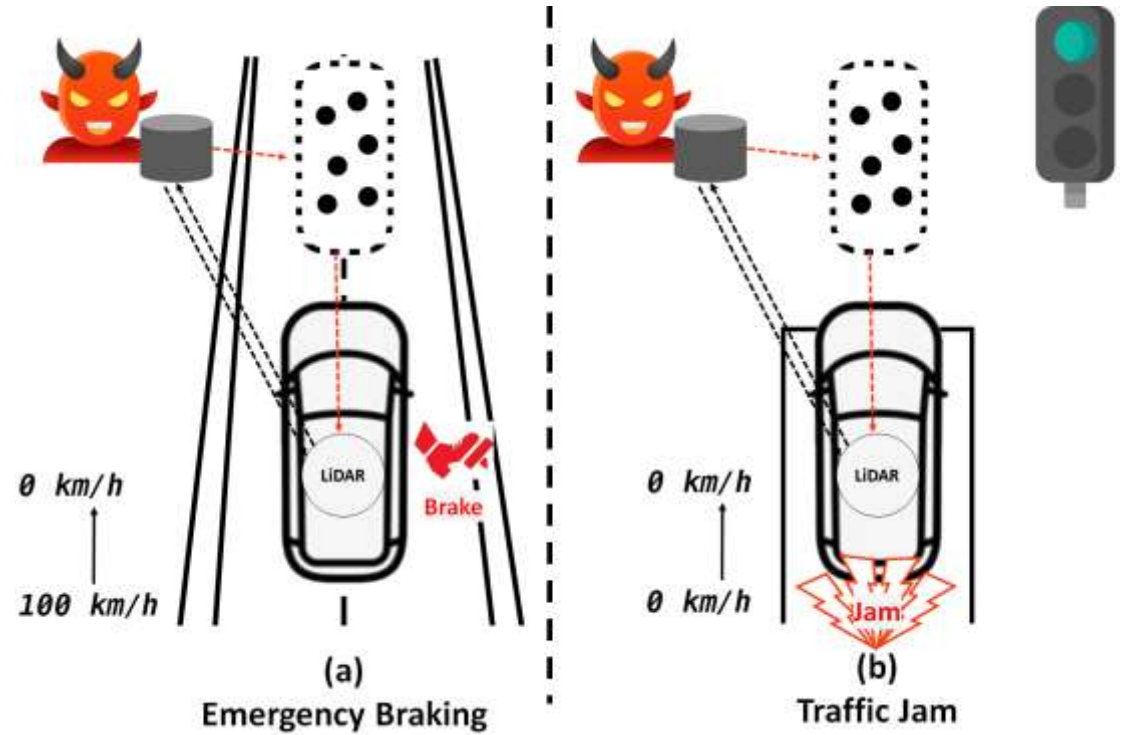
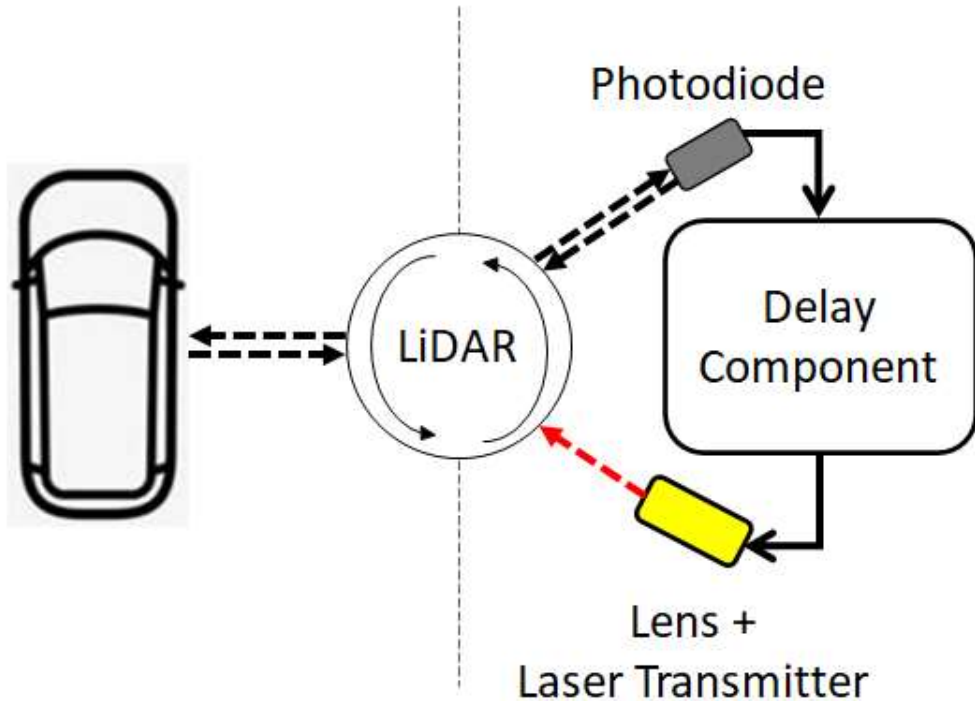


Threats of Appearing Attacks

- Injecting points into LiDAR point clouds
 - Photodiode captures the lasers sent by LiDAR
 - Laser transmitter sent back the fake reflected lasers

- Forging non-existent vehicles to pose threat
 - Forcing the ADS vehicle to emergency brake
 - Keeping the ADS vehicle immobile

