

Multi-Sensor Fusion and V2V Collaboration in Autonomous Vehicles: A Comprehensive Framework for Self-Driving

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Background

- Tesla, Waymo, OpenPilot
- We use a robotic vehicle for testing/research purposes.



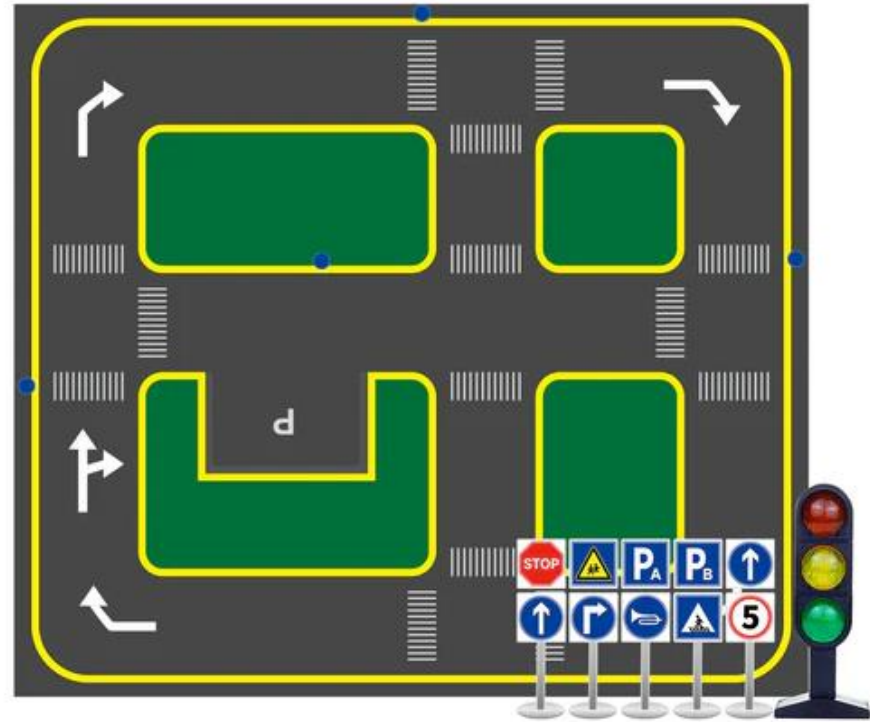
Motivation

- Semi-autonomous driving
 - 1) Lane Detection
 - 2) Vehicle Detection and Tracking
 - 3) Traffic Sign Detection
 - 4) Vehicle-to-Vehicle Collaboration
 - 5) 3D Virtualization



