### PLA-LiDAR:

# Physical Laser Attacks against LiDAR-based 3D Object Detection in Autonomous Vehicle

**Zizhi Jin**<sup>1</sup>, Xiaoyu Ji<sup>1</sup>, Yushi Cheng<sup>1,2</sup>, Bo Yang<sup>1</sup>, Chen Yan<sup>1</sup>, Wenyuan Xu<sup>1</sup>

<sup>1</sup>Zhejiang University, <sup>2</sup>Tsinghua University,











### **LiDAR**

- □ LiDAR (Light Detection and Ranging) is widely used for perception.
- ☐ Correct LiDAR perception provides the foundation for safety in self driving.



**Self-driving Car** 



Robots

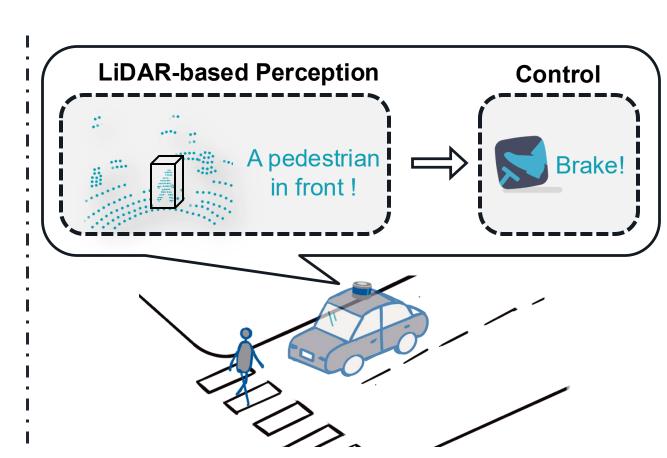


**CVIS** 



**Drones** 

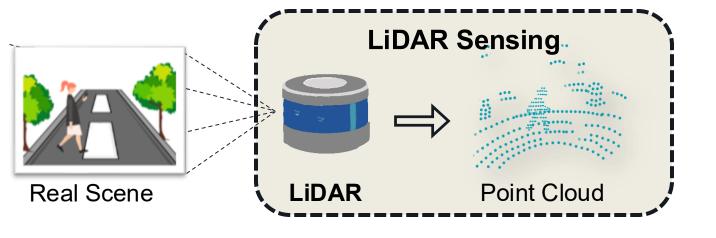
