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## 7. Linear Q-Learning

**Extension Note:** Project 5 due date has been extended by 1 **more** day to **September 6 23:59UTC** .

In this tab, you will implement the Q-learning algorithm with linear function approximation.

Recall the linear approximation we chose.

$$Q(s, c, \theta) = \phi(s, c)^T \theta$$

with

$$\phi(s, c) = \begin{bmatrix} \mathbf{0} \\ \vdots \\ \mathbf{0} \\ \psi_R(s) \\ \mathbf{0} \\ \vdots \\ \mathbf{0} \end{bmatrix}$$

Now, define  $\hat{\theta}_i$  for  $i$  in range  $1, d_C$  so that:

$$\theta = \begin{bmatrix} \hat{\theta}_1 \\ \vdots \\ \hat{\theta}_i \\ \vdots \\ \hat{\theta}_{d_C} \end{bmatrix}$$

With this notation, we get:

$$Q(s, c, \theta) = \psi_R(s)^T \hat{\theta}_c$$

In practice, we can implement  $\hat{\theta}$  as a 2D array, so that

$$\begin{bmatrix} Q(s, 1, \theta) \\ \vdots \\ Q(s, d_C, \theta) \end{bmatrix} = \begin{bmatrix} \hat{\theta}_1^T \\ \vdots \\ \hat{\theta}_{d_C}^T \end{bmatrix} \cdot \psi_R(s)$$

## Epsilon-greedy exploration

1.0/1 point (graded)

Now you will write a function `epsilon_greedy` that implements the  $\varepsilon$ -greedy exploration policy using the current Q-function.

**Hint:** You can access  $Q(s, c, \theta)$  using `q_value = (theta @ state_vector)[tuple2index(action_index, object_index)]`

**Available Functions:** You have access to the NumPy python library as `np` and functions `tuple2index` and `index2tuple`. Your code should also use constants `NUM_ACTIONS` and `NUM_OBJECTS`

```

1 def epsilon_greedy(state_vector, theta, epsilon):
2     """Returns an action selected by an epsilon-greedy exploration policy
3
4     Args:
5         state_vector (np.ndarray): extracted vector representation
6         theta (np.ndarray): current weight matrix
7         epsilon (float): the probability of choosing a random command
8
9     Returns:
```

```

10         (int, int): the indices describing the action/object to take
11         """
12         # TODO Your code here
13         explore = np.random.random() <= epsilon
14         if explore:
15             action_index, object_index = np.random.choice(NUM_ACTIONS, 1)[0], np.random.choice(NUM_OBJECTS, 1)[0]

```

Press ESC then TAB or click outside of the code editor to exit

Correct

```

def epsilon_greedy(state_vector, theta, epsilon):
    """Returns an action selected by an epsilon-greedy exploration policy

    Args:
        state_vector (np.ndarray): extracted vector representation
        theta (np.ndarray): current weight matrix
        epsilon (float): the probability of choosing a random command

    Returns:
        (int, int): the indices describing the action/object to take
    """
    coin = np.random.random_sample()
    if coin < epsilon:
        action_index = np.random.randint(NUM_ACTIONS)
        object_index = np.random.randint(NUM_OBJECTS)
    else:
        q_values = theta @ state_vector
        index = np.argmax(q_values)
        action_index, object_index = index2tuple(index)
    return (action_index, object_index)

```

## Test results

CORRECT

[See full output](#)[See full output](#)

Submit

You have used 1 of 25 attempts

**i** Answers are displayed within the problem

## Linear Q-learning

1.0/1 point (graded)

Write a function `linear_q_learning` that updates the theta weight matrix, given the transition date  $(s, a, R(s, a), s')$ .**Reminder:** You should implement this function locally first. You should test this function along with the next one and make sure you achieve reasonable performance**Hint:** You can access  $Q(s, a, \theta)$  using `q_value = (theta @ state_vector)[tuple2index(action_index, object_index)]`**Available Functions:** You have access to the NumPy python library as `np`. You should also use constants `ALPHA` and `GAMMA` in your code

```

1 def linear_q_learning(theta, current_state_vector, action_index, object_index,
2                       reward, next_state_vector, terminal):
3     """Update theta for a given transition
4
5     Args:

