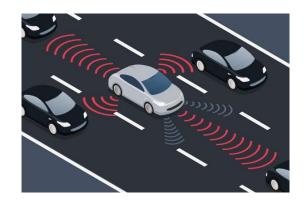
Deep Learning-Based Autonomous Navigation

Mahi, Jordan, James, Bhavik, Ryan Wednesday May 15th, 2019 RSS 2019 Team 11

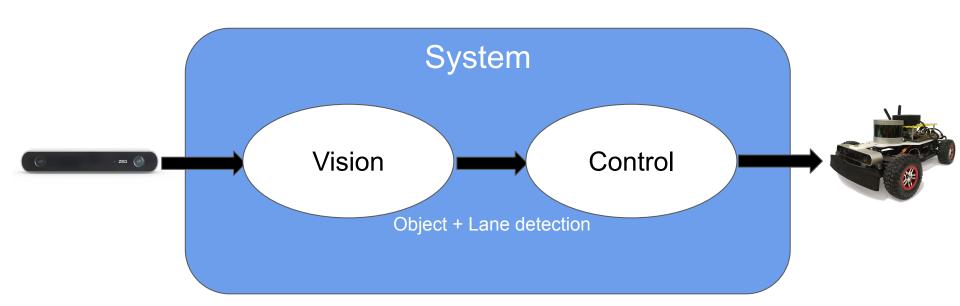
Motivation: Intelligent Autonomous Navigation

- Robust autonomous driving has massive societal implications:
 - Safer, more efficient transportation
 - Greater mobility
 - Increased productivity
- Autonomous driving requires robust ability to:
 - Understand the environment
 - Respond with well-developed controls





Our vision module informs our control module



We trained object detection CNN on custom datasets

YOLOv3 Architecture

- Deep Convolutional Neural Network
- Full Model: 252 layers
- Tiny Model: 44 layers

• Training:

- 1600 annotated custom images from Zed Camera
- Keras + Tensorflow environment on Amazon EC2 Instance
- 2-stage: first stage trains last 3 layers, second stage trains all layers

8 Custom Classes

- 4 signs: disabled parking, pedestrian, yield, and stop signs
- car, person, traffic light, soccer ball (obstacle)