Source: www.velodynelidar.com

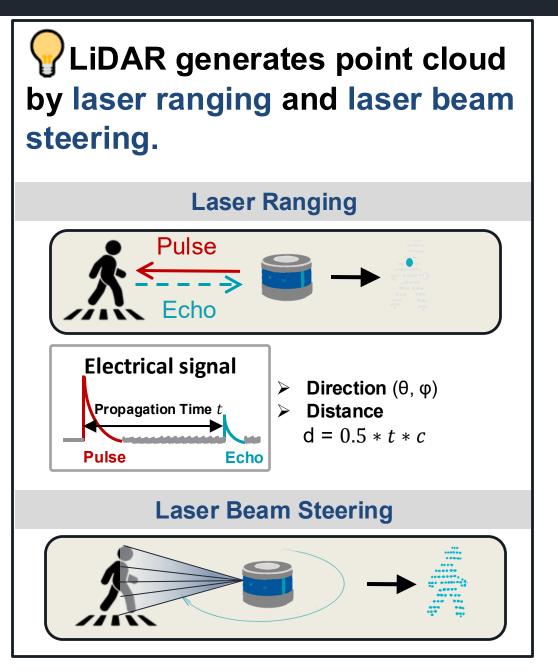






### How Does LiDAR work?

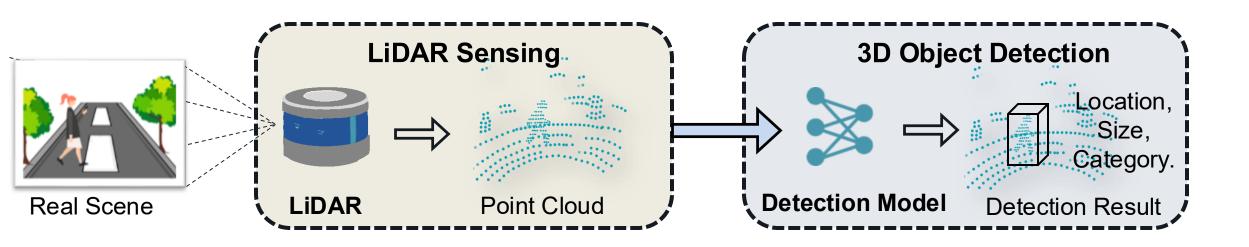








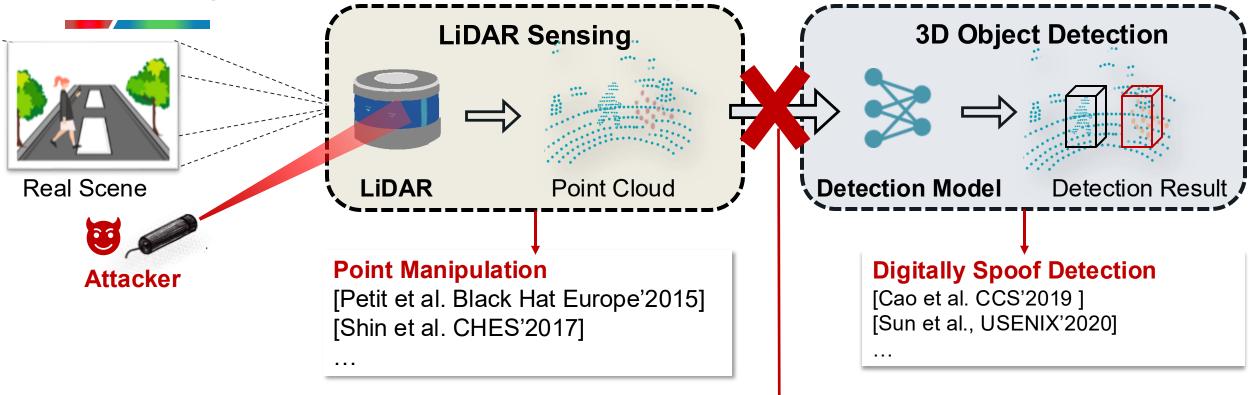
### How Does LiDAR-based Perception work?







Security of LiDAR-based Perception



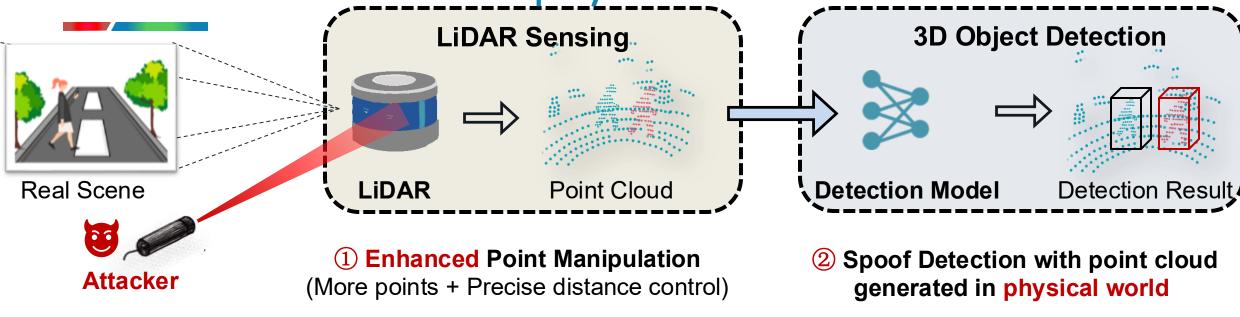


Is it possible to fool 3D object detection using lasers in the physical world?





PLA-LiDAR can achieve physical laser attacks!



### **Contributions of PLA-LiDAR:**

- Enhanced Point Manipulation Capability
- 2. Physical-world Laser Attacks against 3D Object Detection

