# 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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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

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

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

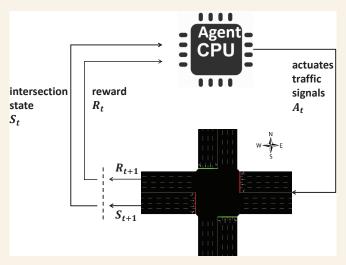


Figure: Gao et al.

# **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

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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
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## State Space:

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- 6 possible incoming roads
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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

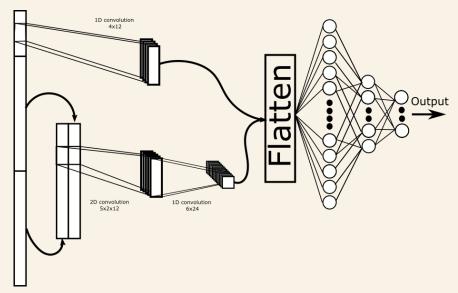
#### Current Method

- 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

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# 4 way, four lane intersection

Traffic Control Agent	Throughput	Avg Delay
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## Results

# 4 way, single lane intersection

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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