Feedback - Homework 3

Help Center

You submitted this quiz on **Sun 3 Jan 2016 2:41 PM PST**. You got a score of **300.00** out of **400.00**. However, you will not get credit for it, since it was submitted past the deadline.

Question 1

Decision Tree

Impurity functions play an important role in decision tree branching. For binary classification problems, let μ_+ be the fraction of positive examples in a data subset, and $\mu_-=1-\mu_+$ be the fraction of negative examples in the data subset.

The Gini index is $1 - \mu_+^2 - \mu_-^2$. What is the maximum value of the Gini index among all $\mu_+ \in [0,1]$?

Your Answer		Score	Explanation
O 1			
0.25			
O 0			
0.5	~	20.00	
0.75			
Total		20.00 / 20.00	

Question 2

Following Question 1, there are four possible impurity functions below. We can normalize each impurity function by dividing it with its maximum value among all $\mu_+ \in [0,1]$. For instance, the classification error is simply $\min(\mu_+,\mu_-)$ and its maximum value is 0.5. So the normalized classification error is $2\min(\mu_+,\mu_-)$. After normalization, which of the following impurity function is equivalent to the normalized Gini index?

Your	Your Answer				Score	Explanation	
			1	,			

 \bigcirc the entropy, which is $-\mu_+ \ln \mu_+ - \mu_- \ln \mu_-$, with

$0\log 0 \equiv 0.$	
• the squared regression error (used for branching in classification data sets), which is by definition $\mu_+(1-(\mu_+-\mu))^2 + \mu(-1-(\mu_+-\mu))^2.$	✓ 20.00
\bigcirc the closeness, which is $1- \mu_+ - \mu $	
onone of the other choices	
\bigcirc the classification error $\min(\mu_+,\mu)$.	
Total	20.00 / 20.00

Question 3

Random Forest

If bootstrapping is used to sample N'=pN examples out of N examples and N is very large. Approximately how many of the N examples will not be sampled at all?

Your Answer		Score	Explanation
$\bigcirc (1 - e^{-1/p}) \cdot N$			
$\bigcirc e^{-1/p} \cdot N$			
$\bigcirc e^{-1} \cdot N$			
$\bullet e^{-p} \cdot N$	✓	20.00	
$\bigcirc (1-e^{-p})\cdot N$			
Total		20.00 / 20.00	

Question 4

Consider a Random Forest G that consists of three binary classification trees $\{g_k\}_{k=1}^3$, where each tree is of test 0/1 error $E_{\rm out}(g_1)=0.1$, $E_{\rm out}(g_2)=0.2$, $E_{\rm out}(g_3)=0.3$. Which of the following is the exact possible range of $E_{\rm out}(G)$?

Your Answer	Score	Explanation

$ 0 \le E_{\text{out}}(G) \le 0.3 $	✓ 20.00
$\bigcirc 0.1 \le E_{\text{out}}(G) \le 0.3$	
$\bigcirc 0.2 \le E_{\text{out}}(G) \le 0.3$	
$\bigcirc 0.1 \le E_{\text{out}}(G) \le 0.6$	
$\bigcirc 0 \le E_{\text{out}}(G) \le 0.1$	
Total	20.00 / 20.00

Question 5

Consider a Random Forest G that consists of K binary classification trees $\{g_k\}_{k=1}^K$, where K is an odd integer. Each g_k is of test 0/1 error $E_{\mathrm{out}}(g_k)=e_k$. Which of the following is an upper bound of $E_{\mathrm{out}}(G)$?

Your Answer	Score	Explanation
	20.00	
$\bigcirc \frac{1}{K+1} \sum_{k=1}^{K} e_k$		
\bigcirc max _{1$\leq k \leq K$} e_k		
$\bigcirc \frac{1}{K} \sum_{k=1}^{K} e_k$		
$\bigcirc \min_{1 \leq k \leq K} e_k$		
Total	20.00 / 20.00	

Question 6

Gradient Boosting

Let ϵ_t be the weighted 0/1 error of each g_t as described in the AdaBoost algorithm (Lecture 208), and $U_t = \sum_{n=1}^N u_n^{(t)}$ be the total example weight during AdaBoost. Which of the following equation expresses U_{T+1} by ϵ_t ?

