Current Progress of V2X Research

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An Overview

Questions and Motivations:

- Efficient traffic management
- Viable and scalable V2X model

How We've Been Answering Them:

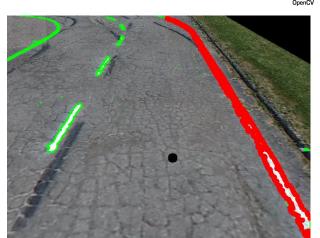
- Creating reliable lane-following algorithm(s)
- Developing viable intersection detection
- Prototyping light-RSU-vehicle connection
- Establishing a modular node architecture
- Simulating V2X capabilities with gazelle



Lane Detection and Following

Tested Approaches on the ACTors:

1) Largest White Contour With an Offset.

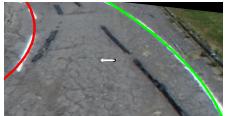


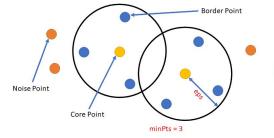
2) Density Based Clustering (Unsupervised Learning)



















Lane Following Demo



Notable Characteristics:

- Successfully lanekeeps in extreme environments (shown: shadow, glare)
- Centered properly within the lane
- Speed retention on turns
- Minimal jerk on straight drives

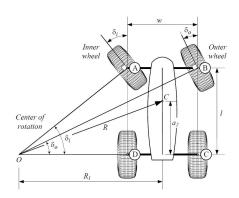
Lane Following

Achievements:

- 3 laps in both inner and outer lanes using each algorithm
- Improved the advanced clustering approach by introducing DBW Native Commands: Actuator Messages and Unified Controller Messages
- IGVC regulatory speed of 5 mph
- Detect the yellow intersection line, stop for a bit, cross the intersection, and ignore the next yellow line

Future Work:

- Birds-eye-view implementation
- Test other approaches: Blob, Hough Lines, and Deep Learning Lane Detection
- Create a table to compare all approaches for the paper
- Choose the best for V2X final demo





Lane Following: Birds-Eye-View

Idea:

- Theoretically, easier curve detection



- First using cv2 function to change the image perspective
- Image processing cropping the image ,threshold, canny
- Gets points of the left and right and computes a least squares function through them
- Computes middle point between left and right lines

Future Improvements

- Tune algorithm to deal with faster speeds and sharp curves
- Connect algorithm to a Deep Learning lane detection





Computing Speed and Detecting Intersections

```
# detect intersection
Create vellow mask
Compute yellow pixel percentage
# lane detection
compute ideal center using CV
if drive.
     # linear speed
     if yellow percentage > threshold:
        Set target speed to 0 for a set time
        Start moving again and ignore yellow detection
        for a set time
     else:
        set target speed to speed limit
     # angular speed
     if mid is left of center by a threshold:
          turn steering wheel left
     elif mid is right of center by a threshold:
        turn steering wheel right
     else:
        go straight
else:
      set target speed to 0
```



Establishing an RSU Strategy: Initial Connection

Our Introduction:

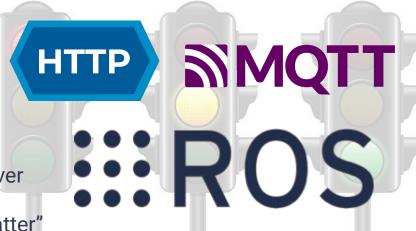
- Understanding sample GPT HTTP Arduino code

The Process:

- Look into alternatives: MQTT, ROSSerial
- Practice with alternative timing algorithms

Where We Are:

- Successfully connected RSU and Arduino over RSU-hosted ROSSerial connection
- Prototyped communication over model "chatter" topic



Establishing an RSU Strategy: Message Types

Traffic light implementation

- The traffic light 'brain' implemented on arduino
- After each switch publishes a messages to topics:
 - Direction, State, time until next change

Custom message type on arduino | Error

Idea: publish a custom ros topic with state and time from arduino to RSU

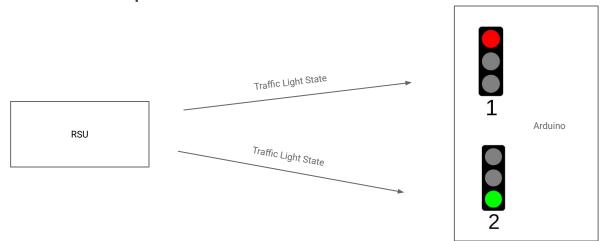
- Made a custom ros topic on ros
- Problem: too much work with installing both ros on windows and arduino on linux
 - Tried WSL, but linux and windows have different file systems
 -> wouldn't be able to connect ros message to arduino
- Decided to come up with another system

```
$ rosmsg info TrafficMessage
[custom_msg_pkg/TrafficMessage]:
bool state
uint16 time
```

