# Team D8 Traffix 

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## Use Case



## THE PROBLEM

Current traffic lights waste time and fuel because they are not optimized for varying traffic conditions
Existing technologies like induction sensors don't adapt to evolving traffic patterns

## STAKEHOLDERS

Local transportation authorities save on long term costs to optimize traffic
Average commuter saves on time wasted while commuting

## OUR SOLUTION



Design a smart traffic light that continuously optimizes light timings based on car/pedestrian density and flow data Replacement to existing traffic lights
Can be implemented in isolation or at city-wide level

## System Specification



## Quantitative Design Requirements

## DESIGN REQUIREMENT

SPECIFICATION
USE CASE JUSTIFICATION


90\% for cars
$80 \%$ for pedestrians

Avg. wait time reduced $>10 \%$ compared to fixed-time light

Users should feel like light timings reflect actual traffic density
Q.O.L. improvement should be noticeable to drivers + pedestrians

Models can handle a minimum of 10 cars at each side of intersection + complex API data

Product is most useful if it can be used to alleviate high-density traffic

< 5 s total between traffic data input and time interval update

Light changes should accurately reflect the current situation

## Implementation - Object Detection

## Overall Solution

- Run on 4 concurrent videos from each side of Fifth \& Craig intersection
- Detect number of pedestrians and cars in each frame with YOLOv3 model
- Determine lane boundaries in order to output number of cars and pedestrians on each side of the intersection
- Currently using hard-coded coordinates as opposed to an edge detection algorithm


## Demo Details

- Object detection code will run on pre-recorded footage
- Display vehicle and pedestrian counts for each side on monitor


## Key Changes \& Tradeoffs

Using YOLOv3 model instead of cascade classifiers

- Haar cascade accuracy was very low $<50 \%$ due to false positives
- YOLOv4 model provided similar accuracy to YOLOv3 but higher latency
Using pre-recorded footage instead of a live camera feed for demo purposes
- Using wired IP cameras (powered with portable batteries) due to inability to access live stream of battery-powered IP cameras



## Implementation - Optimization

## Overall Solution



- Deep Q-learning model with Pytorch
- 2 layered neural network
- Huber loss function
- Toggleable online or offline model
- Called in TraCl script code to constantly update SUMO simulation traffic lights
- Outputs (North-South Green duration, East-West Green duration) to the simulation
- State input:
- queue length, average speed, current light phase, time left in phase


## Demo Details

- Using simulated pedestrian and vehicle counts during demo instead of camera data input
- Vehicles in footage will not respond to simulated light changes leading to optimization not working


## Key Changes

Light interval instead of color action states

- (North-South green duration, East-West green duration) vs North-South at single time/interval
- Safety - delayed updates don't harm upcoming cycles
- Easier to implement - no need for external timing mechanism
Action representation:



## Implementation - Simulation

## Overall Solution

- Using SUMO traffic simulator w/ TraCl Python
- Polled constantly by traffic light circuit to determine current state of physical traffic light
- Lane area detectors to mimic object detection model
- Calibrators to simulate real life traffic flow from TomTom API


## Demo Details

- Plans to implement 3D modeled simulation for demo
- Will also output live state data
- Cars at each side of intersection, average wait time, etc


## Key Changes

- Only using TomTom API instead of TomTom and HERE
- Redundant flow information


## Example Simulation Feed

```
0 EW green
1 EW yellow
2 NS green
3 NS yellow
4 \text { Pedestrian}
```



## Implementation - Circuit

## Overall Solution



- RPi outputs current light state information, sends to Arduino using serial communication
- Arduino uses SPI transmission to update light ON/OFF states stored in the TLC5928 LED Driver chip
- LED Driver outputs are connected to 12 LEDs that model a four-way intersection
- Packaged together as a custom Arduino shield PCB


## Demo Details

- The traffic light circuit will be connected to the RPi output, reflecting the optimized light timing patterns


## Key Changes

- Using an LED driver chip to control an array of individual LEDs, rather than using addressable LED strips



## Testing, Verification, Metrics - Optimization

## HOW WE TESTED

- Comparing average wait time of cars in SUMO simulation with ML model controlling light durations to same periods without using the ML model
- Over 8 periods of 1 hr in simulation time for both trials
- Latency: tested over 10 iterations of interval calculation


## FURTHER IMPROVEMENTS

- Simulation currently does not have a lot of randomness and could be closer to real life environment
- Improve before demo with more route variability

AVERAGE WAIT TIME


## Testing, Verification, Metrics - Object Detection

## HOW WE TESTED

- 100 frames of pre-recorded video at Fifth and Craig intersection; all metrics averaged over those frames
- Compared actual object counts to object vehicle counts
- Maximum vehicles detected on one side with full accuracy was 11



## FURTHER IMPROVEMENTS

- Need to re-test latency when all 4 frames are being processed concurrently; will probably get worse
- Only tested with 3 sides of the intersection because that is the only stable footage we have as of now
- Used hard-coded lane boundaries may test edge detection algorithm


## Testing, Verification, Metrics - Circuit

## HOW WE TESTED

- Partially assembled one of the PCBs and wired it to breadboarded LEDs
- Ran Arduino TB to verify lights transition as intended - Discovered wiring issue with $\mathrm{R}_{\text {IREF }}$
- Connected Arduino to RPi to verify serially-communicated control over light states
- Error statements printed to serial monitor allow us to ensure that no illegal light patterns happen

FURTHER IMPROVEMENTS

- Correctly wired PCBs have been ordered

${ }_{5} \quad$ March 2024
 OpencV
detection detection
algorithm

```
Training new Haar cascade model/ togg . Ankita - Feb 26 - Mar 18
Finding new/faster mo. Ankita - Mar 18-Ap1 Code for pede.. Ankita Detection algor. - Ankit.
```

Testing for use case re.. Ankita, Kail. - Aor 8 - Apr 22
Determine lane boundari... Ankita

## Schedule


*February tasks and deliverables not included, see website schedule for more details

## Key Takeaways

## TRY TO STICK TO WIDELY

 USED TOOLS/LIBRARIES- Better documentation = more gentle learning curve
- More/quicker support