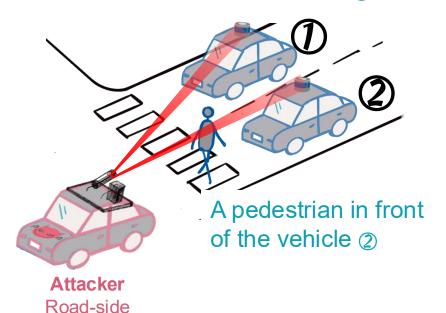


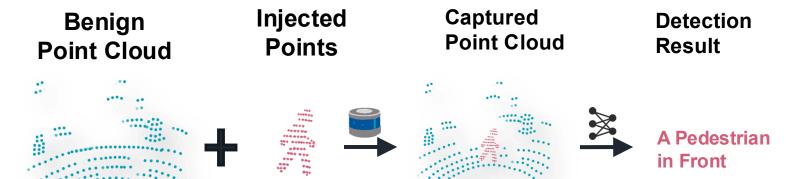


### Threat Model and Attack Goal

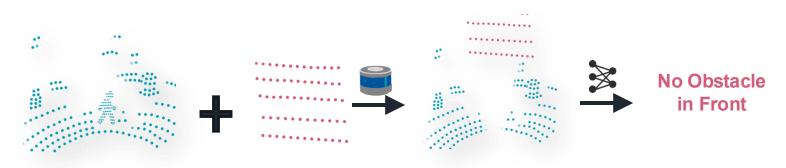
#### **Attack Scenarios**

No obstacles in front of the vehicle ①





**①** Creating: the victim LiDAR perceives a non-existing object

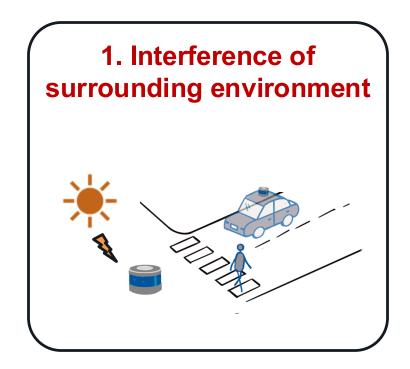


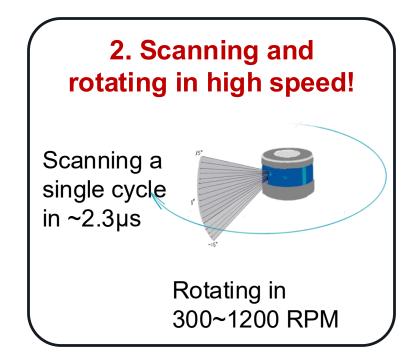
② Hiding: the victim LiDAR fails to perceive an existing object





# Challenges of Physical Laser Attack





```
3. "Curse of Light Speed"

small time error * light speed = distance error

e. g. 1ns * 3 * 10^8 m/_S = 30cm
```

C1: How to make the attack signal be considered as a valid echo?

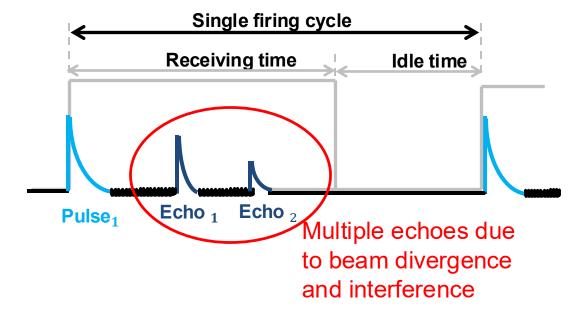
C2: How to have a fine control of the injected point clouds?



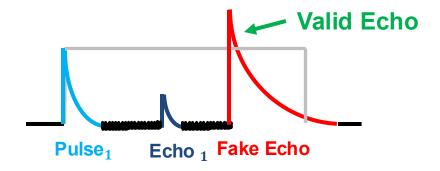


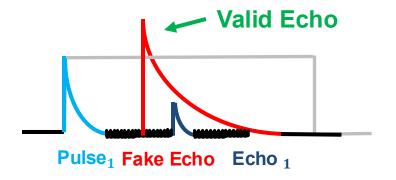
### C1: What is valid echo?

#### □ Return Mode:



Strongest — the strongest echo. ✓
Last —the last (temporally) detected echo.
Dual —both the strongest and last echo.





Vulnerability: Echo can be forged.

### **Insights:**

- A high-power fake echo can be recognized as a valid echo.
- The real echo will be **ignored**.