Normal Detection



The Magic of Such Attacks

1. Practicability

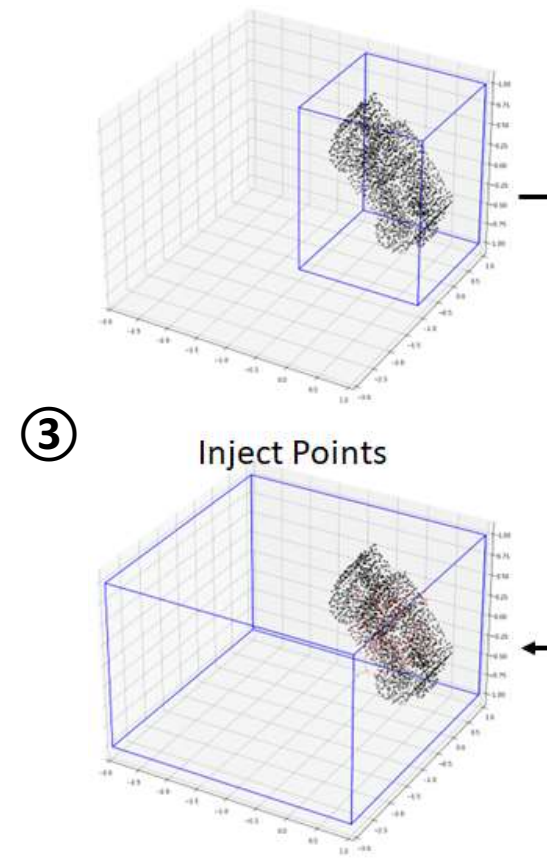
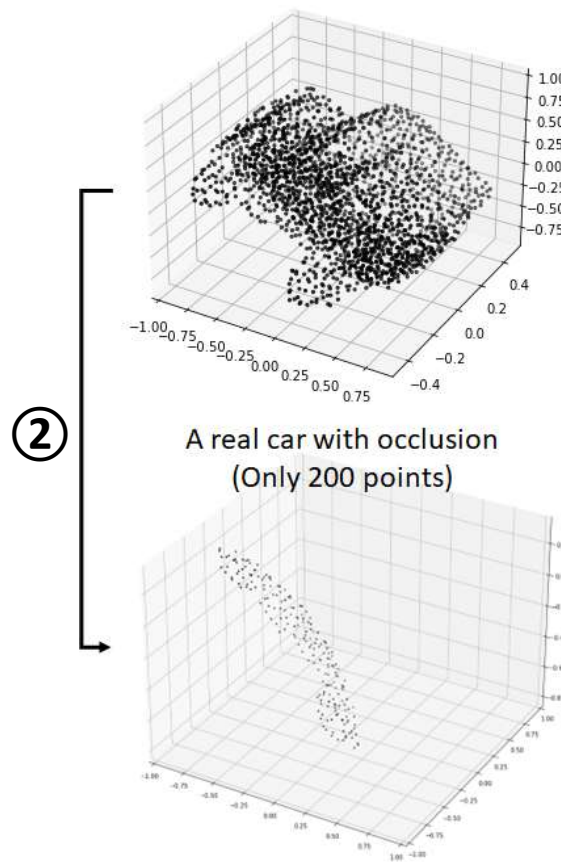
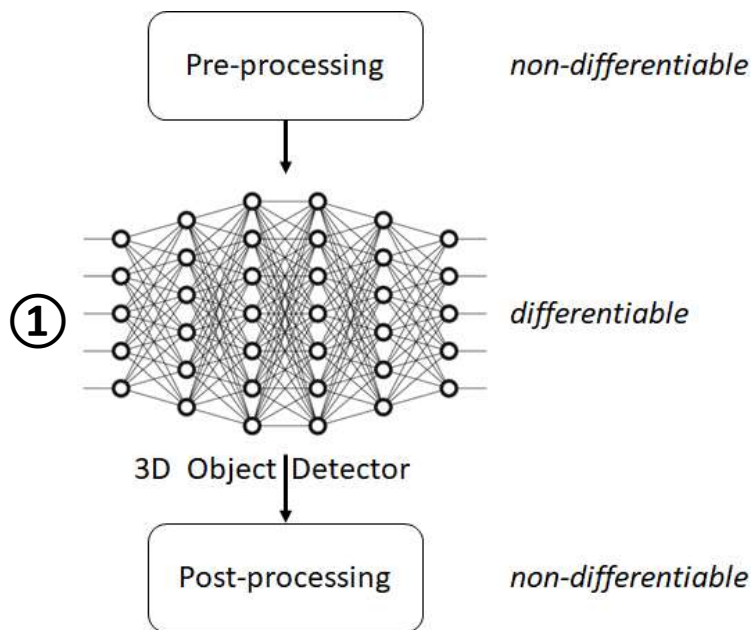
- Reusable traditional adversarial methods (FGSM, PGD, C&W...)