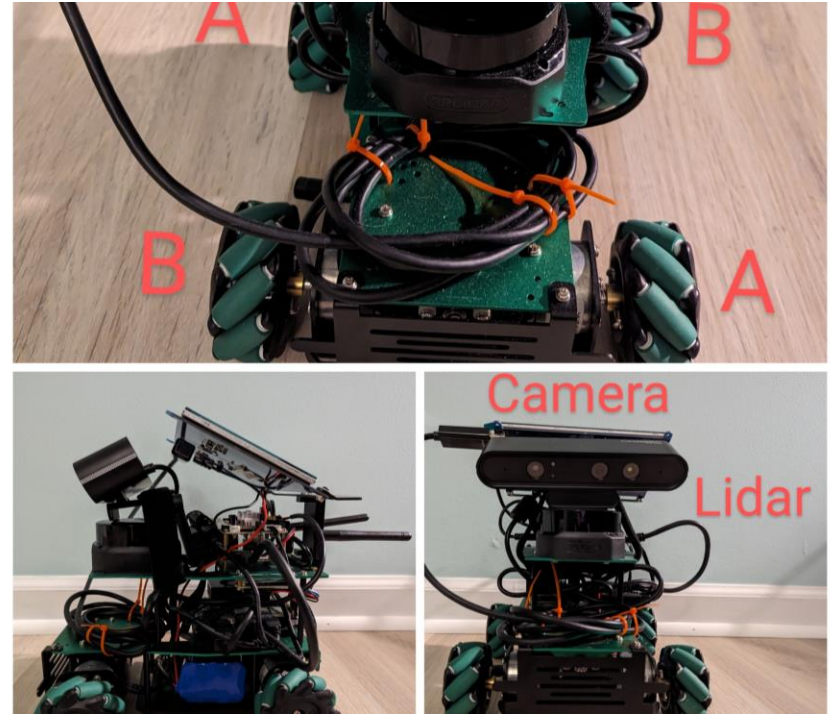
Map

Robotic Map Used
2.8m x 3.2m



Robot

- Robot Used
- Wheels are labeled according to their direction



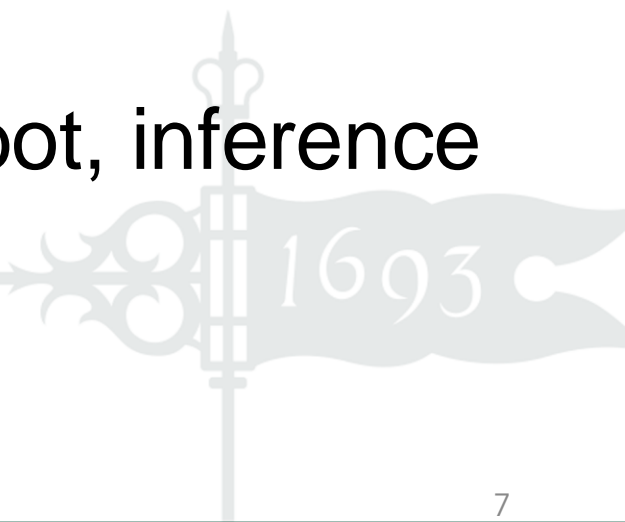
Experiment and Framework Overview

- Lane Detection
- Vehicle Detection and Tracking
- Traffic Sign Detection
- Vehicle-to-Vehicle Collaboration
- 3D Visualization