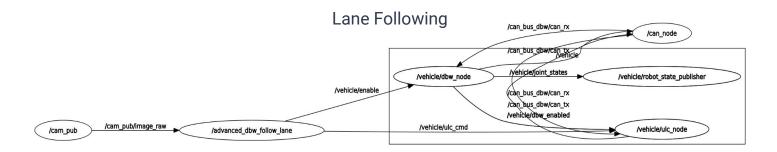
```
state: True
time: 256
state: True
```

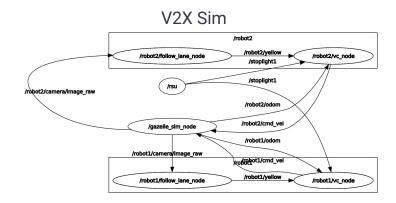
Going Forward: RSU "Host" → Light "Client"

- RSU will run the state and timer of traffic light internally
- RSU only publishes state changes to traffic light node
- Each side of the light will subscribe to topics in that side's ROS namespace

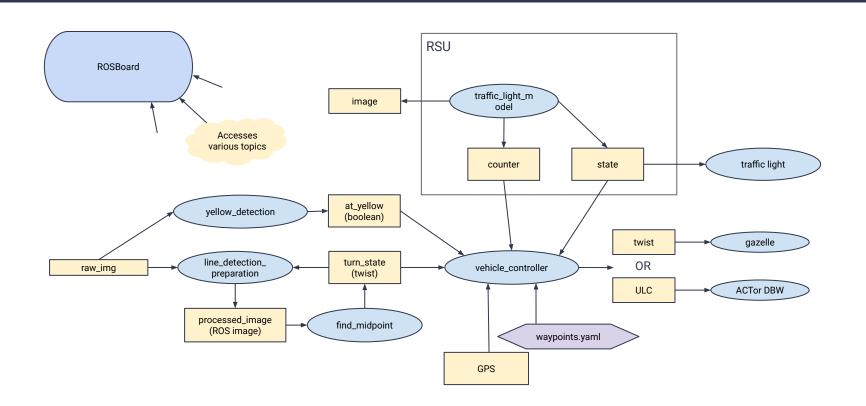


Software Architecture





Conceptualization of Future Architecture



Current Progress on V2X Simulation

- Currently using contour based lane following algorithm
- Developed an adaptive speed algorithm to reduce waiting time at intersections
- Tested algorithm on two different maps with varying number of vehicles
- Created rosboard to visually show stop lights and their times

Capturing Waypoints in Simulation

We use waypoints to solve the "birds eye view" distance calculation problem

Steps:

- 1. Trace expected path of car on simulator map
- Ping position of map on simulator through terminal
- 3. Increment every few centimeters
- 4. Save positions to .yaml file, separating intersection coordinates
- 5. Repeat process for each designated path for car to follow

Improvements:

- Reduce number of waypoints to "checkpoint" type structure on corners only
- Compute a known "distance to intersection" at each node

```
1 intersections:
   - [35.97, 2.14, 0, "E"]
                                #horizontal - E
   - [30.43, -6.20, 1, "N"]
                                #vertical - N
5 waypoints:
   - [30.49, -5.17]
   - [30.43, -4.26]
   - [30.49. -3.34]
   - [30.43, -2.09]
   - [30.37, -0.83]
  - [30.43, 0.43]
   - [30.49, 1.80]
   - [30.60, 2.94]
  - [30.66, 4.26]
  - [30.71, 5.06]
  - [30.71, 6.20]
   - [30.71, 7.11]
  - [30.71.8.14]
   - [30.71. 9.63]
   - [30.66, 10.89]
  - [30.66. 12.09]
  - [30.60, 13.40]
  - [30.71, 14.60]
  - [30.71, 16.03]
  - [30.77, 17.63]
  - [31.06, 19.00]
   - [31.46, 20.14]
  - [32.09, 21.17]
  - [32.89. 22.09]
   - [33.69, 22.83]
   - [34.77, 23.46]
```

Using Waypoints in Simulation

Finding distance using waypoints:

```
Find the closest waypoint to the car's position:
   for every waypoint:
      calculate the distance from car to waypoint
      if distance < lowest distance so far:
        lowest distance
      record the index of lowest distance waypoint</pre>
```

```
Sum the total distance from current waypoint to next intersection:
   total distance = 0
   for every waypoint from waypoints[current] to waypoints[intersection]:
     total distance += distance between waypoint i and waypoint i + 1
```