2. Naturalness

- Difficult for human to distinguish

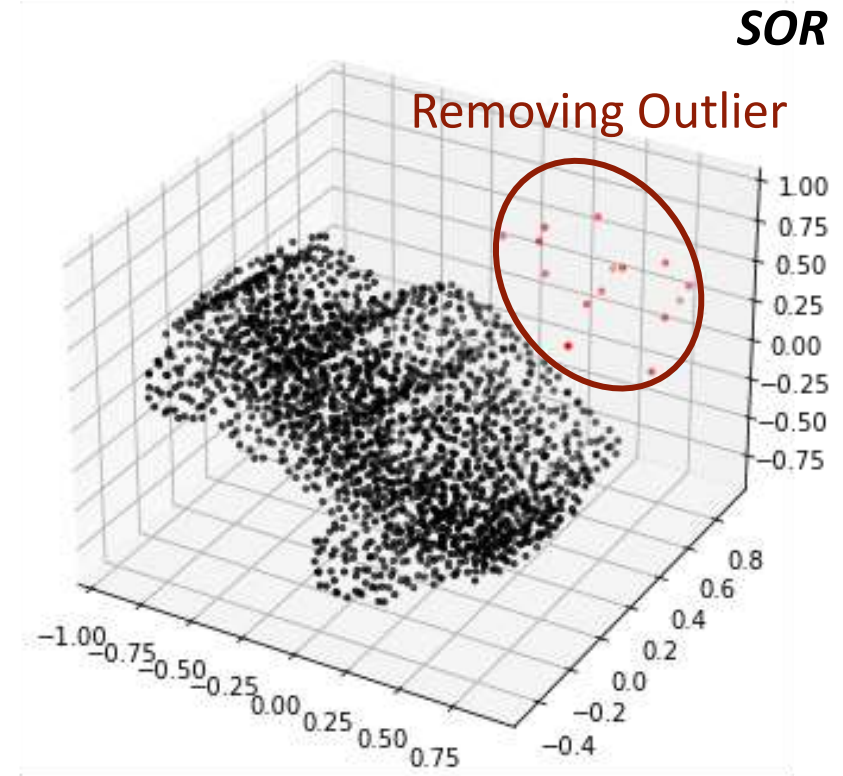
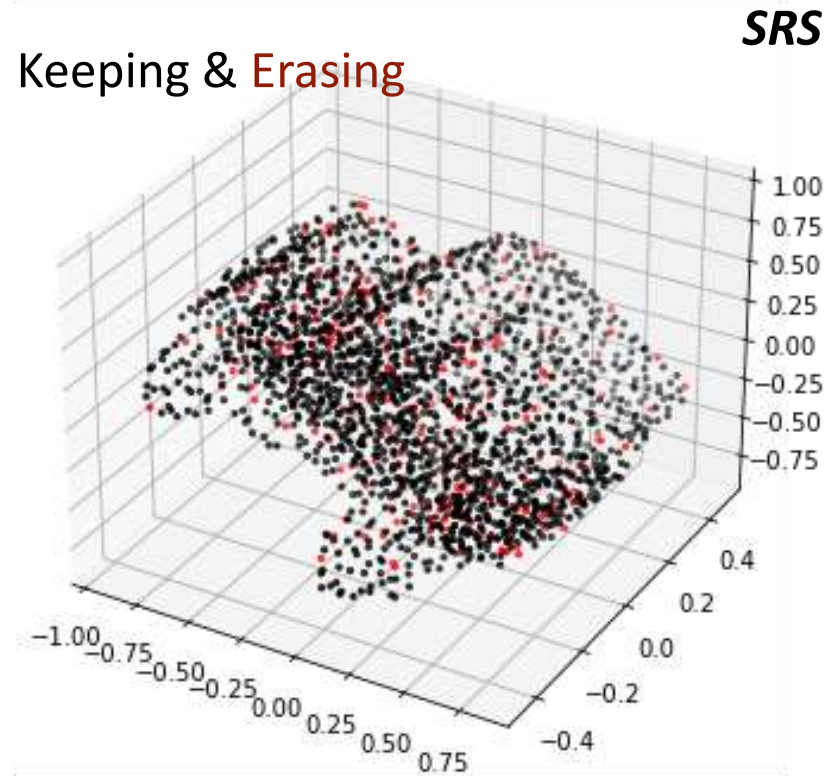
3. Variability