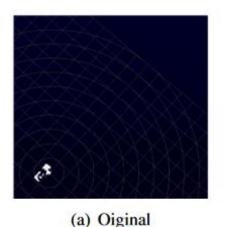


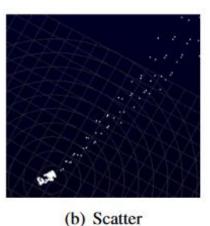


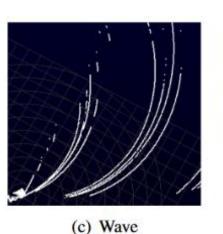
## Preliminary: Point Injection Experiment

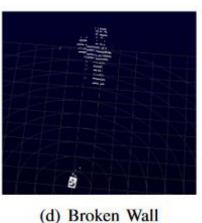
#### **Critical Laser Parameters:**

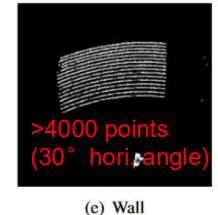
Parameters	Requirements	
Wavelength	Appropriate: Same as LiDAR laser	
Peak Power Power Ppeak = [	High: The higher the better	
<b>Reptition Frequency</b> $f_{rep} = [0 \text{ to } ]$	Precise: According to LiDAR scanning sequence	











SOTA: Cao. CCS'19. 200 points (8° hori. angle)





### C2. Have a fine control of the point clouds: PLA-LiDAR

#### **Step1. Point Cloud Design**

- Design the point cloud to either hide or create objects.
- Requirement: Injectable point cloud.

#### Step2. Laser Signal Design

- Convert the point cloud into laser signals
- Requirement: Fully the 3D information of the point cloud.

#### **Step3. Points Injection**

- Inject the signals into the lidar to generate desired point cloud.
- Requirement: Precise synchronization.

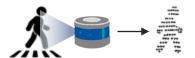


### Step1: Injectable Point Cloud Design

#### Physical Constraints :

- a. Every generated point can only locate on one of the LiDAR's laser rays;
- b. Each laser ray can generate at most one point.

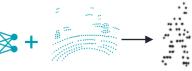
#### 1. Record-based





Pros: 1) Black box. (a) 2) Naturally satisfy the physical constraint.

# 2. Optimization-based ≽+ →



#### **Question formulation:**

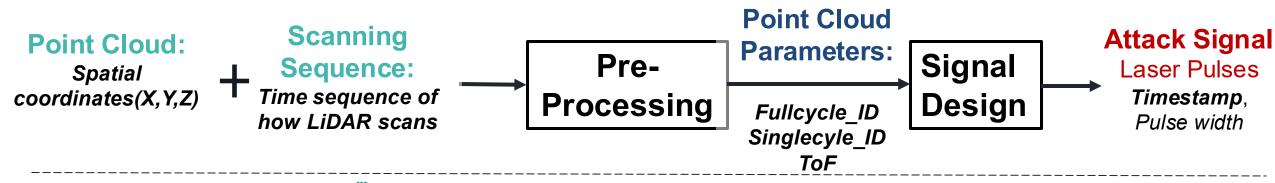
Constrained **Search Space.** 

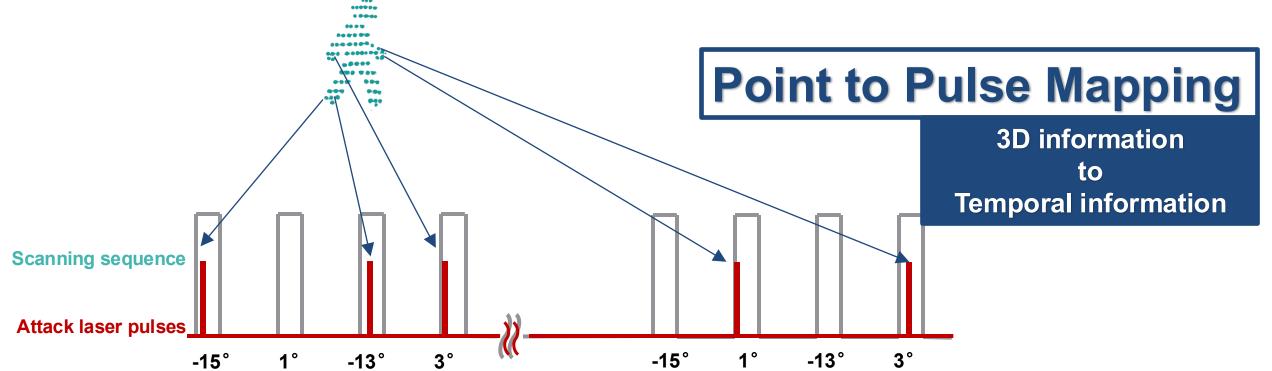
$$\begin{array}{ll} \underset{\mathbf{P'}}{\min} & \mathcal{L}(\mathbf{P} \ ) \\ \text{s.t.} & (R_i^{'},\alpha_i^{'},\omega_i^{'}) \in \operatorname{Loc}^{exp}, i \in [1,n] \\ & |\alpha_i^{'}-\alpha_j^{'}| + |\omega_i^{'}-\omega_j^{'}| \neq 0, i,j \in [1,n] \end{array} \longrightarrow \text{Satisfy constrain a.}$$