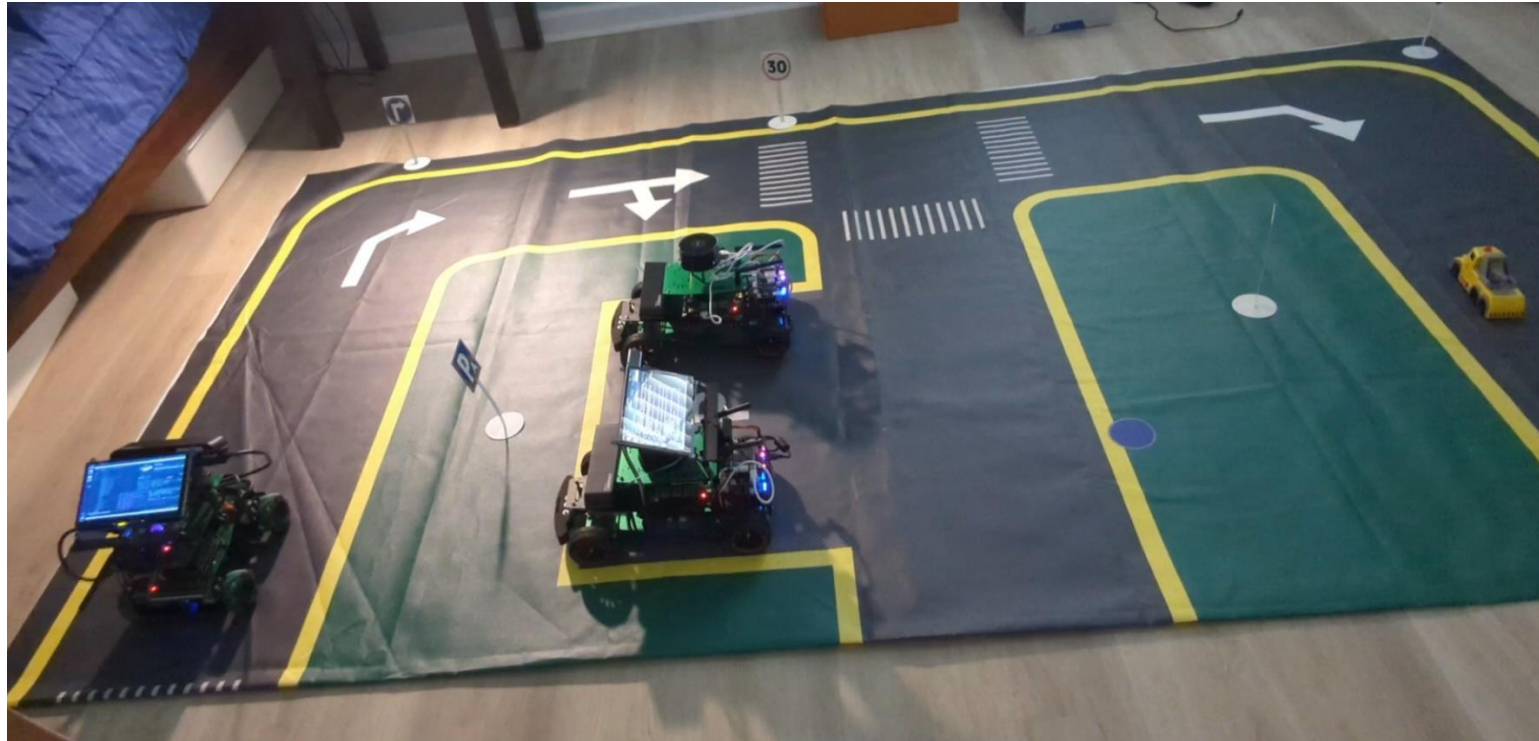


Lane Detection

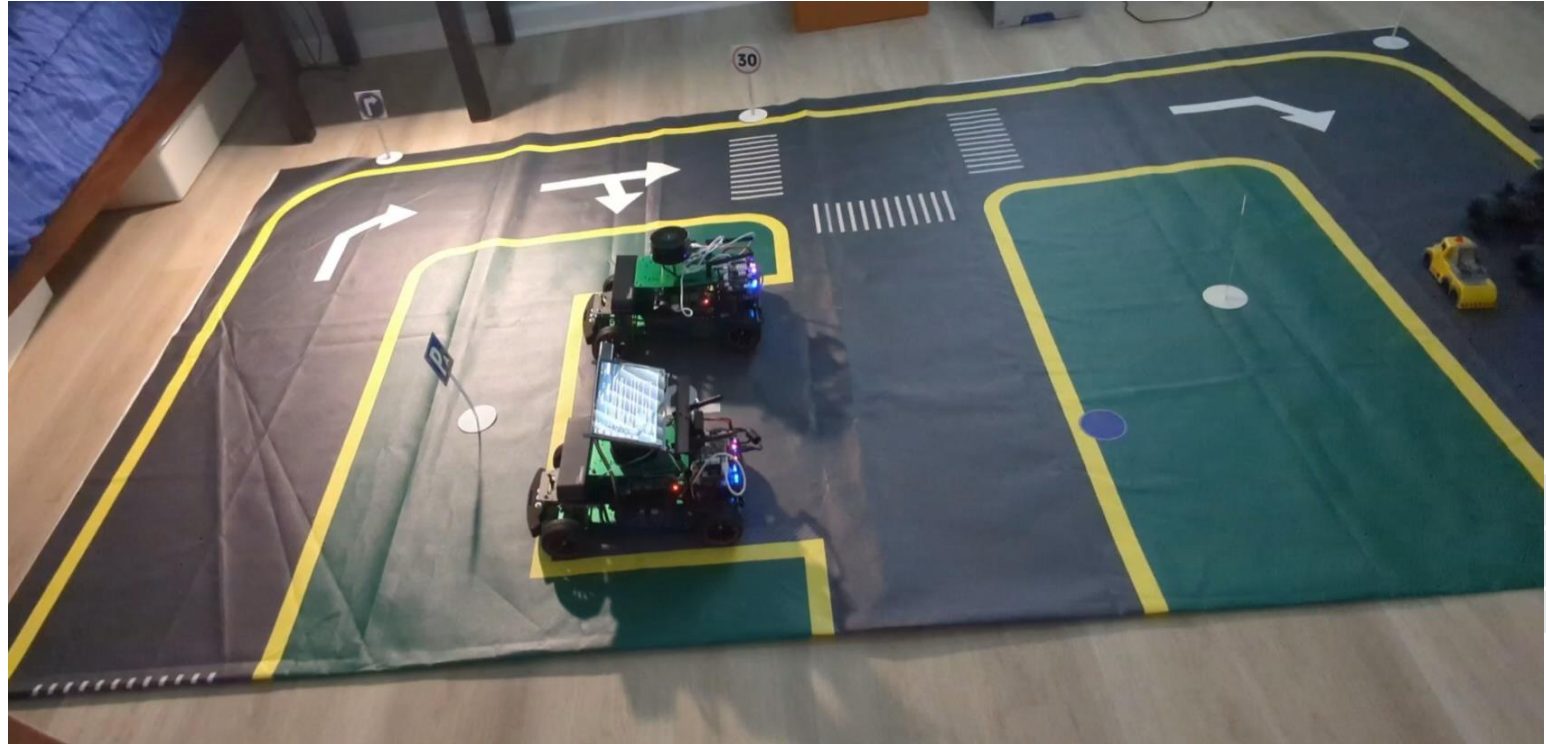
- Recorded vehicle movement around map
- EfficientNet CNN + Long Short-Term Memory (LSTM)
- Trained model is saved to robot, inference is performed on live video