Your Answer	Score	Explanation
$\bigcap_{t=1}^{T} (2\sqrt{\epsilon_t(1-\epsilon_t)})$		

$\bigcirc \sum_{t=1}^{T} \epsilon_t$		
$\bigcirc \sum_{t=1}^{T} (2\sqrt{\epsilon_t(1-\epsilon_t)})$		
$\odot \prod_{t=1}^T \epsilon_t$	× -5.00	
onone of the other choices		
Total	-5.00 / 20.00	

Question 7

For the gradient boosted decision tree, if a tree with only one constant node is returned as g_1 , and if $g_1(\mathbf{x})=2$, then after the first iteration, all s_n is updated from 0 to a new constant $\alpha_1g_1(\mathbf{x}_n)$. What is s_n ?

Your Answer		Score	Explanation
onone of the other choices			
$\bigcirc \min_{1 \le n \le N} y_n$			
O 2			
\bigcirc max $_{1 \le n \le N} y_n$			
	~	20.00	
Total		20.00 / 20.00	

Question 8

For the gradient boosted decision tree, after updating all s_n in iteration t using the steepest η as α_t , what is the value of $\sum_{n=1}^N s_n g_t(\mathbf{x}_n)$?

Your Answer	Score	Explanation
O		
$\bigcirc \sum_{n=1}^{N} y_n s_n$		
	✓ 20.00	

$\bigcirc \sum_{n=1}^{N} y_n^2$		
onone of the other choices		
Total	20.00 / 20.00	

Question 9

Neural Network

Consider Neural Network with sign(s) instead of tanh(s) as the transformation functions. That is, consider Multi-Layer Perceptrons. In addition, we will take +1 to mean logic TRUE, and -1 to mean logic FALSE. Assume that all x_i below are either +1 or -1. Which of the following perceptron

$$g_A(\mathbf{x}) = \operatorname{sign}\left(\sum_{i=0}^d w_i x_i\right).$$

implements

$$OR(x_1, x_2, ..., x_d).$$

Your Answer		Score	Explanation
$\bigcirc (w_0, w_1, w_2, \dots, w_d) = (-d+1, -1, -1, \dots, -1)$			
$\bigcirc (w_0, w_1, w_2, \dots, w_d) = (-d+1, +1, +1, \dots, +1)$			
$ (w_0, w_1, w_2, \dots, w_d) = (d-1, +1, +1, \dots, +1) $	~	20.00	
onone of the other choices			
$\bigcirc (w_0, w_1, w_2, \dots, w_d) = (d - 1, -1, -1, \dots, -1)$			
Total		20.00 / 20.00	

Question 10

Continuing from Question 9, among the following choices of D, which D is the smallest for some 5-D-1 Neural Network to implement $XOR(x_1, x_2, x_3, x_4, x_5)$?

Your Answer Score Explanation

O 9	
O 7	
O 5	
⊙ 3 ×	-5.00
\bigcirc 1	
Total	-5.00 / 20.00

Question 11

For a Neural Network with at least one hidden layer and $\tanh(s)$ as the transformation functions on all neurons (including the output neuron), what is true about the gradient components (with respect to the weights) when all the initial weights $w_{ij}^{(\ell)}$ are set to 0?

Your Answer	Score	Explanation
onone of the other choices		
\bigcirc only the gradient components with respect to $w_{01}^{(L)}$ may be non-zero, all other gradient components must be zero		
\bigcirc only the gradient components with respect to $w_{0j}^{(\ell)}$ for $j>0$ may non-zero, all other gradient components must be zero		
all the gradient components are zero		
$ullet$ only the gradient components with respect to $w_{j1}^{(L)}$ for $j>0$ may be non-zero, all other gradient components must be zero	× -5.00	
Total	-5.00 / 20.00	

Question 12

For a Neural Network with one hidden layer and $\tanh(s)$ as the transformation functions on all neurons (including the output neuron), what is always true about the backprop algorithm when all the initial weights $w_{ij}^{(\ell)}$ are set to 1?