### Step2: Laser Signal Design.

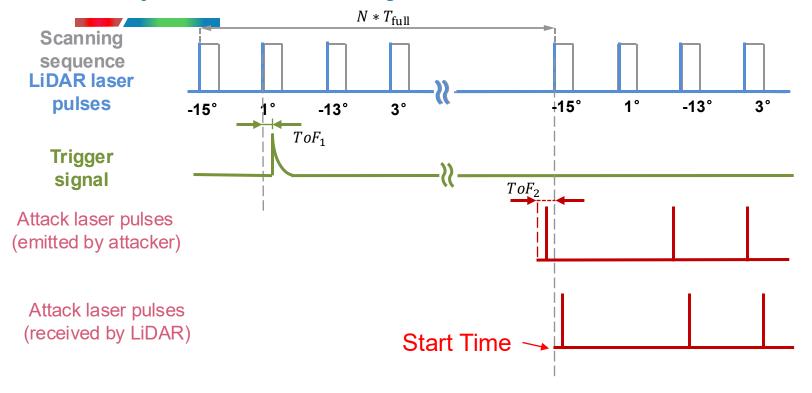


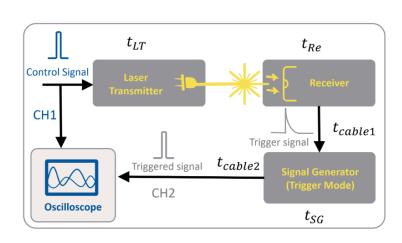






### Step3: Points Injection with Precise Sychronization.





The measurement method of  $t_{device}$ 

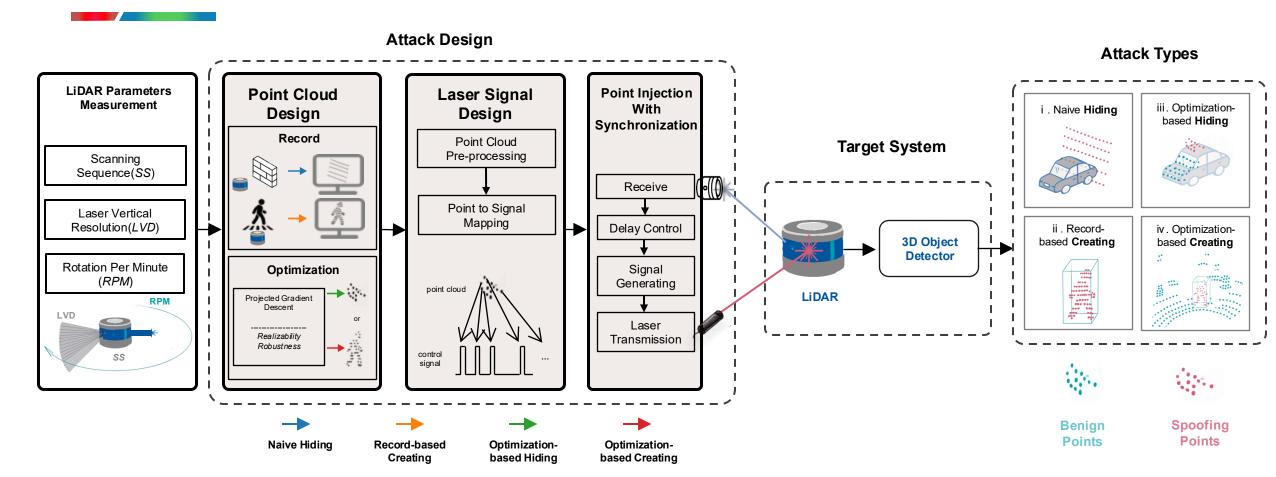
Empirically, time error should be within 3 nanosecond.