Lane Detection Demo Video

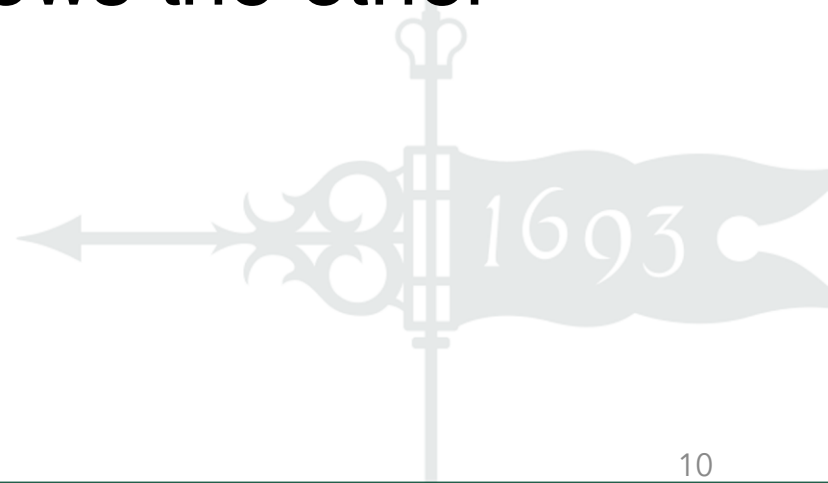


Lane Detection Demo Video

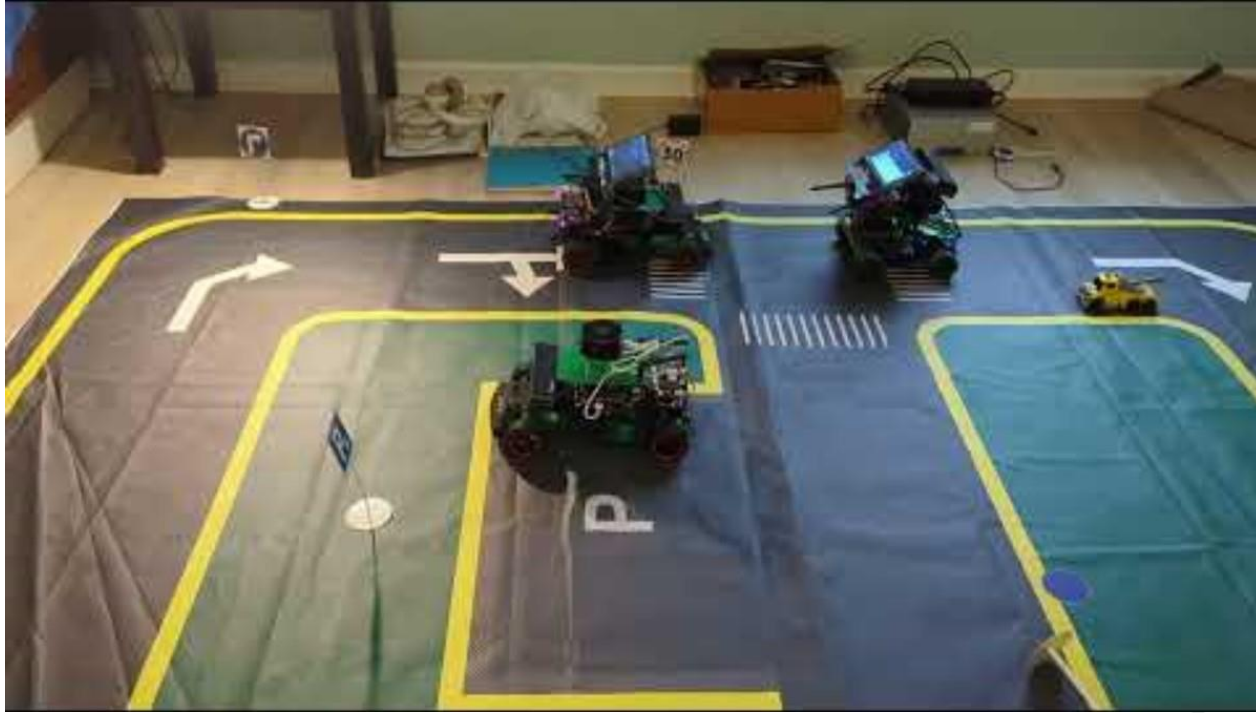


Vehicle Detection and Tracking

- LiDAR sensor for distance calculations/object detection
- Robot identifies and follows the other robot/vehicle



Vehicle Tracking Demo Video



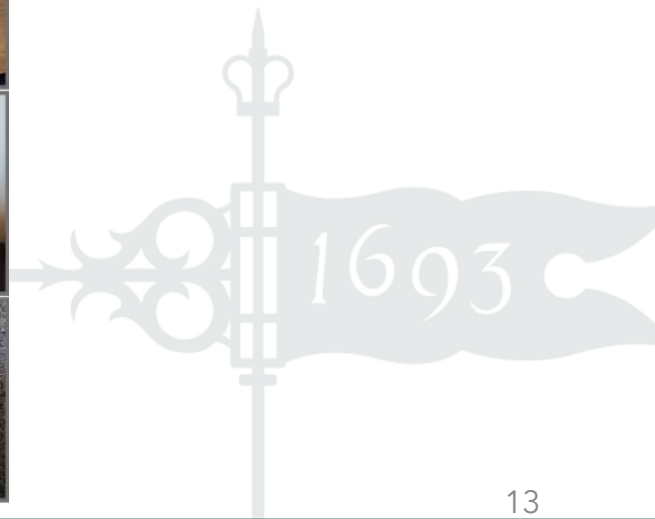
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Traffic Sign Detection

- You Only Look Once (YOLO)
- Augmented images for training
- Robot checks if object is detected, if so open perform call the model