- Various attack goals



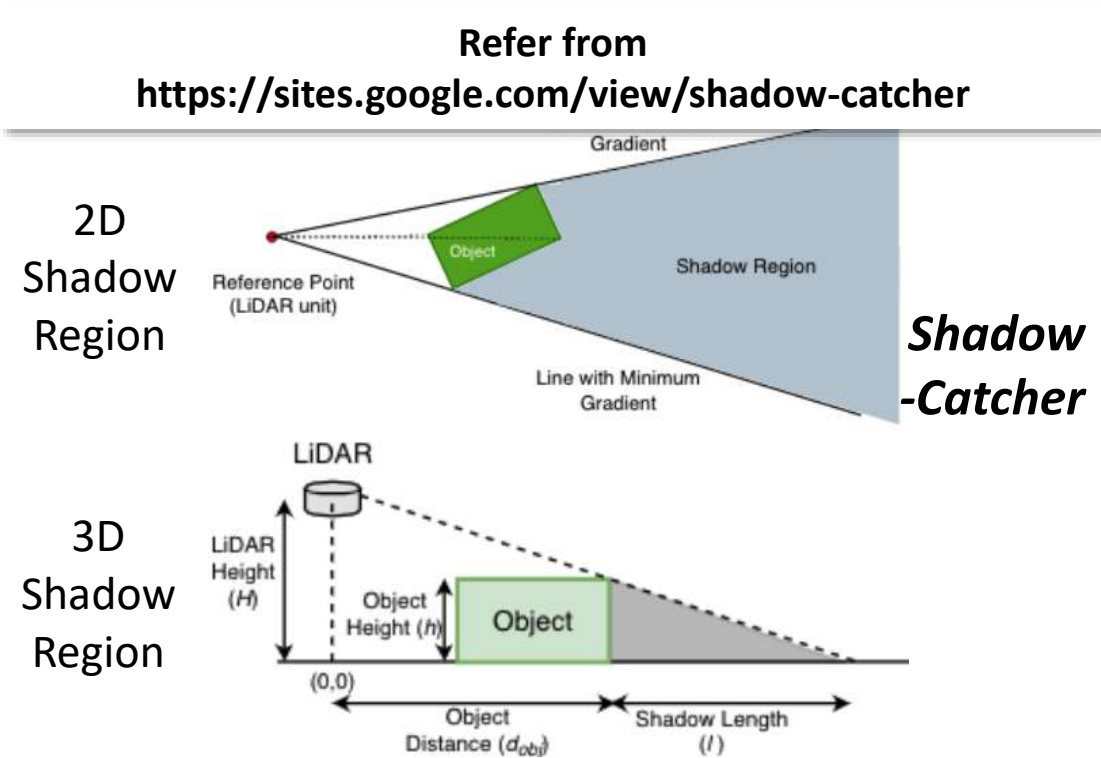
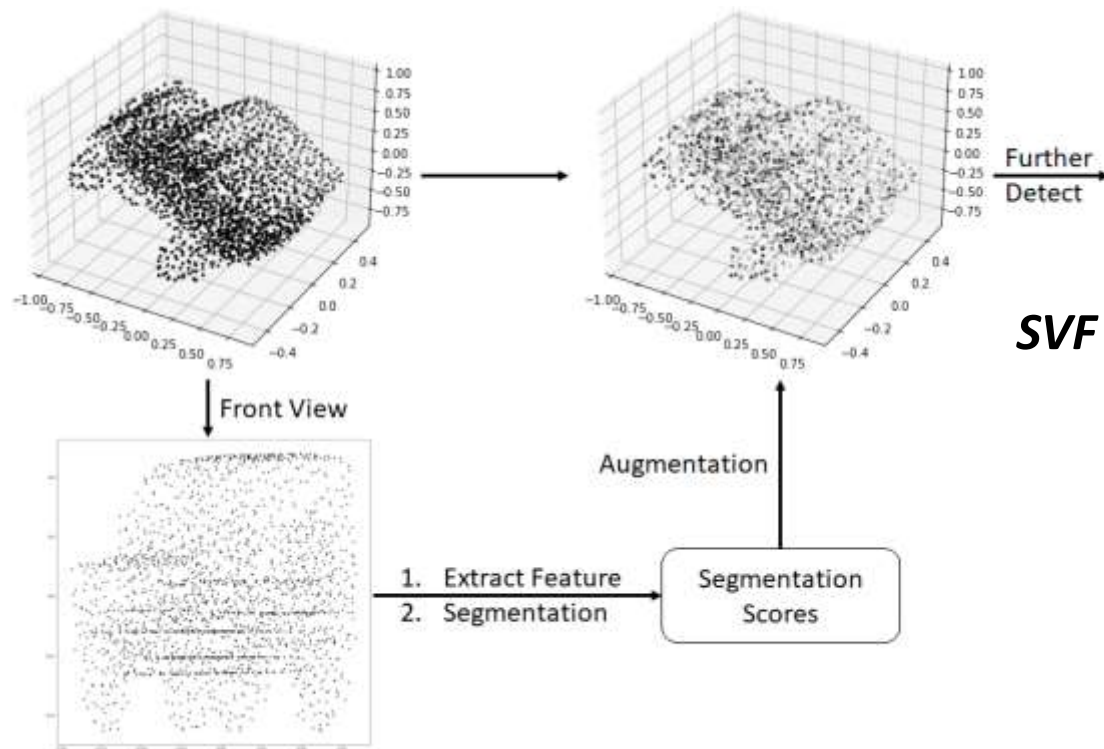
Existing Defense Methods

- **Universal Defenses**
 - Initial Motivation: Improving the robustness of PC models against noise
 - SRS and SOR



Existing Defense Methods

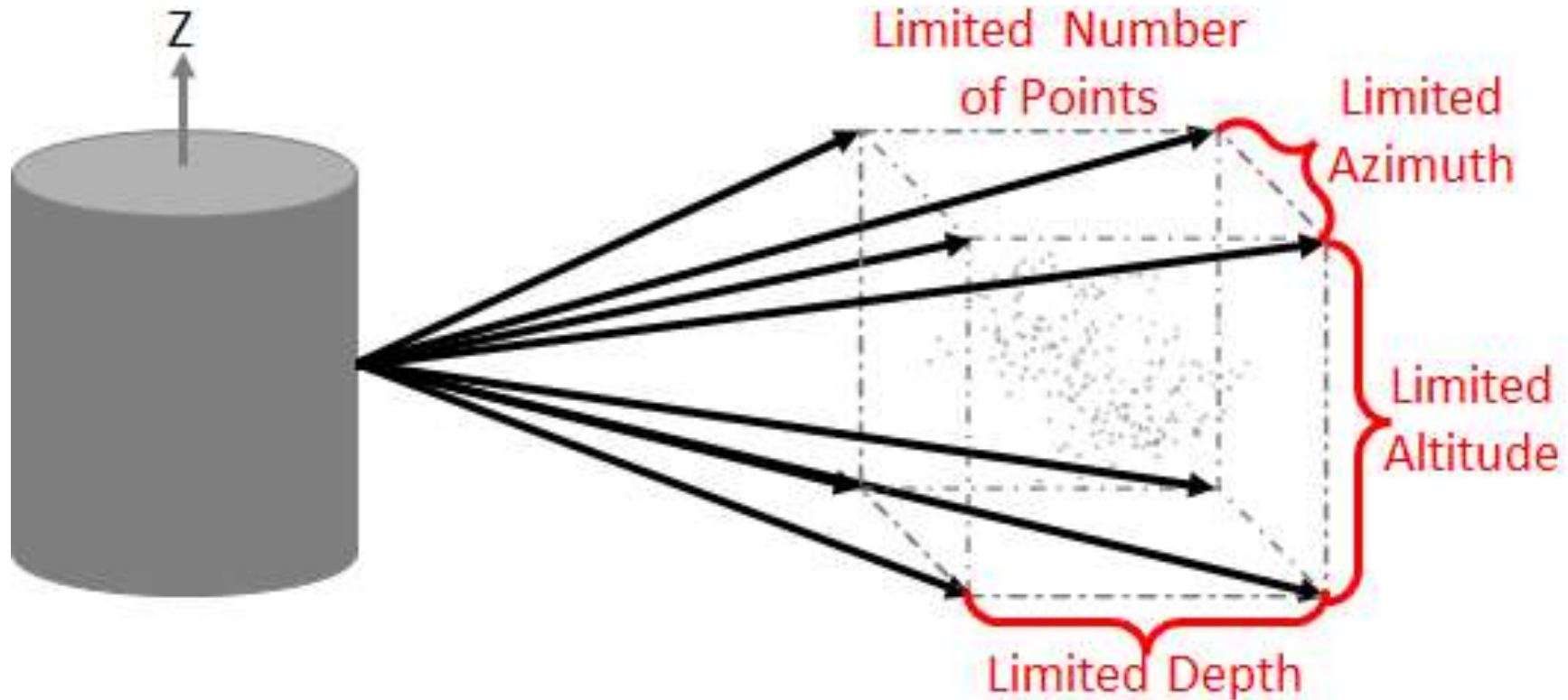
- **Specific Defenses**
 - Initial Motivation: Mitigating specific attack methods
 - SVF, CARLO and Shadow-Catcher