The YOLO Pipeline



Real-time Object Detection

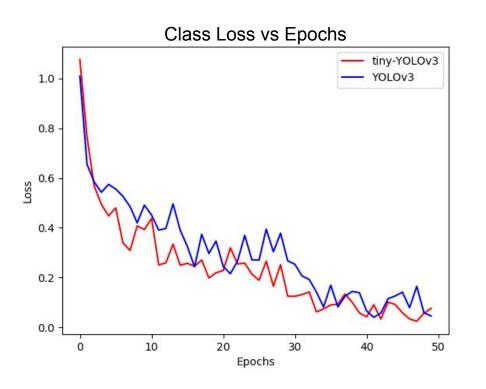


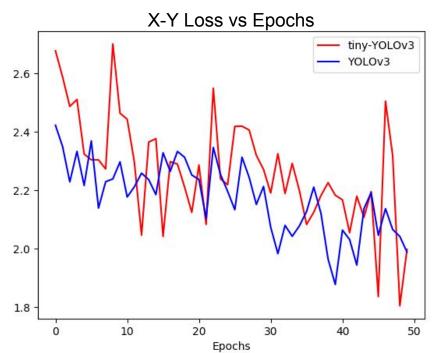
Detection of Disabled Parking Sign



Detection of Pedestrian Sign

Tiny vs Full YOLOv3 Validation Error





Vision: Lane Segmentation

Original Image

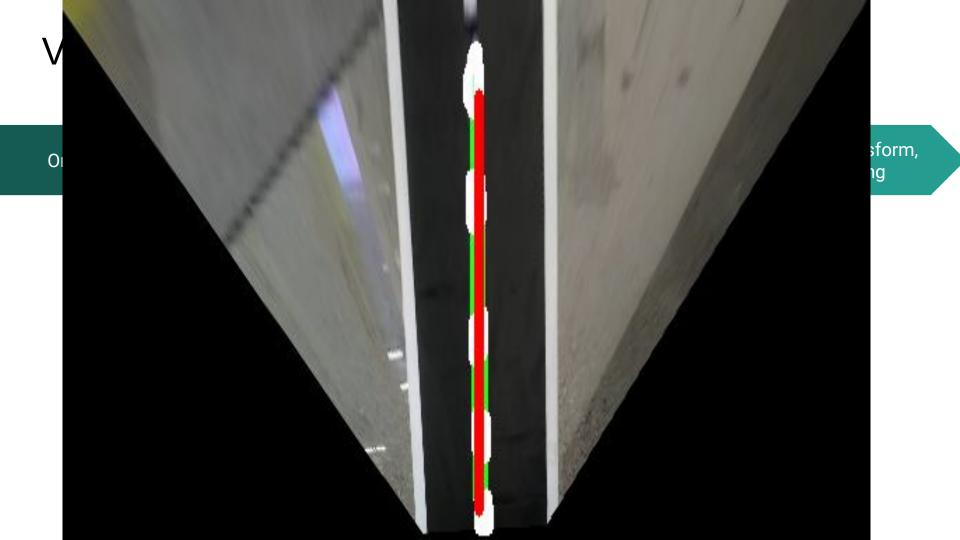
Perspective Warping

Segmentation

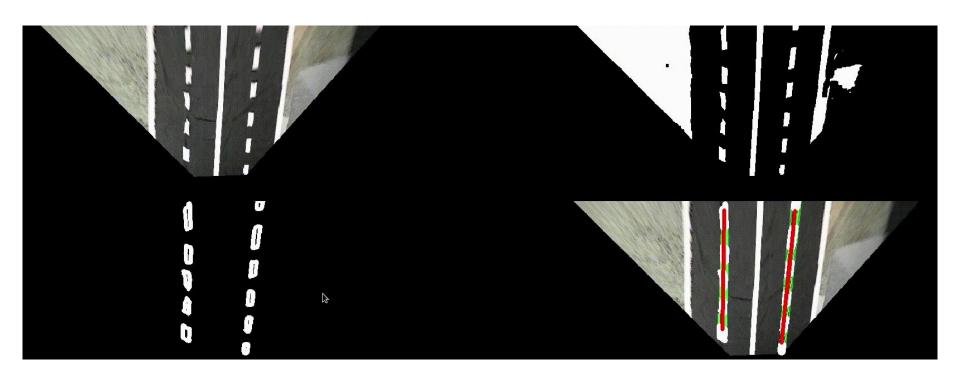
Contours Filtering

Clustering

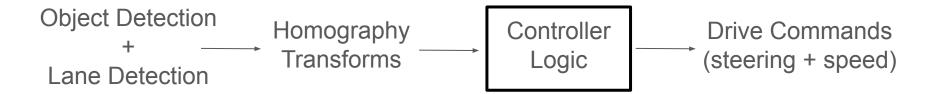
Clustering



Lane Segmentation in Real Time



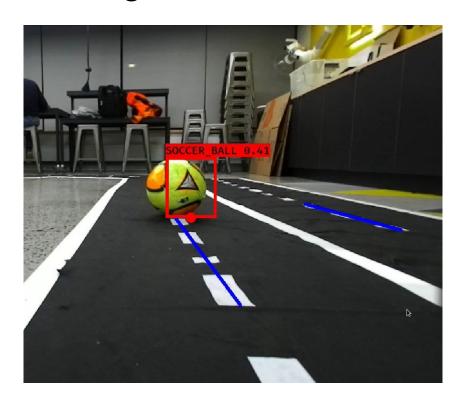
Controller: High Level



Three states:

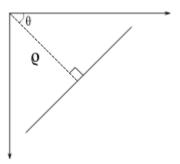
- 1) **Autonomous Driving**: lane changing/following, turning, stopping, traffic rules, etc.
- 2) **TA car following**: PD control based on relative distance/location from TA car
- 3) **Parking**: valid parking space detection + PD control