Traffic Sign Detection



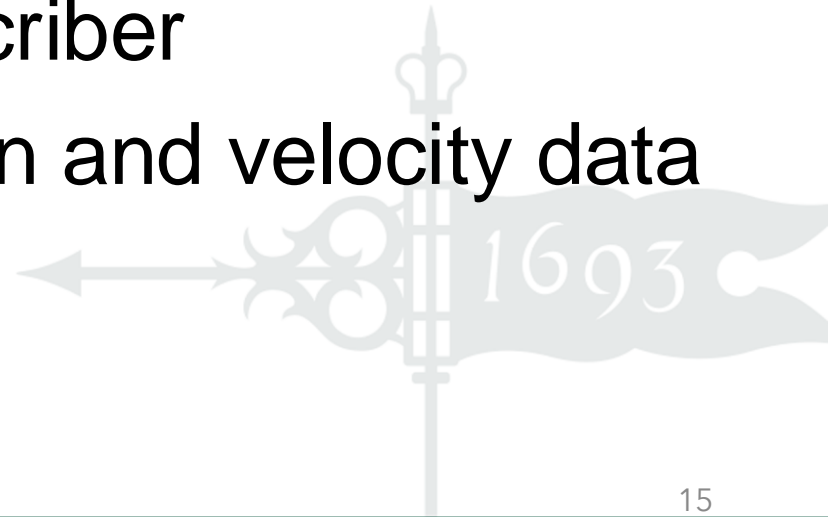
Traffic Sign Demo Video



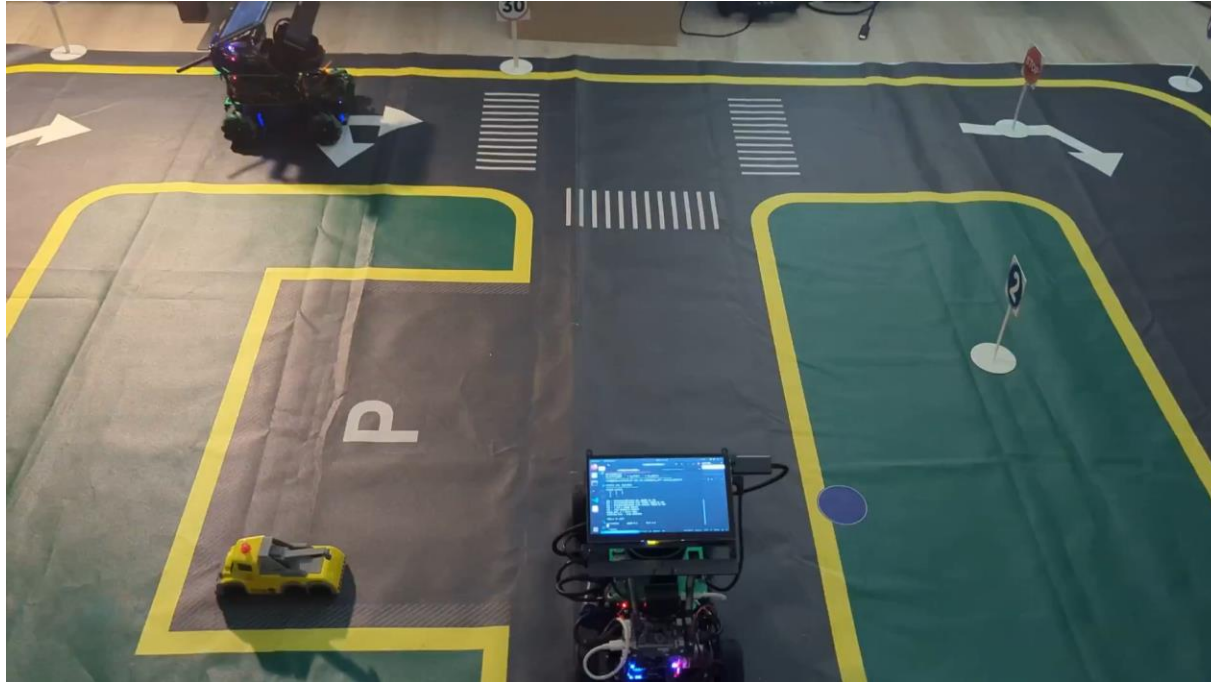
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Vehicle-to-Vehicle Collaboration

- Data Distributed Services (DDS) used
- Robot 1 creates a publisher
- Robot 2 creates a subscriber
- Send each other position and velocity data



