

**Exercise 1**

(a) Using Theorem 3.6: With  $|\mathcal{H}| = 3^3 = 27$

$$\Pr_{T \mathcal{D}^m} (\forall h \in \mathcal{H} : |err_T(h) - err_D(h)| \leq \epsilon) > 1 - \delta$$

$$\Pr_{T \mathcal{D}^m} (\forall h \in \mathcal{H} : |err_T(h) - err_D(h)| \leq \epsilon) > 0.9$$

$$\Rightarrow \delta = 0.1$$

$$m \geq \frac{1}{2\epsilon^2} \log \left( \frac{2|\mathcal{H}|}{\delta} \right)$$

$$143 \geq \frac{1}{2\epsilon^2} \log \left( \frac{2 \cdot 3^3}{0.1} \right)$$

$$143 \geq \frac{1}{2\epsilon^2} (\log(54) - \log(0.1))$$

$$143 \geq \frac{1}{2\epsilon^2} (\log(54) - \log(0.1))$$

$$\epsilon^2 \geq \frac{(\log(54) - \log(0.1))}{143 \cdot 2}$$

$$|\epsilon| \geq \sqrt{\frac{(\log(54) - \log(0.1))}{286}}$$

$$\Rightarrow \epsilon \geq \sqrt{\frac{(\log(54) - \log(0.1))}{286}}$$

$$\Pr_{T \mathcal{D}^m} (\forall h \in \mathcal{H} : |err_T(h) - err_D(h)| \leq \epsilon) > 0.9$$

$$\Pr_{T \mathcal{D}^m} \left( \forall h \in \mathcal{H} : |0.03 - err_D(h)| \leq \sqrt{\frac{(\log(54) - \log(0.1))}{286}} \right) > 0.9$$

$$\Rightarrow err_D(h) \leq 0.03 + \sqrt{\frac{(\log(54) - \log(0.1))}{286}} \simeq 0.208149 \simeq 0.21$$

(b) Using Theorem 3.4:

$$\Pr_{T \mathcal{D}^m} (\forall h \in \mathcal{H} : \text{if } h \text{ is consistent with } T, \text{ then } err_D(h) \leq \epsilon) > 1 - \delta$$

$$\Pr_{T \mathcal{D}^m} (\forall h \in \mathcal{H} : \text{if } h \text{ is consistent with } T, \text{ then } err_D(h) \leq 0.01) > 0.9$$

$$\Rightarrow \epsilon = 0.01, \delta = 0.1$$

$$m \geq \frac{1}{\epsilon} \ln \left( \frac{|\mathcal{H}|}{\delta} \right)$$

$$m \geq \frac{1}{0.01} \ln \left( \frac{3^3}{0.1} \right)$$

$$m \geq 100(\ln(27) - \ln(0.1)) \simeq 559.84$$

$$\Rightarrow m \geq 560$$

## Exercise 2

## Exercise 3

## Exercise 4

See Referencesappendix for code.

```
1 Final probabilities: [0.4 0.4 0.2]
2
3 Tracked weight vectors:
4
5 Round: 1 Weights: [1. 1. 1.]
6 Round: 2 Weights: [0.5 0.5 1. ]
7 Round: 3 Weights: [0.5 0.25 0.5 ]
8 Round: 4 Weights: [0.25 0.25 0.25]
9 Round: 5 Weights: [0.25 0.125 0.125]
10 Round: 6 Weights: [0.125 0.125 0.0625]
11 Round: 7 Weights: [0.125 0.0625 0.04419417]
```

## Exercise 5

See Referencesappendix for code.

```
1 Probabilities:
2
3 Round: 1 Probabilities: [0.33 0.33 0.33]
4 Round: 2 Probabilities: [0.46 0.27 0.27]
5 Round: 3 Probabilities: [0.28 0.51 0.21]
6 Round: 4 Probabilities: [0.25 0.42 0.33]
7
8 Weight vectors:
9
10 Round: 1 Weights: [1. 1. 1.]
11 Round: 2 Weights: [2.83 1. 1. ]
12 Round: 3 Weights: [2.83 8.48 1. ]
13 Round: 4 Weights: [2.83 8.48 5.32]
```