Limitations of Existing Attacks

- **Two Common Limitations**

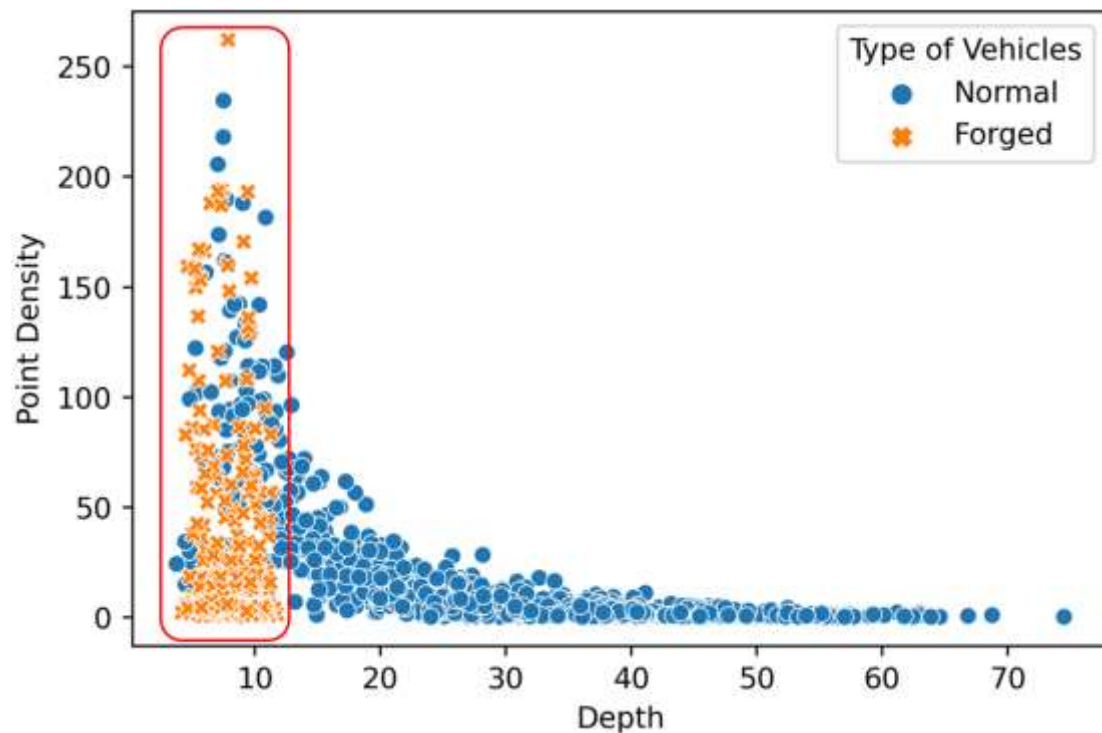
1. Constrained by the **attack device** → the **position** and **number** of forged points
2. Constrained by the **attack goal** → the **shape** of forged objects