Vehicle-to-Vehicle Collaboration Demo Video



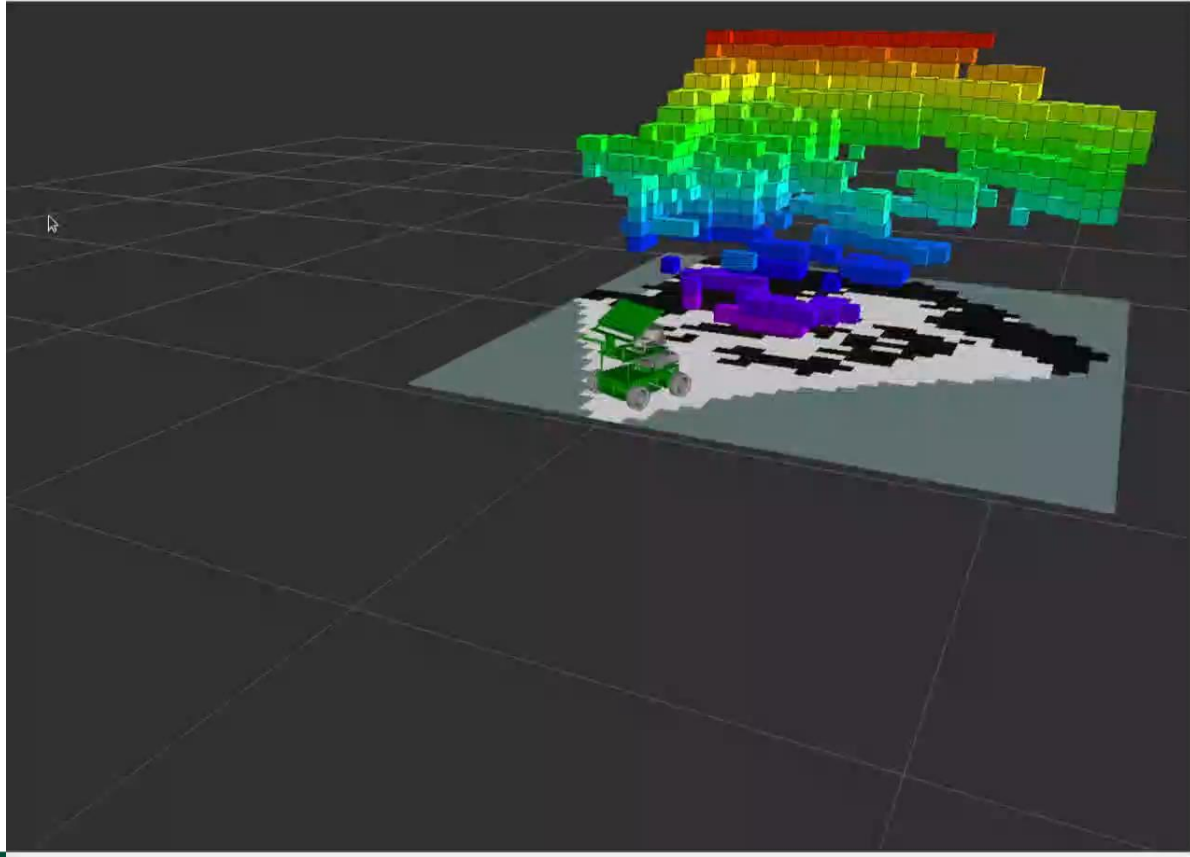
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3D Virtualization

- LiDAR sensor + camera
- Display objects in RVIZ
- Uses Octomapping



3D Virtualization Demo Video

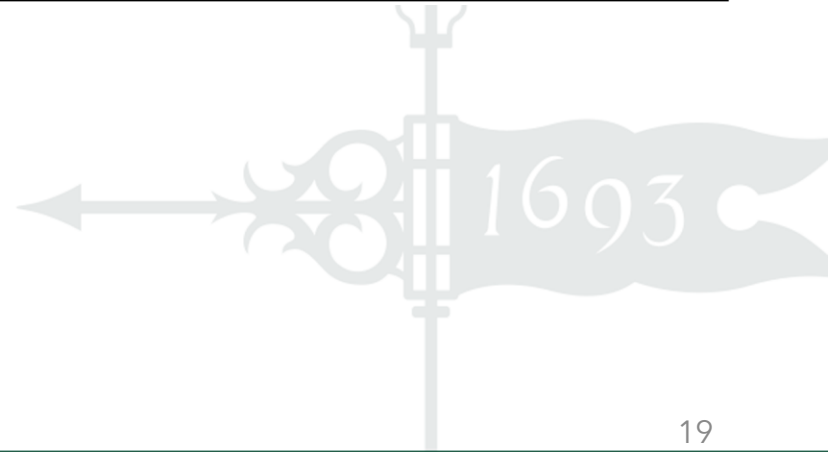


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Traffic Sign Detection Results