The delay should be set:  $t_{delay} = t_{align} - ToF_1 - ToF_2 - t_{device}$  where  $t_{align} = N * T_{full} - T_{sfc}$ 





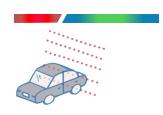
# PLA-LiDAR: System Design.







### **Physical-World Attacks**



**Naïve Hiding** 

#### **Ground Truth**



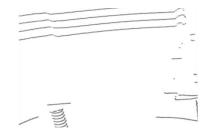
Benign point cloud



Point Cloud under attack



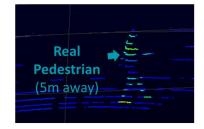
Detection under attack

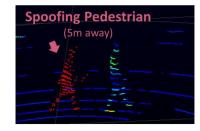


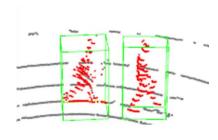


Record-based Creating





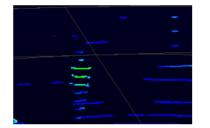


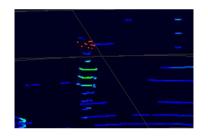


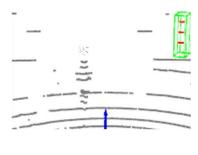


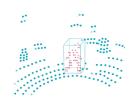
Optimization Hiding





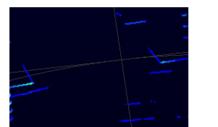


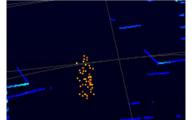


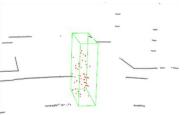


Optimization Creating













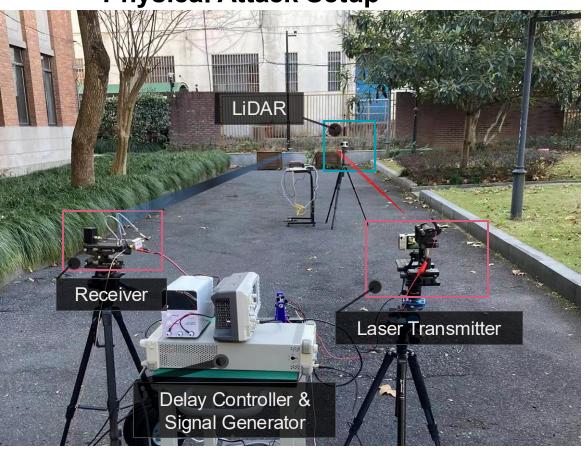
### **Evaluation**

- Physical-world attack Evaluation
  - Four Attacks
  - Two LiDARs: VLP-16, RS-16



 Three Models: Second, Pointpillar, Apollo

#### **Physical Attack Setup**







## Physical-World Attacks

Overall Performance

90.5% 48.8%

Detector LiDAR		Attack Types			
Detector	Model	Nai-Hide	Rec-Create	Opt-Hide	Opt-Create
SECOND	VLP-16	100%	98%	38%	72%
SECOND	RS-16	100%	86%	33%	61%
PoinPillar	VLP-16	100%	64%	79%	15%
1 omi mai	RS-16	100%	51%	68%	12%
Apollo	VLP-16	100%	98%	77%	37%
	RS-16	100%	89%	73%	21%

VLP-16: 73.2%

RS-16: 66.2%

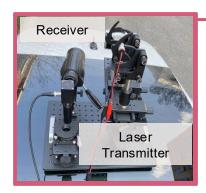
#### **Obersavations:**

- 1. VLP-16 (73.2%) is more vulnerable than RS-16 (66.2%).
  - Period randomization can mitigates our attacks.
- 2. Naive attack (90.5%) is better than optimization-based attack (48.8%).
  - Optimization attack has a higher requirement on timing.

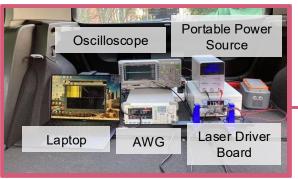




# Feasibility Study on Moving Vehicle



(a) A receiver and a laser transmitter on the car roof.



(b) The attack equipment in the car trunk.



Setup of moving experiment.