Defense Insight

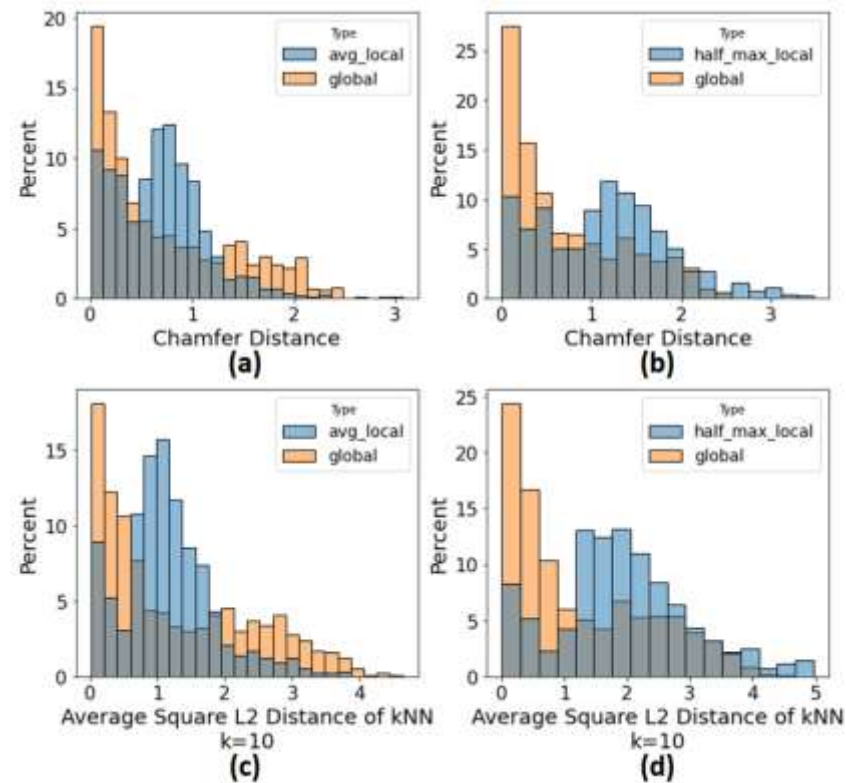
1. On the Position and Number

- the distributions of point density and depth are different



2. On the Shape

- the local difference is mostly larger than the global difference



LiDARs in Automated Driving System

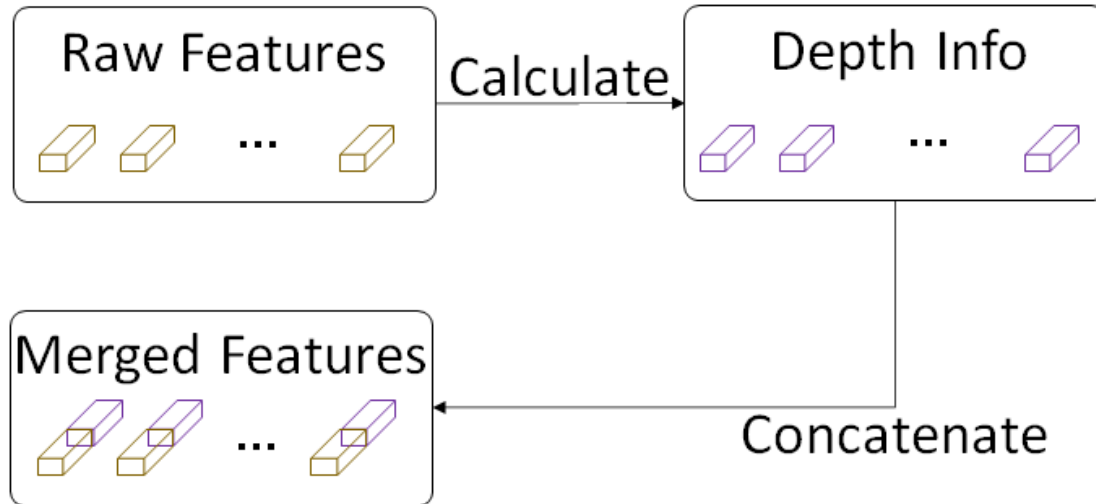
1. On the Position and Number

- Modeling the depth-density relation

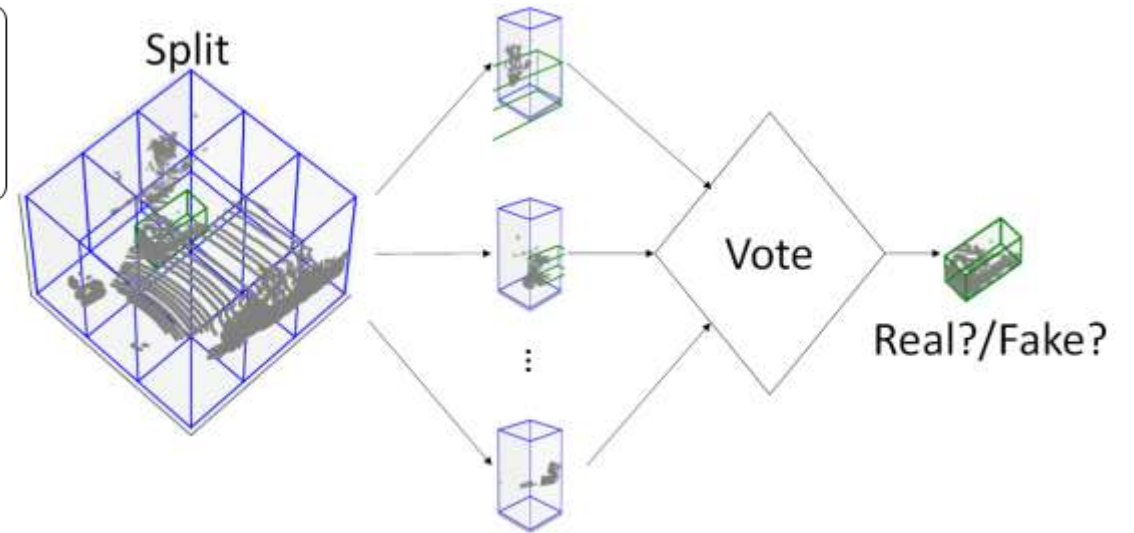
2. On the Shape

- Deploying local detector + Voting

Explicit depth feature & Implicit density feature

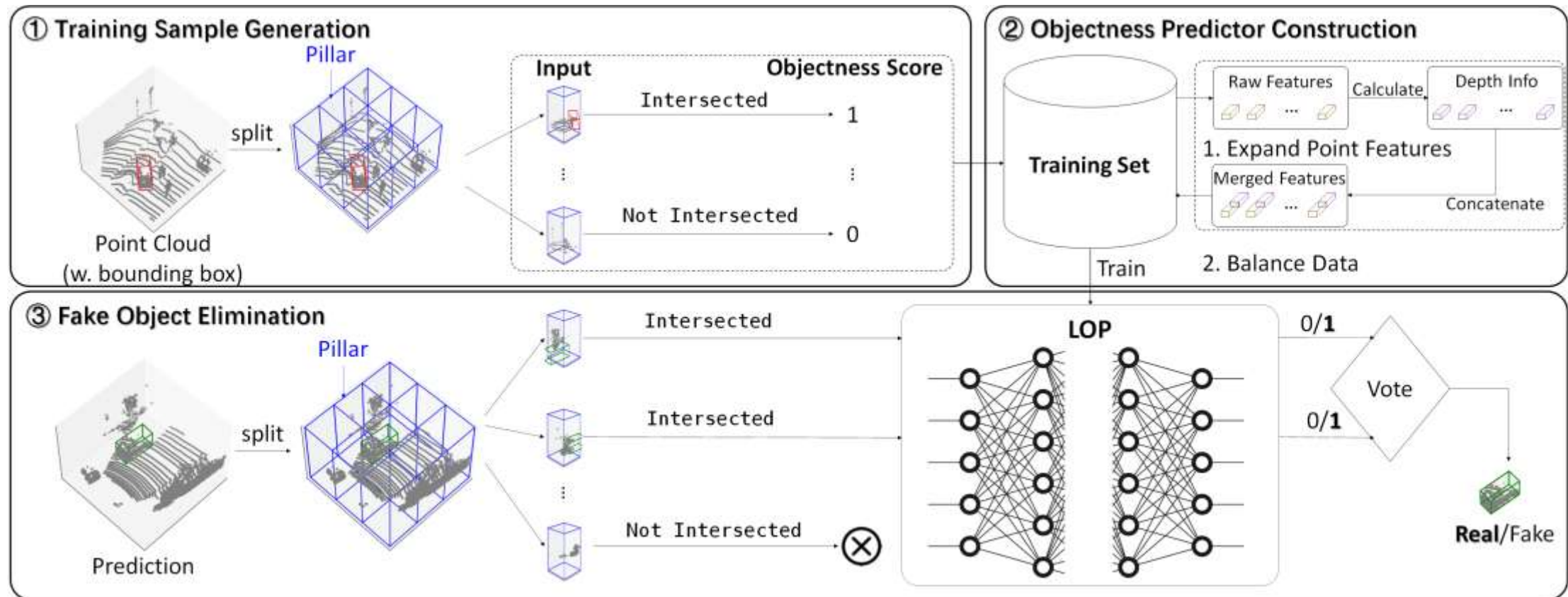


Split the prediction by splitting input space



Our Proposed Method

