Feedback - 作業一

You submitted this quiz on **Tue 17 May 2016 7:39 PM CST**. You got a score of **400.00** out of **400.00**. However, you will not get credit for it, since it was submitted past the deadline.

Question 1 Which of the following problems are best suited for machine learning? (i) Classifying numbers into primes and non-primes (ii) Detecting potential fraud in credit card charges (iii) Determining the time it would take a falling object to hit the ground (iv) Determining the optimal cycle for traffic lights in a busy intersection (v) Determining the age at which a particular medical test is recommended **Your Answer Explanation** Score (i), (ii), (iii), and (iv) (i) and (ii) (ii), (iv), and (v) 20.00 (i), (ii), (iii), and (iv) none of the other choices Total 20.00 / 20.00

Question 2

For Questions 2-5, identify the best type of learning that can be used to solve each task below. Play chess better by practicing different strategies and receive outcome as feedback

Your Answer	Score	Explanation
supervised learning		

onone of other choices		
• reinforcement learning	~	20.00
active learning		
unsupervised learning		
Total		20.00 / 20.00
Question 3		

Question 3 Categorize books into groups without given topic	cs	
Your Answer	Score	Explanation
supervised learning		
reinforcement learning		
unsupervised learning	20.00	
onone of other choices		
active learning		
Total	20.00 / 20.00	

Recognize whether there is a face in the picture by a thousand face pictures and ten thousand non-face pictures.

Your Answer	Score	Explanation
active learning		
none of other choices		
reinforcement learning		
unsupervised learning		

supervised learning	~	20.00
Total		20.00 / 20.00

Selectively schedule experiments on mice to quickly evaluate the potential of cancer medicines

Your Answer		Score	Explanation
supervised learning			
onone of other choices			
active learning	~	20.00	
reinforcement learning			
unsupervised learning			
Total		20.00 / 20.00	

Question 6

Question 6-8 are about Off-Training-Set error.

Let $\mathcal{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N, \mathbf{x}_{N\!H}, \dots, \mathbf{x}_{N\!H}\}$ and $\mathcal{Y} = \{-1, +1\}$ (binary classification). Here the set of training examples is $\mathcal{D} = \left\{ (\mathbf{x}_n, y_n) \right\}_{n=1}^N$, where $y_n \in \mathcal{Y}$, and the set of test inputs is $\left\{ \mathbf{x}_{N\!H} \right\}_{\ell=1}^L$. The Off-Training-Set error~(OTS) with respect to an underlying target f and a hypothesis g is

$$E_{OTS}(g,f) = \frac{1}{L} \sum_{\ell=1}^{L} \left[g(\mathbf{x}_{N\!+\!\ell}) \neq f(\mathbf{x}_{N\!+\!\ell}) \right].$$

Consider $f(\mathbf{x}) = +1$ for all \mathbf{x} and

$$g(\mathbf{x}) = \begin{cases} +1, & \text{for } \mathbf{x} = \mathbf{x}_k \text{ and } k \text{ is odd and } 1 \le k \le N + L \\ -1, & \text{otherwise} \end{cases}$$

 $E_{OTS}(g,f)=$? (Please note the difference between floor and ceiling functions in the choices)

Your Answer	Score	Explanation
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	✓ 20.00
$\bigcirc \frac{1}{L} \times (\lfloor \frac{N+L}{2} \rfloor - \lceil \frac{N}{2} \rceil)$	
$\bigcirc \frac{1}{L} \times (\lceil \frac{N+L}{2} \rceil - \lceil \frac{N}{2} \rceil)$	
onone of the other choices	
$\bigcirc \frac{1}{L} \times (\lceil \frac{N+L}{2} \rceil - \lfloor \frac{N}{2} \rfloor)$	
Total	20.00 / 20.00

We say that a target function f can "generate" \mathcal{D} in a noiseless setting if $f(\mathbf{x}_n) = y_n$ for all $(\mathbf{x}_n, y_n) \in \mathcal{D}$.

For all possible $f: \mathcal{X} \to \mathcal{Y}$, how many of them can generate \mathcal{D} in a noiseless setting? Note that we call two functions f_1 and f_2 the same if $f_1(\mathbf{x}) = f_2(\mathbf{x})$ for all $\mathbf{x} \in \mathcal{X}$.