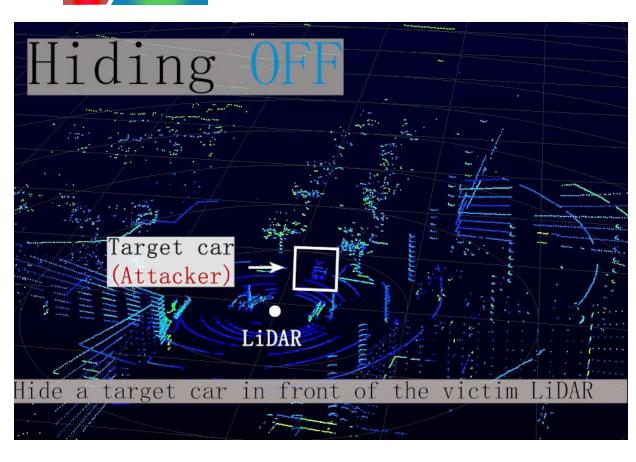
(c) The victim car (Apollo KiT) with a VLP-16.

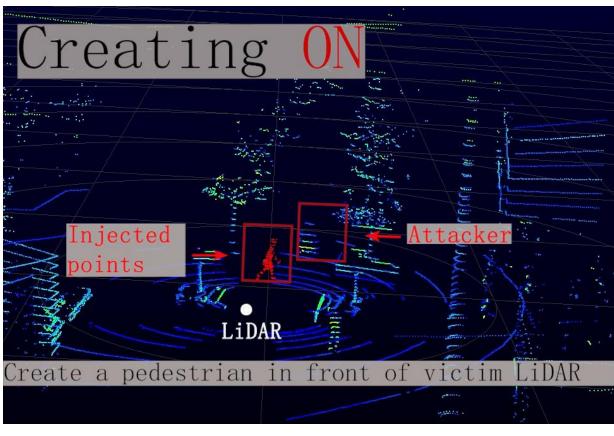
- Improvement to conquer Jitter when moving
- $\triangleright$  A large-diameter telescope ( $\Phi$  = 50 mm) to expand the receiver's receiving area
- > Large spot diameter (8 cm), and use a high-power laser diode (P<sub>peak</sub> = 300 W)
- Attack success rate:
- Hiding Attacks 94.1% (16/17 trials)
- Creating Attacks 78.9% ASR (15/19 trials)





# Feasibility Study on Moving Vehicle - Demo





Hiding Attacks

**Creating Attacks** 





# **Potential Mitigation**

### 1. LiDAR Improvement

- Pulse Encoding
- Pulse Randomizing and Scanning Period Randomizing

### 2. Security Redundancy

- Multi-LiDAR Fusion
- Multi-Sensor Fusion





# **Summary**

- Proposed the PLA-LiDAR attack
  - > 20 times more points than prior works.
  - > 4 types of attacks.
- Extensive physical experiments
  - > 2 LiDARs + 3 Detection Models
  - Stationary + Moving
- Show the physical threats of lasers against LiDAR-based object detection!