```

```
6      theta (np.ndarray): current weight matrix
7      current_state_vector (np.ndarray): vector representation of current state
8      action_index (int): index of the current action
9      object_index (int): index of the current object
10     reward (float): the immediate reward the agent receives from playing current command
11     next_state_vector (np.ndarray): vector representation of next state
12     terminal (bool): True if this episode is over
13
14     Returns:
15         None
16     """
```

Press ESC then TAB or click outside of the code editor to exit

Correct

```
def linear_q_learning(theta, current_state_vector, action_index, object_index,
                    reward, next_state_vector, terminal):
    """Update theta for a given transition

    Args:
        theta (np.ndarray): current weight matrix
        current_state_vector (np.ndarray): vector representation of current state
        action_index (int): index of the current action
        object_index (int): index of the current object
        reward (float): the immediate reward the agent receives from playing current command
        next_state_vector (np.ndarray): vector representation of next state
        terminal (bool): True if this episode is over

    Returns:
        None
    """
    q_values_next = theta @ next_state_vector
    maxq_next = np.max(q_values_next)

    q_values = theta @ current_state_vector
    cur_index = tuple2index(action_index, object_index)
    q_value_cur = q_values[cur_index]

    target = reward + GAMMA * maxq_next * (1 - terminal)

    theta[cur_index] = theta[cur_index] + ALPHA * (
        target - q_value_cur) * current_state_vector
```

## Test results

CORRECT

[See full output](#)[See full output](#)

Submit

You have used 4 of 25 attempts

**i** Answers are displayed within the problem

## Evaluate linear Q-learning on Home World game

1/1 point (graded)

Adapt your `run_episode` function to call `linear_Q_learning` and evaluate your performance using hyperparameters:

Set `NUM_RUNS` = 5, `NUM_EPIS_TRAIN` = 25, `NUM_EPIS_TEST` = 50,  $\gamma = 0.5$ , `TRAINING_EP` = 0.5, `TESTING_EP` = 0.05 and the learning rate  $\alpha = 0.001$ .

Please enter the *average episodic rewards* of your Q-learning algorithm when it converges.

0.3262413375213623

✓ Answer: 0.37

Submit

You have used 2 of 6 attempts

**i** Answers are displayed within the problem



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| ? <a href="#">[Staff] Epsilon-greedy exploration</a>  | 6  |
| <a href="#">Hi, Over all this course i got along all subjects with more or less commodity but with this topic (Linear Q-learning) i can't get the big picture. I follo...</a> |    |
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| <a href="#">I get the error again when I submit epsilon-greedy.</a>   |    |
| 💬 <a href="#">Why don't we achieve optimal results?</a>   | 1  |
| <a href="#">OK, same question as before. I received a green tick for my average episodic rewards after convergence. However, this number falls a fair bit sho...</a>          |    |
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| <a href="#">if terminal is True, then max_q_value_nxt = 0 else max_q_value_nxt is max of theta time time next state vector y = reward + (GAMMA * max_q_va...</a>              |    |

Struggling with Linear Q-learning

- ? Where am I going wrong? q\_value = (theta @ current\_state\_vector)[tuple2index(action\_index, object\_index)] max q\_value next = 0 if terminal else ... 21 new\_
- ? Epsilon-greedy exploration - Hints for getting started 3  
Hi All, I find it a bit difficult to make progress with this problem. Would you please provide any useful hints. Perhaps, how this function is differen...
- ? [Staff] Evaluate linear Q-learning on Home World game 21 new\_ 26  
Community TA
- ✓ [staff] Average episodic reward 11  
Hi, my model converges with given in the task parameters, all functions are marked correct by the grader. But I can't figure out where should I pi...
- 💬 doesn't converge 4

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