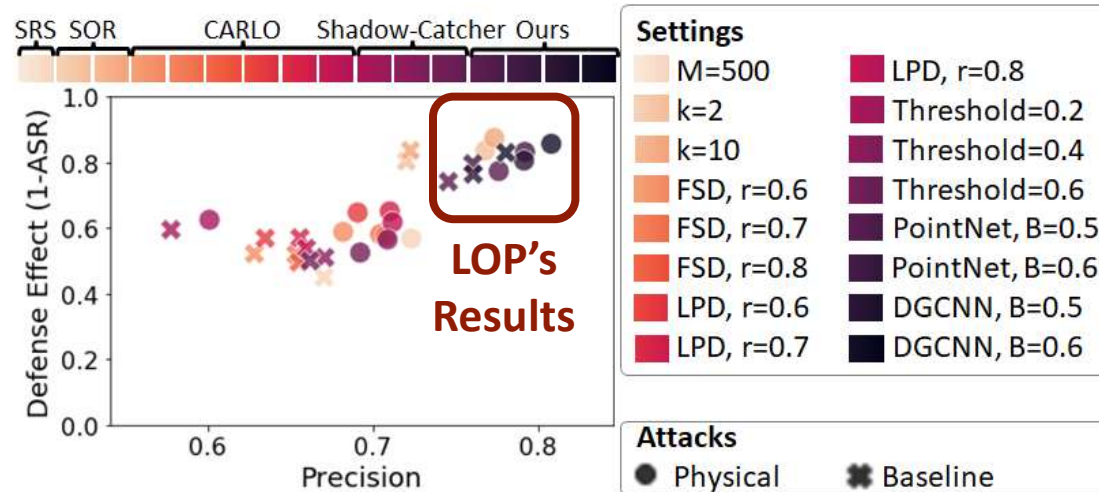
- **Local Objectness Predictor**
 - **Plug-and-Play Design** (**No need to retrain the whole detector*)



Defense Effectiveness

- More improvement on the performance and robustness of protected 3D object detectors

- Acceptable costs on memory and slightly larger costs on time



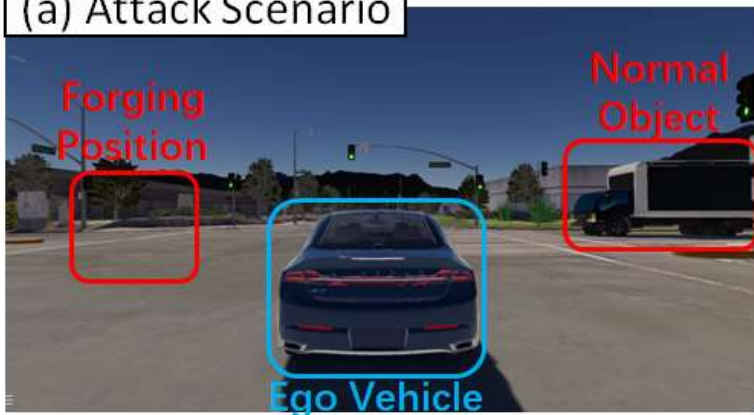
	Time per sample (s)	GPU Mem (MB)	CPU Mem (MB)
None	0.060±0.005	1477	2551
SRS	0.069±0.007	1473	2549
SOR	0.114±0.005	5827	2516
Carlo (LPD)	0.503±0.003	1477	2552
Carlo (FSD)	2.463±0.005	1477	2506
Shadow-Catcher	0.089±0.002	1477	2551
Ours (PointNet)	1.341±0.011	2283	2518
Ours (DGCNN)	1.589±0.013	3747	2506