# PLA-LiDAR: Physical Laser Attacks against LiDAR-based 3D Object Detection in Autonomous Vehicle













#### **Demo Website:**

https://sites.google.com/view/physicallidar-attack



xji@zju.edu.cn, wyxu@zju.edu.cn



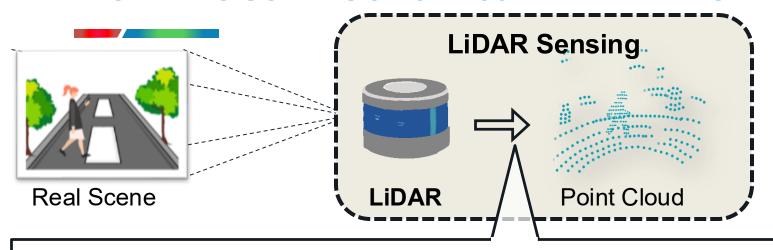


**USSLAB Website:** www.usslab.org

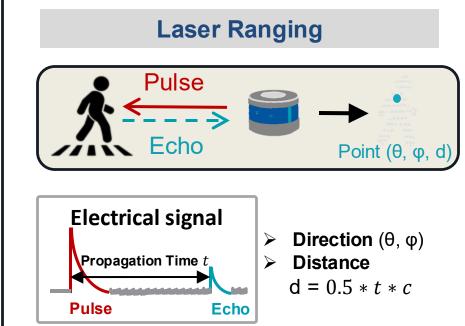


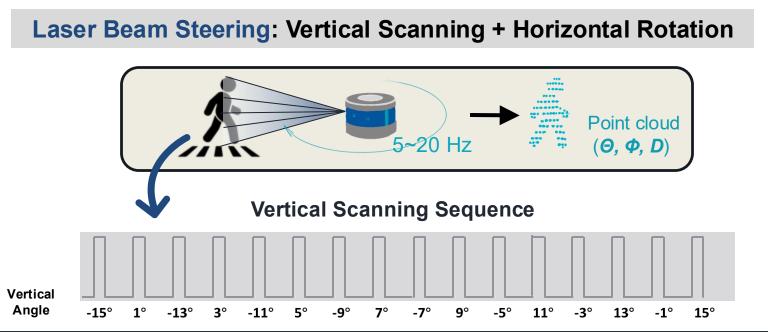


### How Does Mechanical LiDAR work?



LiDAR generates point cloud by laser ranging and laser beam steering.

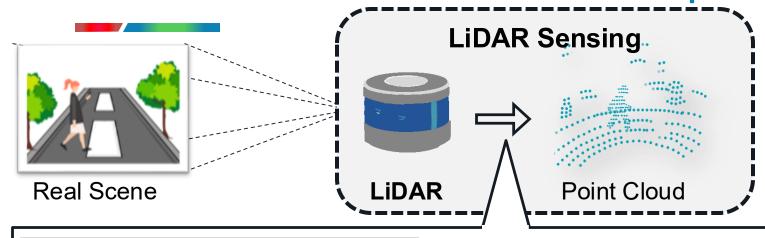






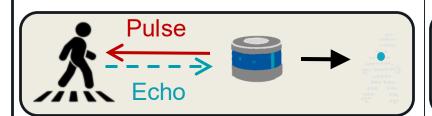


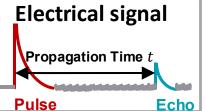
### How Does LiDAR-based Perception work?





Laser Ranging

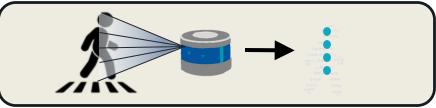




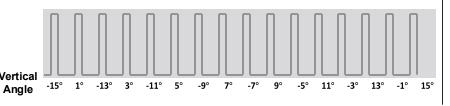
- $\triangleright$  Direction (θ, φ)
- > Distance
- d = 0.5 \* t \* c

#### Point Array:

**Vertical Steering** 

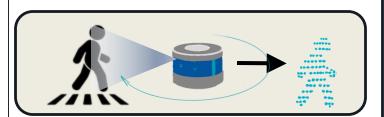


#### Scanning sequence in vertical



#### **Point Cloud:**

Vertical & Horizontal Steering



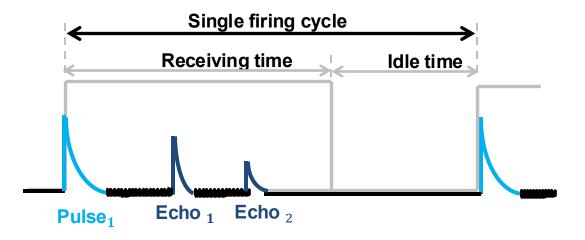
300 ~ 1200 RPM (Rotating per Minute)



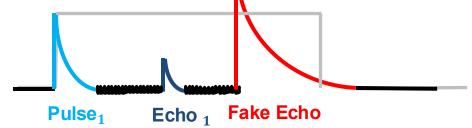


### Dive into Mechanical LiDAR - Return Mode

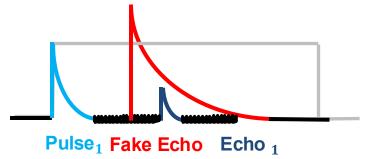
☐ Return Mode: Strongest or Last or Dual



Return Mode	Valid Echo	Point Number
Strongest (Default)	Echo <sub>1</sub>	1
Last	Echo <sub>2</sub>	1
Dual	Echo <sub>1</sub> & Echo <sub>2</sub>	2



Return Mode	Valid Echo
Strongest	Fake Echo
Last	Fake Echo
Dual	Echo 1 & Fake Echo



Return Mode	Valid Echo
Strongest	Fake Echo
Last	Echo <sub>1</sub> / Fake Echo (saturation)
Dual	Echo 1 & Fake Echo