CS540 ANSWER SHEET

Name _____ Email _____

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Midterm Examination CS540-1: Introduction to Artificial Intelligence Fall 2014 20 questions, 5 points each

INSTRUCTIONS: Choose ONE answer per question. WRITE YOUR ANSWERS ON THE ANSWER SHEET. WE WILL NOT GRADE ANSWERS ON OTHER PAGES. ON THE ANSWER SHEET WRITE DOWN THE ANSWERS ONLY – DO NOT INCLUDE INTERMEDIATE STEPS OR DERIVATIONS. BE SURE TO INCLUDE YOUR NAME AND EMAIL ON THE ANSWER SHEET, TOO.

1. What would be the output tree of HAC using complete linkage on the following 1D

(B) (-3 ((-1 0) (4 (8 10))))

points: -3, -1, 0, 4, 8, 10? (A) ((-3 (-1 0)) (4 8) 10))

	(*C) ((-3 (- above	1 0)) (4 (8 10)	(D)	(((-3 (-1 0))	4) (8 10))	(E) none of the		
2.	Consider a l	oinary classific	ation task in	\mathbb{R}^2 with the	training data:			
	Class P: $(1,1)$, $(2,1)$, $(3,2)$, $(-1,3)$							
	Class N: (2,	3), (3,3), (-1,-1)					
	Use 3-NN with Euclidean distance. What's the label for test point (2,2), before a after adding another class N training point (2.2, 1.8) to the training set?							
	(A) P, P of ties.	(B) P, N	(C) N, P	(D) N,	N (*E)	Ambiguous because		
3.	3. Let $f()$ be a continuous differentiable function in \mathbb{R} . If the derivative $f'(x) = 0$ at which one of the following is impossible at x ?							
	(A) local mi	inimum	(B) local max	ximum	(C) global min	nimum		
	(D) global n	naximum	(*E) Gradie	nt descent or	f() starting a	t x will move away.		
4.	Which funct	tion is convex of	on \mathbb{R} ?					
	$(A) -x^2$ above	(*B) $-x +$	- 2 (C)	x^3 (D	$\log_2(x^2)$	(E) none of the		
5.	5. Let k-means clustering starts with the centroids (2,3), (4,1). The data points are (7,10), (0,-2), (-3,4), (2,5). After one iteration of k-means (assigning the points to clusters and updating the centroids), what is the distortion?							
	(A) 2	(B) 24	(C) 46	(D) 68	(*E) none of	the above		
6.	6. In \mathbb{R}^4 what is the city-block distance (also known as $L1$ distance or Manhattan distance between two points $(1, 2, 3, 4)$ and $(8, 7, 6, 5)$?					(anhattan distance)		
	(A) 9	(B) 9.1652	(*C) 16	(D) 16.	1652 (E)) none of the above		

	(*E) They all can.						
9.	A decision tree has depth d (a decision tree where the root is a leaf node has $d=1$). All its internal node has c children. The tree is also complete, meaning all leaf nodes are at depth d . If we require each leaf node to contain at least 15 training examples, what is the minimum size of the training set?						
	(*A) $15c^{d-1}$ (B) $15c^d$ (C) $15(1+c+c^2+\ldots+c^{d-1})$						
	(D) $15(1+c+c^2++c^d)$ (E) none of the above						
10.	A bag contains a red ball, a green ball, and a blue ball. Randomly draw a ball from the bag with equal probability. What is the entropy of the outcome?						
	(A) $\log_2(2)$ (*B) $\log_2(3)$ (C) $\log_3(2)$ (D) $\log_3(3)$ (E) none of the above						
11.	. If a feature (i.e. attribute) and the class label are independent (in a probability sense), what is the mutual information between them?						
	(A) negative (*B) 0 (C) positive (D) can be anything (E) none of the above						
12.	A linear SVM in \mathbb{R}^5 has $w = (1, 0, 1, 0, 1)^{\top}$ and $b = -10$. What class does it predict on test point $x = (5, 4, 0, 1, 0)$?						
	(*A) -1 (B) 1 (C) -5 (D) 5 (E) none of the above						
13.	A linear SVM with slack variables has the objective function $\frac{1}{2}w^{\top}w + C\sum_{i}\xi_{i}$. Which statement is FALSE when $C=0$?						
	(*A) The resulting decision boundary can correctly separate all training points. (B) The slack variables ξ_i cannot be negative.						
	(C) There is always a solution w .						
(D) All inequality constraints can be satisfied.							
	(E) It is possible for a training point to be on the decision boundary.						
14.	I. Suppose an SVM has $w = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ and $b = -1$. What is the actual distance between the two planes defined by $w^{\top}x + b = -1$ and $w^{\top}x + b = 1$?						
	3						

7. A test set $(x_1, y_1), \dots (x_{1000}, y_{1000})$ contains labels $y_i = (-1)^i$ for $i = 1 \dots 1000$. A classifier simply predicts y = 1 all the time. What is this classifier's test accuracy?

(D) 1

(C) SVM

(E) none of the above

(D) neural network

(*C) 0.5

8. Which classifier can never produce a linear decision boundary in \mathbb{R}^2 ?

(A) 0

(A) kNN

(B) 0.001

(B) decision tree

	$(A) \ \frac{1}{\sqrt{14}}$	(*B) $\frac{2}{\sqrt{14}}$	$(C) \frac{\sqrt{11}}{2}$	(D) $\sqrt{11}$	(E) none of the above
15.	If $K(x, x')$ is	s a kernel with i	nduced feature	representation	$\phi(x)$, and $G(x, x')$ is another
	kernel with	induced featur	e representation	on $\theta(x)$, then it	t is known that $H(x, x') =$
	aK(x,x') +	bG(x, x') is als	o a kernel if a	a > 0, b > 0. V	What is the induced feature
	representati	on of H ?			

(B) $a\phi(x)+b\theta(x)$ (C) $\begin{pmatrix} a\phi(x) \\ b\theta(x) \end{pmatrix}$ (*D) $\begin{pmatrix} \sqrt{a}\phi(x) \\ \sqrt{b}\theta(x) \end{pmatrix}$ (A) $\sqrt{a}\phi(x) + \sqrt{b}\theta(x)$ (E) none of the above

16. Due to some bizarre reason one of the inputs to a sigmoid perceptron is positive infinity! What output value is possible for this perceptron?

(C) 0.540(D) 1 (A) 0(B) 0.5(*E) all of the above

17. For gradient descent to find the minimum of the function $f(x) = 0.25x^2$, which step size α ensures convergence?

(*B) 2(A) 0(C) 4(D) 6 (E) none of the above

18. Which type of classifier cannot handle multiclass classification problems?

(C) SVM (*E)(A) kNN (B) decision tree (D) neural network They all can.

19. Bag A contains 5 white balls and 7 black balls. Bag B contains 3 whites and 12 black. A fair coin is flipped; if it is Heads, a ball is drawn from bag A, and if it is Tails, a ball is drawn from bag B. Suppose that this experiment was done and a white ball was drawn. What is the probability that this ball was in fact taken from bag B? (i.e. that the coin flip was Tails.)

(A) $\frac{17}{37}$ (C) $\frac{17}{35}$ (D) $\frac{12}{35}$ (*B) $\frac{12}{37}$ (E) none of the above

20. Which of the following is impossible?

(A) $P(A \mid B) > P(A)$ (B) $P(A \mid B) < P(A)$ (D) P(A, B) < P(A) (E) $P(A, B) = P(A \mid B)$ (*C) P(A, B) > P(A)