

# Joint RL meeting

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Brown University

#### Outline

1. Reducing the number of features in function approximation

2. Deep Reinforcement Learning – first draft

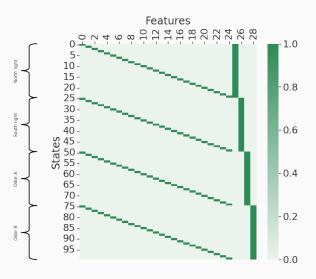
#### Outline

1. Reducing the number of features in function approximation

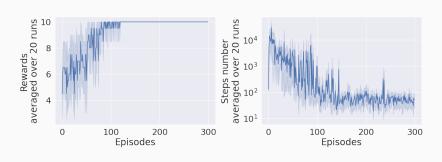
2. Deep Reinforcement Learning – first draft

# Features matrix – allocentric agent

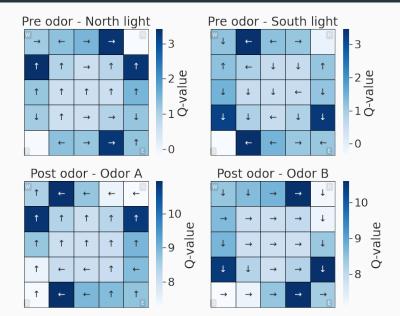
→ Reduced from 100 to 25 locations + 4 cues



# Rewards and steps – allocentric agent

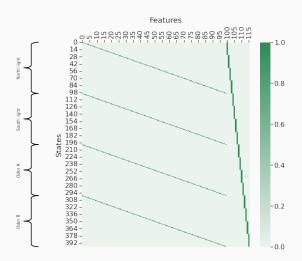


# Q-values learned – allocentric agent

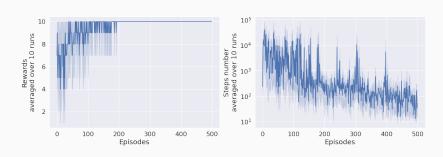


# Features matrix – egocentric agent

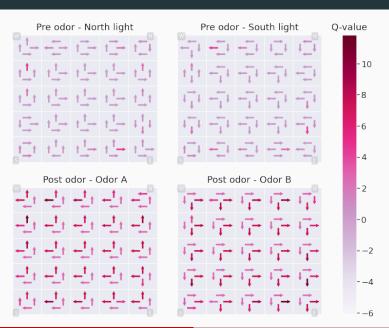
→ Reduced from 400 to 100 locations (25 locations x 4 head directions) + 16 cues (4 cues x 4 head directions)



# Rewards and steps – egocentric agent



# Q-values learned – egocentric agent



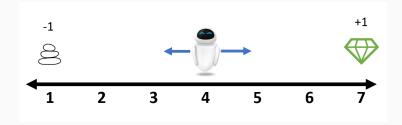
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#### Outline

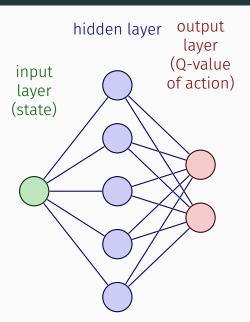
Reducing the number of features in function approximation

2. Deep Reinforcement Learning – first draft

# Toy task: Random Walk 1D



#### Network used

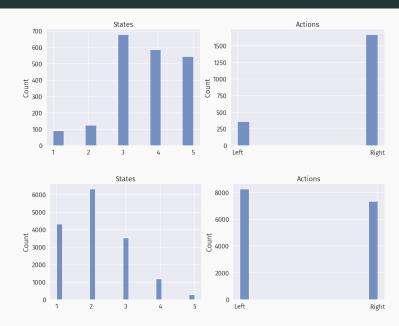


### Algorithm

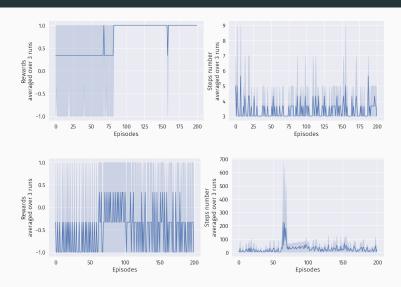
#### Algorithm 1: An algorithm with caption

```
Data: X = [S_1, S_2, S_3, ..., S_n]
Data: y = [q_1, q_2, q_3, ..., q_n]
Result: v = x^n
q' \leftarrow r + \gamma q;
Loss \leftarrow (y - \hat{y}_{pred}) \times \Delta W;
while N \neq 0 do
    if N is even then
        X \leftarrow X \times X;
       N \leftarrow \frac{N}{2}; /* This is a comment */
    else
         if N is odd then
       end
    end
end
```

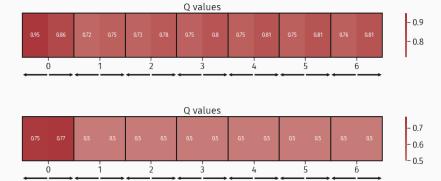
#### States and actions



# Rewards and steps



### Q-values learned



# Questions ?