Steering controller finds the best unobstructed lane



- 1) Check for obstacles in current lane
- 2) Best lane: smallest difference to lane of previous timestep (in ρ , θ space)

$$\rho = x \cos \theta + y \sin \theta$$



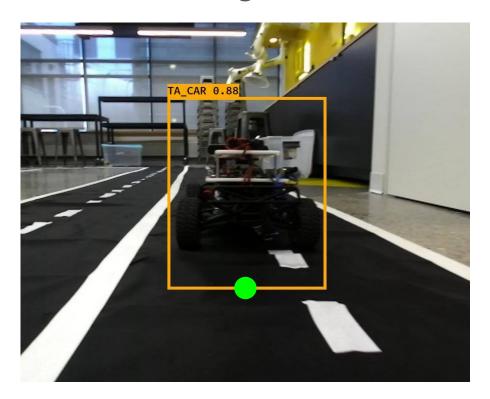
3) Angle calculation by Pure Pursuit



Limitations we faced, and our solutions

Problems		Our solutions
1	Once at an intersection, robot can no longer see the intersection due to the camera's limited field of view	Plan out turns in advanceHardcode drive action(Ideally: more cameras)
2	Object detection does not recognize objects in every frame	 Rely on detected objects in the last frame(s)
3	YOLO often fails to load and crashes other nodes	Wait many hoursStressful hours(Ideally: explore options)

TA car following uses PD controller



PD control on midpoint of bottom of bounding box

Desired distance: 1 meter

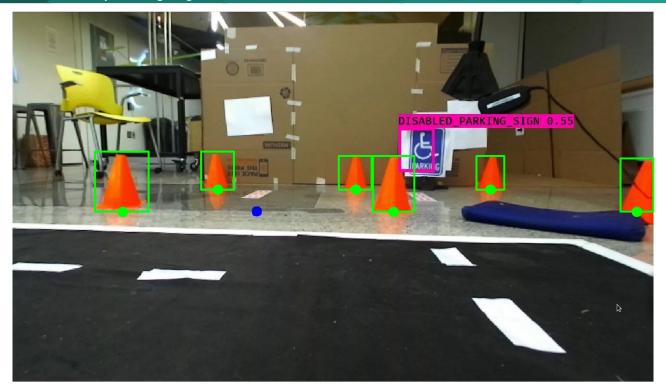
Parking controller uses color segmentation and YOLO

Color segment + bound orange cones

YOLO on disabled parking sign Stage 1: Find spot with front 3 cones

Stage 2: Move to midpoint of back 2 cones

Stage 3: Stop when < 2 cones





What We Learned

Object detection methods:

- YOLOv3 Deep Convolutional Neural Networks
- Color segmentation

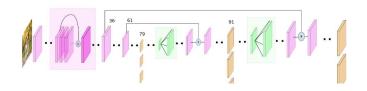
Lane detection methods:

- Hough Transform
- Edge Detection
- Clustering

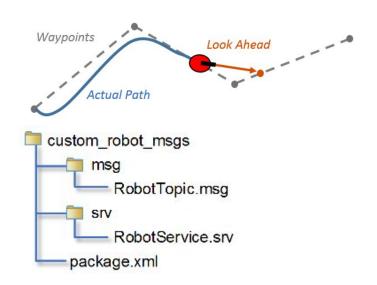
Applications of pure pursuit control:

- Parking
- Turning
- Lane following

Developing our own ROS package







What We Can Improve on

Object detection

- More data (e.g. lighting and placement variations) to decrease overfitting
- Other network architecture for higher inference speed

Lane detection

More robust algorithms, especially around the corners

Controller

Higher driving speed for greater practicality

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With thanks to Marwa Abdulhai, Muyan Lin, Prof. Carlone, Prof. Karaman, and the rest of the RSS Staff!





