# The Intersection of Reinforcement Learning and Traffic Light Scheduling

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## What is the issue

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## More Seriously:

Traffic congestion is becoming more and more of a problem for both humans and the environment.

In 2013:

- 65 hours
- 3.1 megatons

My project seeks to solve this problem of traffic congestion by controlling the traffic lights.

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| Neural Networks (NN)         | Function approximators made from col-    |
|                              | lections of "neurons"                    |

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| Action Space                 | The variables that we will control with our output                                                    |
| Reward                       | Some value dependent on the state that tells us how we are doing                                      |
| Neural Networks (NN)         | Function approximators made from collections of "neurons"                                             |
| Reinforcement Learning (RL)  | An algorithm which uses positive and negative rewards to "teach" a NN to produce the behavior we want |

# **Q** Learning

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# Q Learning

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Train a neural network to approximate the reward of choosing a particular action at each time

$$Q(s_t, a_t, \pi) = \mathbb{E}\left\{\sum_{k=0}^{\infty} \gamma^k r_{t+k} \middle| s_t, a_t, \pi\right\}$$

For some policy  $\pi$ .

To choose an action, we take

$$a_t = \underset{a}{\operatorname{arg\,max}} Q(s_t, a, \pi)$$

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# Q Learning

Difficult to know what action to pick at each time.

Train a neural network to approximate the reward of choosing a particular action at each time

$$Q(s_t, a_t, \theta) = \mathbb{E}\left\{\sum_{k=0}^{\infty} \gamma^k r_{t+k} \middle| s_t, a_t, \theta\right\}$$

where the weights to our NN are  $\theta$ .

To choose an action, we take

$$a_t = \underset{a}{\operatorname{arg max}} Q(s_t, a, \theta)$$

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# Q Learning: Simply put

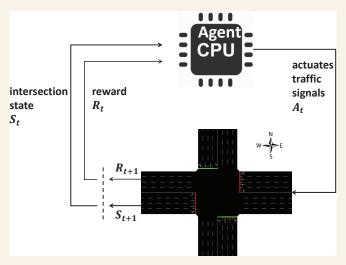


Figure: Gao et al.

# **Current Approaches**

Current methods that cities implement to limit congestion include

- Fixed timing
- Adaptive control
- Coordinated control



# **Approach**

# **Action Space:**

12 values

- 3 possible pairs of inbound edges
- 4 total non-red light settings for each pair

# Approach

## **Action Space:**

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## State Space:

72 values

- 6 possible incoming roads
- 5 lane types
- 2 features per lane type
- "One hot" vector holding index of current light signal

# Approach

#### **Action Space:**

12 values

- 3 possible pairs of inbound edges
- 4 total non-red light settings for each pair

## State Space:

72 values

- 6 possible incoming roads
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- 2 features per lane type
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## **Reward Function:**

1 value
Sum the squares of all
lane waiting times
Take the difference
across consecutive
time steps

# Differences from Previous Research: State Space

#### Previous Research

- Omnipotent state space
- Positions of cars in lanes
- Speeds of cars in lanes
- Fixed sizes

#### Current Method

- Still somewhat Big Brother
- Number of cars in each lane
- Number of cars under speed threshold
- Can be extended to any number/type of lanes

## State Space

The variables and characteristics we can observe to use as input

# Differences from Previous Research: Action Space

#### Previous Research

- Fixed timing
- Fixed path (ie yellow follows green)
- Only works for 4 way intersections

#### Current Method

- No fixed timing
- No fixed path
- Can be generalized for any intersection\*

#### **Action Space**

The variables that we will control with our output

## Differences from Previous Research: Reward

#### Previous Research

- Raw difference in wait time
- Linear wait

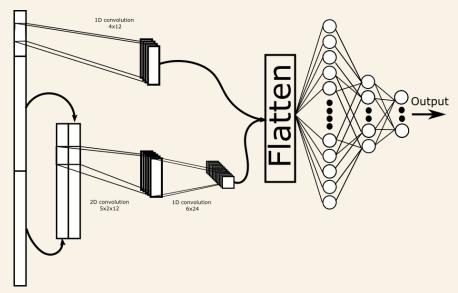
#### Current Method

- Uses proportion of change in wait
- Squares wait times per vehicle

#### Reward

Some value dependent on the state that tells us how we are doing

# Model Architecture



# Demo

# Results

# 4 way, single lane intersection

| Traffic Control Agent | Throughput | Avg Delay |
|-----------------------|------------|-----------|
| Fixed timing          | 141.0      | 19        |
| Q-learning            | 115.00     | 119.8     |

# Results

# 4 way, single lane intersection

| Traffic Control Agent | Throughput | Avg Delay |
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| Fixed timing          | 141.0      | 19        |
| Q-learning            | 115.00     | 119.8     |

# 4 way, four lane intersection

| Traffic Control Agent | Throughput | Avg Delay |
|-----------------------|------------|-----------|
| Fixed timing          | 0          | 0         |
| Q-learning            | 0          | 0         |

## Results

# 4 way, single lane intersection

| Traffic Control Agent | Throughput | Avg Delay |
|-----------------------|------------|-----------|
| Fixed timing          | 141.0      | 19        |
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## 4 way, four lane intersection

| Traffic Control Agent | Throughput | Avg Delay |
|-----------------------|------------|-----------|
| Fixed timing          | 0          | 0         |
| Q-learning            | 0          | 0         |

## 6 way, two lane intersection

| Traffic Control Agent | Throughput | Avg Delay |
|-----------------------|------------|-----------|
| Fixed timing          | 0          | 0         |
| Q-learning            | 0          | 0         |

## Roadblocks

Obviously did not get as far as I had hoped What were some of the problems?

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Understanding theory  $\neq$  Easy implementation

Sometimes code doesn't work how you would expect

## Conclusion

Lots to explore now

Compare to existing RL algorithms

Evaluate for 3, 5, and 6 way intersections