can further reduce by multi-processing

Simulation Experiments

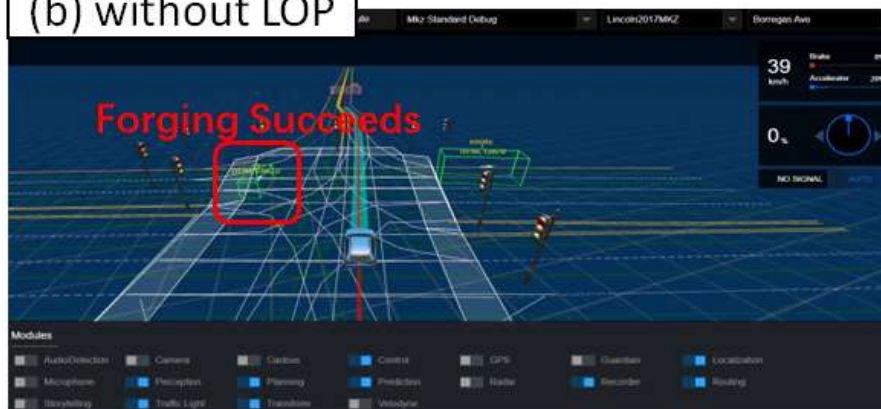
- The performance of Apollo 6.0.0 deployed with LOP, evaluated in LGSVL simulator

(a) Attack Scenario

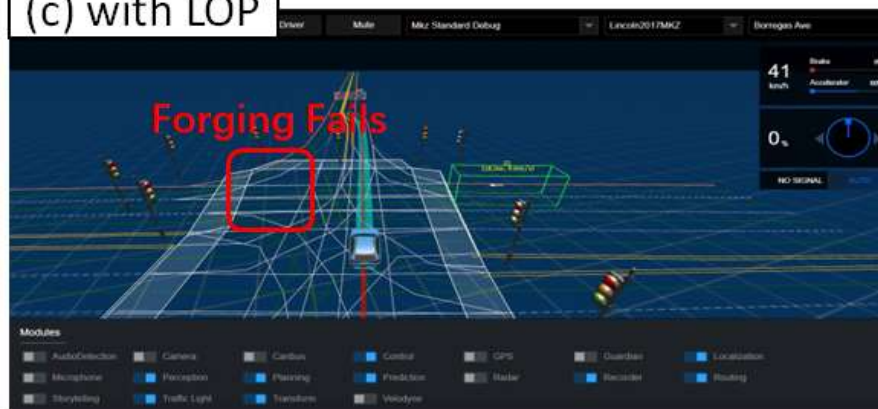


	Precision	ASR	time cost (ms)	FPS
Apollo 6.0.0 (w/o. LOP)	8.33%	53.66%	33.36ms	29.97
Apollo 6.0.0 (w/. LOP)	100.00%	0.00%	42.48ms	23.54

(b) without LOP



(c) with LOP

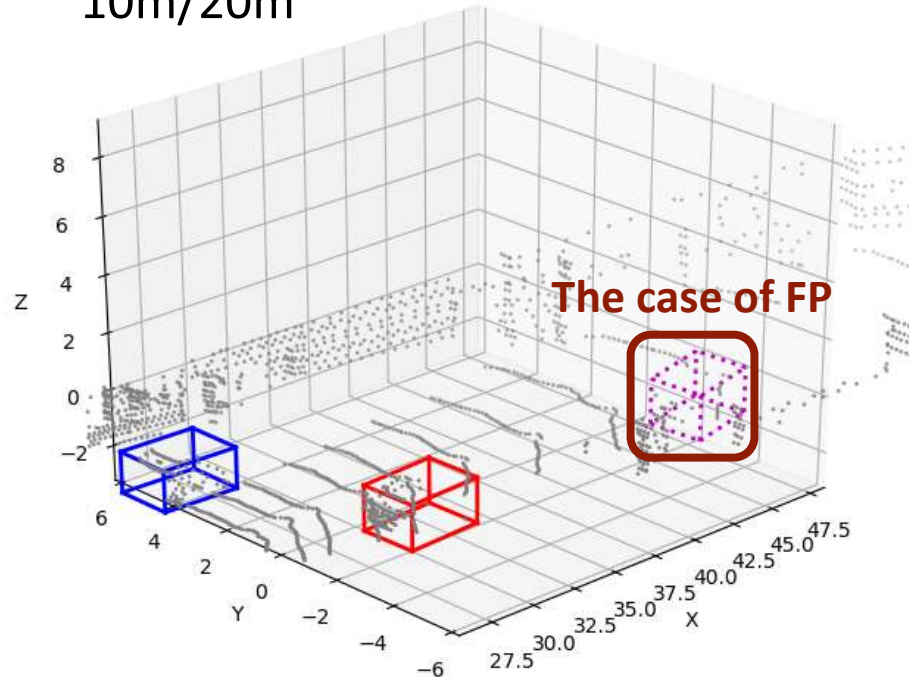


Future Directions

Direction 1

The Existence of False positives

- Farther objects are harder to detect
 - 12.95%/16.53% of FP with depth < 10m/20m



Direction 2

The Upgrade of Attack Device

- The maximum of forged points is already up to 2500

The Fig.7 in PLA-LiDAR (S&P 2023)



Take Away Message

1. We conclude the limitations of existing appearing attacks
2. We propose a plug-and-play defense method LOP
3. We prove the effectiveness of our LOP online and offline



↓ Full paper of our LOP



More Research on AI Security



Thank you for your Audience!

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