Your Answer		Score	Explanation
$\odot 2^L$	~	20.00	
$\supset 2^N$			
$\sum_{k=0}^{N+L}$			
) 1			
none of the other choices			
Total		20.00 / 20.00	

Question 8

A determistic algorithm $\mathcal A$ is defined as a procedure that takes $\mathcal D$ as an input, and outputs a hypothesis g. For any two deterministic algorithms $\mathcal A_1$ and $\mathcal A_2$, if all those f that can "generate" $\mathcal D$ in a noiseless setting are equally likely in probability,

Your Answer	Score	Explanation
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 \bigcirc For any given f that "generates" \mathcal{D} ,

$$\left\{ E_{OTS} \left(\mathcal{A}_1(\mathcal{D}), f \right) \right\} = \left\{ E_{OTS} \left(\mathcal{A}_2(\mathcal{D}), f \right) \right\}.$$

none of the other choices

● 20.00

$$\mathbb{E}_f\Big\{E_{OTS}\big(\mathcal{A}_1(\mathcal{D}),f\big)\Big\} = \mathbb{E}_f\Big\{E_{OTS}\big(\mathcal{A}_2(\mathcal{D}),f\big)\Big\}.$$

 \bigcirc For any given f' that does not "generate" \mathcal{D} ,

$$\left\{ E_{OTS} \left(\mathcal{A}_1(\mathcal{D}), f' \right) \right\} = \left\{ E_{OTS} \left(\mathcal{A}_2(\mathcal{D}), f' \right) \right\}.$$

 $\mathbb{E}_f\Big\{E_{OTS}\big(\mathcal{A}_1(\mathcal{D}),f\big)\Big\} = \mathbb{E}_f\Big\{E_{OTS}\big(f,f\big)\Big\}.$

Total 20.00 / 20.00

Question 9

For Questions 9-12, consider the bin model introduced in class.

Consider a bin with infinitely many marbles, and let μ be the fraction of orange marbles in the bin, and ν is the fraction of orange marbles in a sample of 10 marbles.

If $\mu=0.5$, what is the probability of $\nu=\mu$?

Your Answer		Score	Explanation
0.39			
0.24	~	20.00	
none of the other choices			
0.90			
0.56			
Total		20.00 / 20.00	

If $\mu = 0.9$, what is the probability of $\nu = \mu$?

Your Answer		Score	Explanation
0.24			
none of the other choices			
0.90			
0.56			
0.39	~	20.00	
Total		20.00 / 20.00	

Question 11

If $\mu = 0.9$, what is the probability of $\nu \leq 0.1$?

Your Answer		Score	Explanation
onone of the other choices			
	~	20.00	
0.1×10^{-9}			
$0.8.5 \times 10^{-9}$			
$\bigcirc 1.0 \times 10^{-9}$			
Total		20.00 / 20.00	

Question 12

If $\mu=0.9$, what is the bound given by Hoeffding's Inequality for the probability of $\nu\leq0.1$?

Your Answer Score Explanation

$\bigcirc 5.52 \times 10^{-10}$		
$\bigcirc 5.52 \times 10^{-4}$		
O none of the other choices		
	~	20.00
$\bigcirc 5.52 \times 10^{-8}$		
Total		20.00 / 20.00

Questions 13-14 illustrate what happens with multiple bins using dice to indicate 6 bins. Please note that the dice is not meant to be thrown for random experiments in this problem. They are just used to bind the six faces together. The probability below only refers to drawing from the bag.

Consider four kinds of dice in a bag, with the same (super large) quantity for each kind.

A: all even numbers are colored orange, all odd numbers are colored green

B: all even numbers are colored green, all odd numbers are colored orange

C: all small (1-3) are colored orange, all large numbers (4-6) are colored green

D: all small (1-3) are colored green, all large numbers (4-6) are colored orange

If we pick 5 dice from the bag, what is the probability that we get five orange 1's?

Your Answer		Score	Explanation
	~	20.00	
onone of the other choices			
$\bigcirc \frac{46}{256}$			
$\bigcirc \frac{1}{256}$			
$\bigcirc \frac{31}{256}$			
Total		20.00 / 20.00	

Question 14

If we pick 5 dice from the bag, what is the probability that we get "some number" that is purely orange?