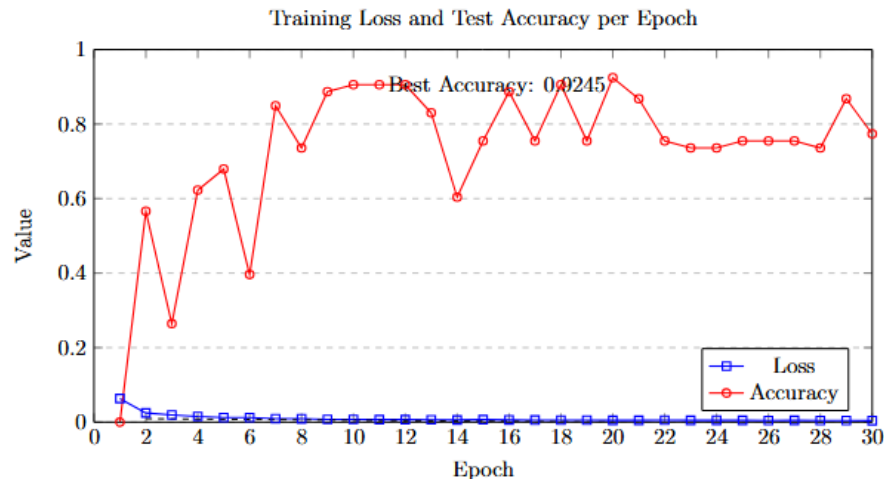
- YOLO model validation results
- Overall mAP of 94%

| Class | Images | Instances | P | R | mAP@50 | mAP@50-95 |
|--------|--------|-----------|-------|-------|--------|-----------|
| all | 50 | 57 | 0.982 | 0.833 | 0.945 | 0.766 |
| not_30 | 50 | 9 | 1.000 | 0.982 | 0.995 | 0.803 |
| 30 | 50 | 8 | 0.985 | 1.000 | 0.995 | 0.819 |
| right | 50 | 9 | 0.965 | 0.667 | 0.870 | 0.685 |
| 2 | 50 | 8 | 1.000 | 0.531 | 0.900 | 0.682 |
| 1 | 50 | 7 | 0.951 | 1.000 | 0.995 | 0.788 |
| alarm | 50 | 8 | 1.000 | 0.774 | 0.949 | 0.807 |
| left | 50 | 8 | 0.971 | 0.875 | 0.913 | 0.778 |



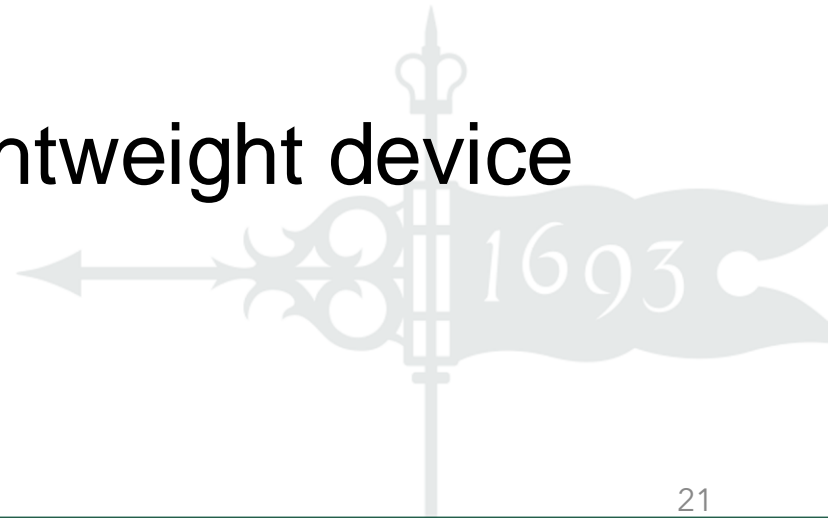
Lane Detection Results

- Best test accuracy was reached at Epoch 20
- 92.45%
- Early stopped called at Epoch 30



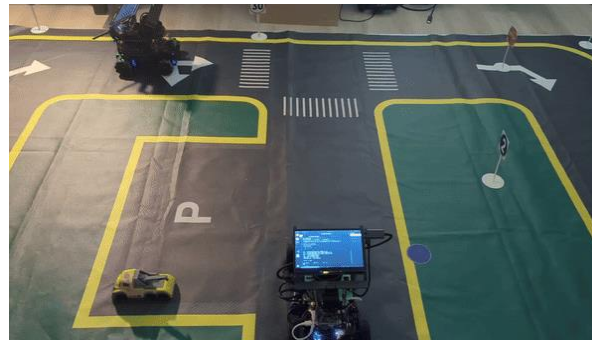
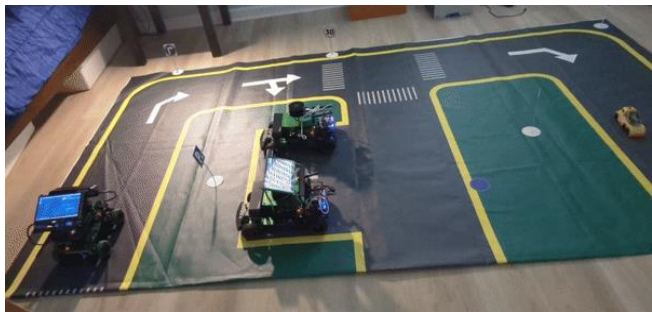
Challenges - Solutions

- OpenCV Incorrect Lane Detection
- Limitations of OpenPilot's Model
- Robot Turning Issue
- Optimizing Code for Lightweight device