## Exercise 6

## Appendix

### Code for Exercise 4

```
1 import numpy as np
2
3
4 def mwu_algorithm(loss_matrix, events, rounds, alpha):
5     # initial weight vector of 1s
6     weights = np.ones((loss_matrix.shape[0]))
7     weights_tracking = {}
8     weights_tracking[0] = weights
9     # more convenient to loop through rounds and events
10    rounds_arr = [i for i in range(rounds)]
11    for round, event in zip(rounds_arr, events):
```

```
12     # getting the current probabilities, not really needed here
13     p = probabilities(weights)
14     # need to use event-1 as events start at 1 but indexing at 0
15     weights = np.power((1 - alpha), loss_matrix[:, event-1]) * weights
16     # loss isn't really needed
17     loss = calculate_loss(loss_matrix, p, event-1)
18     weights_tracking[round+1] = weights
19
20     return p, weights_tracking
21
22 def probabilities(weights):
23     return weights / np.sum(weights)
24
25 def calculate_loss(loss_matrix, probabilities, event):
26     return np.sum(probabilities * loss_matrix[:, event])
27
28
29 loss_matrix = np.array([[0,1,1,0],
30                        [1,0,1,1],
31                        [1,1,0,0.5]])
32
33 observed_events = [3,1,2,1,2,4]
34
35 p_6, weights_tracking = mwu_algorithm(loss_matrix, observed_events, 6, alpha
36                                     =0.5)
37
38 print(f'Final probabilities: {p_6}\n')
39 print(f'Tracked weight vectors: \n')
40 for key, val in weights_tracking.items():
41     print(f'Round:\t{key + 1}\tWeights:\t{val}')
```

## Code for Exercise 5

```
1 import numpy as np
2 from copy import deepcopy
3
4 def exp3(gamma, rounds, actions, rewards):
5     weights = np.ones((len(actions)))
6     rounds_arr = [i for i in range(rounds)]
7     n = len(actions)
8     # for tracking weights and probabilities
9     weights_tracking = {}
10    probabilities_tracking = {}
11    weights_tracking[0] = np.ones(len(actions))
12    probabilities_tracking[0] = probability_dist(weights, gamma)
13    for round, action in zip(rounds_arr, actions):
14        probabilities = probability_dist(weights, gamma)
15        probabilities_tracking[round] = probabilities
16        reward = rewards[action]
17        weights = update_weights(weights, reward, probabilities, action,
18                                gamma)
19        weights_tracking[round + 1] = deepcopy(weights)
20
21    probabilities_tracking[rounds] = probability_dist(weights, gamma)
22    return weights_tracking, probabilities_tracking
23
24 def probability_dist(weights, gamma):
25     return (1 - gamma) * (weights / np.sum(weights)) + gamma / len(weights)
26
27 def update_weights(weights, reward, probabilities, action, gamma):
28     n = len(weights)
```

```
28     # only update chosen action
29     weights[action] = weights[action] * np.exp((gamma * reward) / (n *
30     probabilities[action]))
31     return weights
32
33 action_seq = np.array([ 1, 2, 3 ])
34 rewards = np.array([ 3, 5, 3 ]) * np.log(2)
35
36 weights, probs = exp3(gamma=0.5, rounds=3, actions=action_seq - 1, rewards=
37     rewards)
38 print(f'Probabilities: \n')
39 for key, val in probs.items():
40     print(f'Round:\t{key + 1}\tProbabilities:\t{val.round(2)}')
41
42 print(f'\nWeight vectors: \n')
43 for key, val in weights.items():
44     print(f'Round:\t{key + 1}\tWeights:\t{val.round(2)}')
```