Your Answer		Score	Explanation
31 256	~	20.00	
<u>46</u> 256			
$\frac{8}{256}$			
none of the other choices			
$\frac{1}{256}$			
		20.00 / 20.00	

Question 15

For Questions 15-20, you will play with PLA and pocket algorithm. First, we use an artificial data set to study PLA. The data set is in

https://d396qusza40orc.cloudfront.net/ntumlone%2Fhw1%2Fhw1_15_train.dat

Each line of the data set contains one (\mathbf{x}_n, y_n) with $\mathbf{x}_n \in \mathbb{R}^4$. The first 4 numbers of the line contains the components of \mathbf{x}_n orderly, the last number is y_n .

Please initialize your algorithm with $\mathbf{w}=0$ and take $\mathrm{sign}(0)$ as -1

Implement a version of PLA by visiting examples in the naive cycle using the order of examples in the data set. Run the algorithm on the data set. What is the number of updates before the algorithm halts?

Your Answer		Score	Explanation
○ ≥ 201 updates			
31 - 50 updates	~	20.00	
11 - 30 updates			
< 10 updates			
51 - 200 updates			
Total		20.00 / 20.00	

Implement a version of PLA by visiting examples in fixed, pre-determined random cycles throughout the algorithm. Run the algorithm on the data set. Please repeat your experiment for 2000 times, each with a different random seed. What is the average number of updates before the algorithm halts?

Your Answer		Score	Explanation
11 - 30 updates			
0 < 10 updates			
31 - 50 updates	~	20.00	
51 - 200 updates			
) ≥ 201 updates			
Total		20.00 / 20.00	

Question 17

Implement a version of PLA by visiting examples in fixed, pre-determined random cycles throughout the algorithm, while changing the update rule to be

$$\mathbf{w}_{t+1} \leftarrow \mathbf{w}_t + \eta y_{n(t)} \mathbf{x}_{n(t)}$$

with $\eta=0.5$. Note that your PLA in the previous Question corresponds to $\eta=1$. Please repeat your experiment for 2000 times, each with a different random seed. What is the average number of updates before the algorithm halts?

Your Answer	Score	Explanation
51 - 200 updates		
○ 11 - 30 updates		
○ < 10 updates		

31 - 50 updates	~	20.00
$\bigcirc \ge 201$ updates		
Total		20.00 / 20.00

Next, we play with the pocket algorithm. Modify your PLA in Question 16 to visit examples purely randomly, and then add the 'pocket' steps to the algorithm. We will use

https://d396qusza40orc.cloudfront.net/ntumlone%2Fhw1%2Fhw1_18_train.dat

as the training data set \mathcal{D} , and

https://d396qusza40orc.cloudfront.net/ntumlone%2Fhw1%2Fhw1_18_test.dat

as the test set for "verifying" the g returned by your algorithm (see lecture 4 about verifying). The sets are of the same format as the previous one.

Run the pocket algorithm with a total of 50 updates on \mathcal{D} , and verify the performance of

 \mathbf{w}_{POCKET} using the test set. Please repeat your experiment for 2000 times, each with a different random seed. What is the average error rate on the test set?

Your Answer		Score	Explanation
0.4 - 0.6			
0.6 - 0.8			
○ ≥ 0.8			
0.2 - 0.4			
• < 0.2	~	20.00	
Total		20.00 / 20.00	

Question 19

Modify your algorithm in Question 18 to return \mathbf{w}_{50} (the PLA vector after 50 updates) instead of $\hat{\mathbf{w}}$ (the pocket vector) after 50 updates. Run the modified algorithm on \mathcal{D} , and verify the

performance using the test set. Please repeat your experiment for 2000 times, each with a different random seed. What is the average error rate on the test set?

Your Answer	Score	Explanation
0.4 - 0.6		
0.2 - 0.4	✓ 20.00	
$\bigcirc \ge 0.8$		
0.6 - 0.8		
O < 0.2		
Total	20.00 / 20.00	

Question 20

Modify your algorithm in Question 18 to run for 100 updates instead of 50, and verify the performance of \mathbf{w}_{POCKET} using the test set. Please repeat your experiment for 2000 times, each with a different random seed. What is the average error rate on the test set?

Your Answer		Score	Explanation
• < 0.2	~	20.00	
$0 \ge 0.8$			
0.2 - 0.4			
0.6 - 0.8			
0.4 - 0.6			
Total		20.00 / 20.00	
Total		20.00 / 20.00	