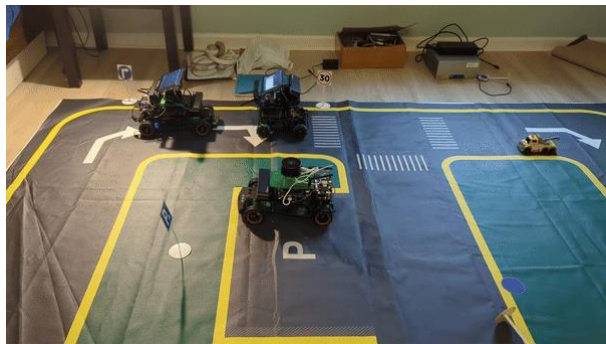


Overview/Summary

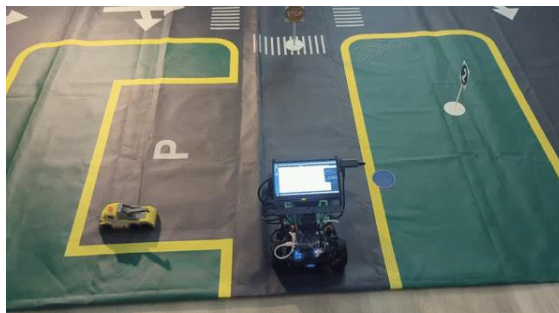
Lane Detection



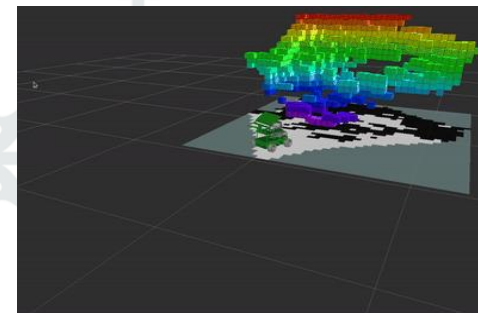
Multi-Robot
communication



Vehicle Tracking

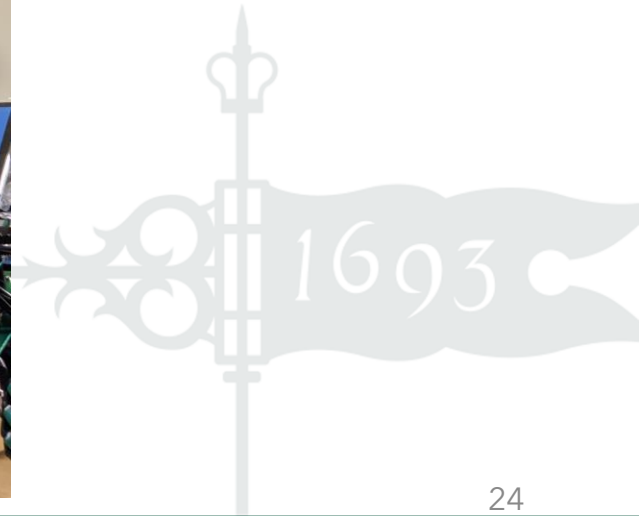
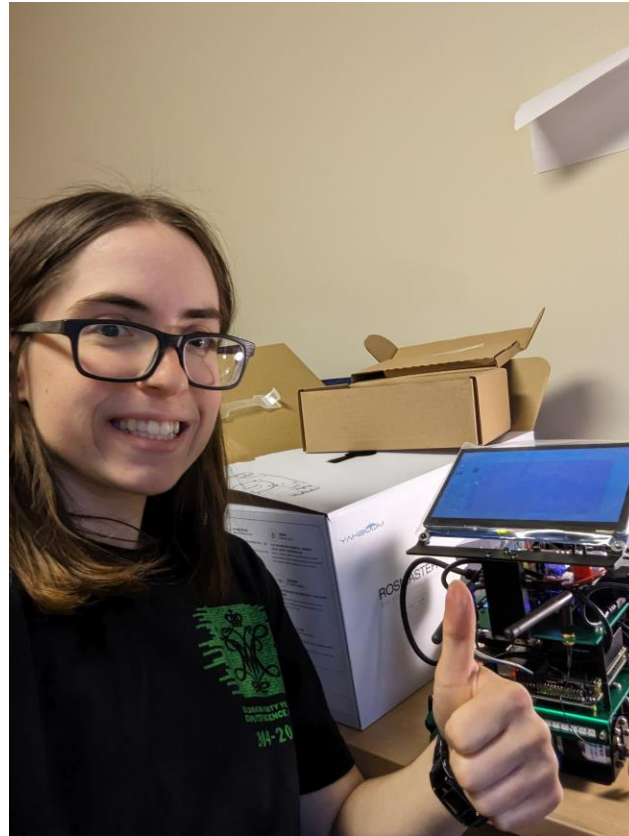


Sign Detection



3D Visualization

Thank You!



GitHub

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