Your Answer	Score	Explanation
onone of the other choices		
$w_{ij}^{(1)} = w_{(i+1)j}^{(1)}$ for $1 \le i < d^{(0)} - 1$ and all j		
\bigcirc all $w_{j1}^{(2)}$ for $j>0$ are different		
O the gradient components with respect to all $w_{ij}^{(\ell)}$ are zero		
$w_{ij}^{(1)} = w_{i(j+1)}^{(1)}$ for all i and $1 \le j < d^{(1)} - 1$	✓ 20.00	
Total	20.00 /	
	20.00	

Question 13

Experiments with Decision Tree

Implement the simple C&RT algorithm without pruning using the Gini index as the impurity measure as introduced in the class. For the decision stump used in branching, if you are branching with feature i and direction s, please sort all the $x_{n,i}$ values to form (at most) N+1 segments of equivalent θ , and then pick θ within the median of the segment. Run the algorithm on the following set for training:

hw3 train.dat

and the following set for testing:

hw3_test.dat

How many internal nodes (branching functions) are there in the resulting tree G?

Your Answer		Score	Explanation
O 12			
<u> </u>			
<u> </u>			
8	×	-5.00	
<u> </u>			
Total		-5.00 / 20.00	

Question 14

Continuing from Question 13, which of the following is closest to the $E_{\rm in}$ (evaluated with 0/1 error) of the tree?

Score	Explanation
✓ 20.00	
20.00 / 20.00	
	✓ 20.00

Question 15

Continuing from Question 13, which of the following is closest to the $E_{\rm out}$ (evaluated with 0/1 error) of the tree?

Your Answer		Score	Explanation
0.05			
0.25			
0.35			
0.15	~	20.00	
0.00			
Total		20.00 / 20.00	

Question 16

Now implement the Bagging algorithm with N'=N and couple it with your decision tree above to make a preliminary random forest G_{RS} . Produce T=300 trees with bagging. Repeat the

experiment for 100 times and compute average $E_{\rm in}$ and $E_{\rm out}$ using the 0/1 error. Which of the following is true about the average $E_{\rm in}(g_t)$ for all the 30000 trees that you have generated?

Your Answer		Score	Explanation
$\bigcirc 0.09 \le \text{average } E_{\text{in}}(g_t) < 0.12$			
$\odot 0.03 \le \text{average } E_{\text{in}}(g_t) < 0.06$	~	20.00	
$\bigcirc 0.12 \le \text{average } E_{\text{in}}(g_t) < 0.50$			
$\bigcirc 0.00 \le \text{average } E_{\text{in}}(g_t) < 0.03$			
$\bigcirc 0.06 \le \text{average } E_{\text{in}}(g_t) < 0.09$			
Total		20.00 / 20.00	

Question 17

Continuing from Question 16, which of the following is true about the average $E_{\rm in}(G_{RF})$?

Score	Explanation
✓ 20.00	
20.00 / 20.00)
	✓ 20.00

Question 18

Continuing from Question 16, which of the following is true about the average $E_{\text{out}}(G_{RF})$?

Your Answer	Score	Explanation
$\bigcirc 0.03 \le \text{average } E_{\text{out}}(G_{RF}) < 0.06$		

\bullet 0.06 \leq average $E_{\text{out}}(G_{RF}) < 0.09$	✓ 20.00
$\bigcirc 0.00 \le \text{average } E_{\text{out}}(G_{RF}) < 0.03$	
$\bigcirc 0.09 \le \text{average } E_{\text{out}}(G_{RF}) < 0.12$	
\bigcirc 0.12 \leq average $E_{\text{out}}(G_{RF}) < 0.50$	
Total	20.00 / 20.00

Question 19

Now, `prune' your decision tree algorithm by restricting it to have one branch only. That is, the tree is simply a decision stump determined by Gini index. Make a random `forest' G_{RS} with those decision stumps with Bagging like Questions 16-18 with T=300. Repeat the experiment for 100 times and compute average $E_{\rm in}$ and $E_{\rm out}$ using the 0/1 error.

Which of the following is true about the average $E_{\mathrm{in}}(G_{RS})$?

Score	Explanation
✓ 20.00	
20.00 / 20.00)
	✓ 20.00

Question 20

Continuing from Question 19, which of the following is true about the average $E_{\rm out}(G_{RS})$?

Your Answer	Score	Explanation
$\bigcirc 0.06 \le \text{average } E_{\text{out}}(G_{RS}) < 0.09$		
$\bigcirc 0.09 \le \text{average } E_{\text{out}}(G_{RS}) < 0.12$		
$\bigcirc 0.03 \le \text{average } E_{\text{out}}(G_{RS}) < 0.06$		
$0.00 \le \text{average } E_{\text{out}}(G_{RS}) < 0.03$		