Adaptive Speed Algorithm

```
when stop lights change state:
      calculate distance to closest intersection
      time to intersection = distance / current speed
      # works good on simulation
      if light state is red:
             if time to intersection < time to change state:
                    set speed to distance / time to change state (arrive too soon, SLOW DOWN)
             else:
                    keep speed
      # logic to improve (weird behaviour sometimes)
      if light state is green:
             if time to change state < time to intersection > and time to next change state:
                    set speed to distance / time to change next state (arrive too late, SLOW DOWN)
             else:
                    keep speed
```

Rosboard

- Initially planned to use Flask
- Then we switched to Rosboard introduced by Devson
- Display stoplight status on the intersection and time left for the state to change

```
# Red at horizontal intersections, green at vertical intersections while (time.time() - curr) < 10:

stop = Bool(data=True)
left_time = Float64(data=10 - (time.time() - curr)) # Time remaining stop, publ.publish(stop)
stop, publ.publish(stop)
stop, publ.publish(stop)
state, pub.publish(statel_msg)
rate.sleep()

curr = time.time()

# Green at horizontal intersections, red at vertical intersections while (time.time() - curr) < 10:
stop = Bool(data=10 - (time.time() - curr)) # Time remaining stop, publ.publish(stop)
stop, publ.publish(stop)
stop, publ.publish(stop)
stop, publ.publish(stop)
stop, publ.publish(stop)
stop, publ.publish(stop)
stop, publ.publish(stateZ_msg)
rate.sleep()
```

```
rospack = rospkg.RosPack()
package_path = rospack.get_path('v2x_sim')
statel_path = f"{package_path}/maps/gr.png"
state2_path = f"{package_path}/maps/rg.png" # Assuming you meant to have different images
# Load images
state1 = cv2.imread(state1_path)
state2 = cv2.imread(state2_path)
```



Lot H Course Sim

Accomplishments:

- Changing light states
- Light states recognized by cars
- Adaptive speed algorithm

Flaws:

- Lane Following
- Light states at runtime
- Not slowing down enough for red lights
- Stopping before crosswalk

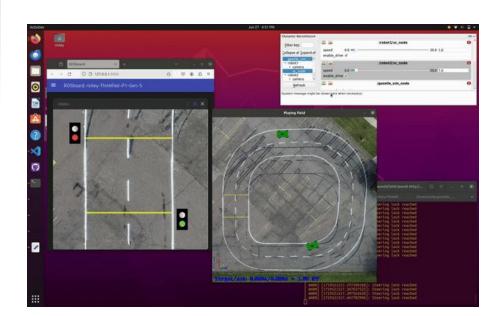


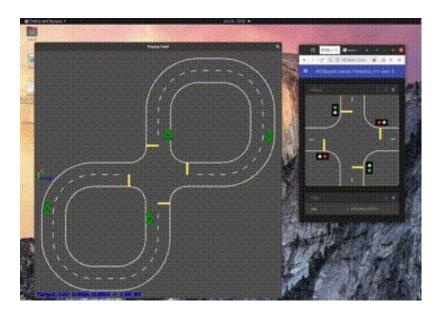
Figure 8 Sim

Accomplishments:

- Changing light states
- Light states recognized by cars
- Adaptive speed algorithm
- Working with 4 vehicles

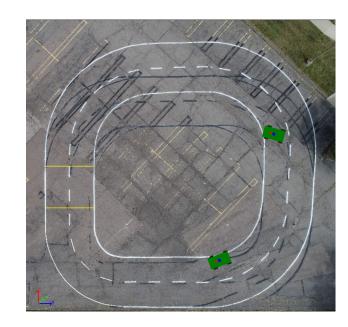
Flaws:

- Lane Following
- Light states at runtime
- Not slowing down enough for red lights
- Stopping before crosswalk



Future Goals for V2X Simulation

- Create a dependable lane following algorithm that works in both simulation and real-life
- Implement a yellow light functionality to avoid really slow speeds
- Adjust adaptive speed algorithm to never stop at red lights
- Begin adapting simulation to real life



Summary and Plans Moving Forward

Summary:

- A robust lane following algorithm developed; viable alternatives on the way
- Proof-of-concept RSU connection established
- Adaptive speed algorithm working within simulation for:
 - Both circle "Lot H" course and figure 8 "intersection" course
 - Red and green lights
 - Up to four cars

Future Plans:

- Modularize codebase to make issues more discrete (single-node architecture likely not maintainable)
- Begin testing all project components (RSU, lane following, adaptive speed algorithm) together outside of the simulation
- Begin working on writing lane following and V2X papers