

Name:

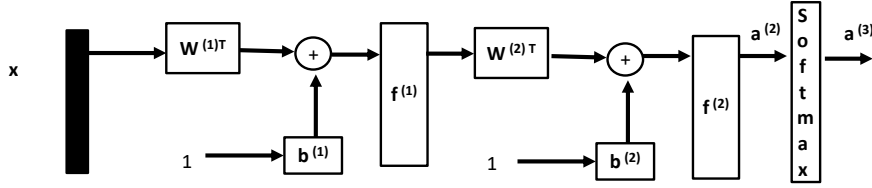
USC ID:

Notes (Please read very carefully):

- Write your name and ID number in the spaces above.
- No books, cell phones or other notes are permitted. Only two letter size cheat sheets (back and front) and a calculator are allowed.
- Problems are not sorted in terms of difficulty. Please avoid guess work and long and irrelevant answers.
- Show all your work and your final answer. Simplify your answer as much as you can.
- Open your exam only when you are instructed to do so.
- The exam has 5 questions, 12 pages, and 25 points extra credit.
- In online exams, legible copies SCANNED via phone applications must be submitted in one pdf file, not pictures of answer sheets.
- Make sure you submit ALL pages of your answers. Answers submitted after the exam is adjourned WILL NOT BE ACCEPTED.

Problem	Score	Earned
1	25	
2	25	
3	25	
4	25	
5	25	
Total	125	

1. Consider the following MLP



where

$$\mathbf{W}^{(1)} = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 2 & 1 \end{bmatrix}$$

$$\mathbf{W}^{(2)} = \begin{bmatrix} -1 & 2 & -1 \\ 1 & 2 & 0 \end{bmatrix}$$

$$\mathbf{b}^{(1)} = \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}, \quad \mathbf{b}^{(2)} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix},$$

Also assume that the components of $\mathbf{f}^{(1)}$, and $\mathbf{f}^{(2)}$ are linear, i.e. is $f_1(x) = f_2(x) = x$. The second layer is followed by a softmax layer, and the output of each neuron in the softmax layer is calculated as:

$$a_j^{(3)} = \frac{e^{a_j^{(2)}}}{e^{a_1^{(2)}} + e^{a_2^{(2)}} + e^{a_3^{(2)}}}$$

where $a_j^{(3)}$ is the output of j^{th} neuron in the third layer and $a_j^{(2)}$ is the output of the j^{th} neuron in the second layer.

Answer the following questions:

- How many neurons are there in the first, and second layers, respectively?
- How many classes (K) does the classification problem performed by this network involve?
- Explain why this network is exactly equivalent to performing multinomial regression using linear logits and write down $p_k(\mathbf{x}) = \Pr(Y = k | \mathbf{X} = \mathbf{x})$ for $k = 1, \dots, K$.

Solution:

2. We are trying to estimate the temperature of consecutive years based on *observations* on tree ring sizes. Possible ring sizes are Very Small = VS, Small = S, Medium = M, Large = L, and Very Large = VL. Years can be Cold = C or Hot = H. Assume that we observed VS, VL tree ring sizes in two consecutive years. Also, Assume that $\pi = [0.3 \ 0.7]$ shows the initial distribution of C and H, respectively. Calculate the probability $P(O)$ that each of following HMMs have given rise to the observation $O = \{VL, S\}$ to determine which one is more likely to be fit to the observation. First rows of A_1, B_1, A_2, B_2 represent C and second rows represent H.

$$(a) \quad \begin{array}{cc} C & H \end{array} \quad \begin{array}{ccccc} VS & S & M & L & VL \end{array}$$

$$A_1 = \begin{bmatrix} 0.7 & 0.3 \\ 0.1 & 0.9 \end{bmatrix} \quad B_1 = \begin{bmatrix} 0.2 & 0.3 & 0.3 & 0.1 & 0.1 \\ 0.2 & 0.2 & 0.2 & 0.3 & 0.1 \end{bmatrix}$$

$$(b) \quad \begin{array}{cc} C & H \end{array} \quad \begin{array}{ccccc} VS & S & M & L & VL \end{array}$$

$$A_2 = \begin{bmatrix} 0.6 & 0.4 \\ 0.4 & 0.6 \end{bmatrix} \quad B_2 = \begin{bmatrix} 0.1 & 0.2 & 0.2 & 0.1 & 0.4 \\ 0.5 & 0.2 & 0.1 & 0.1 & 0.1 \end{bmatrix}$$

Solution:

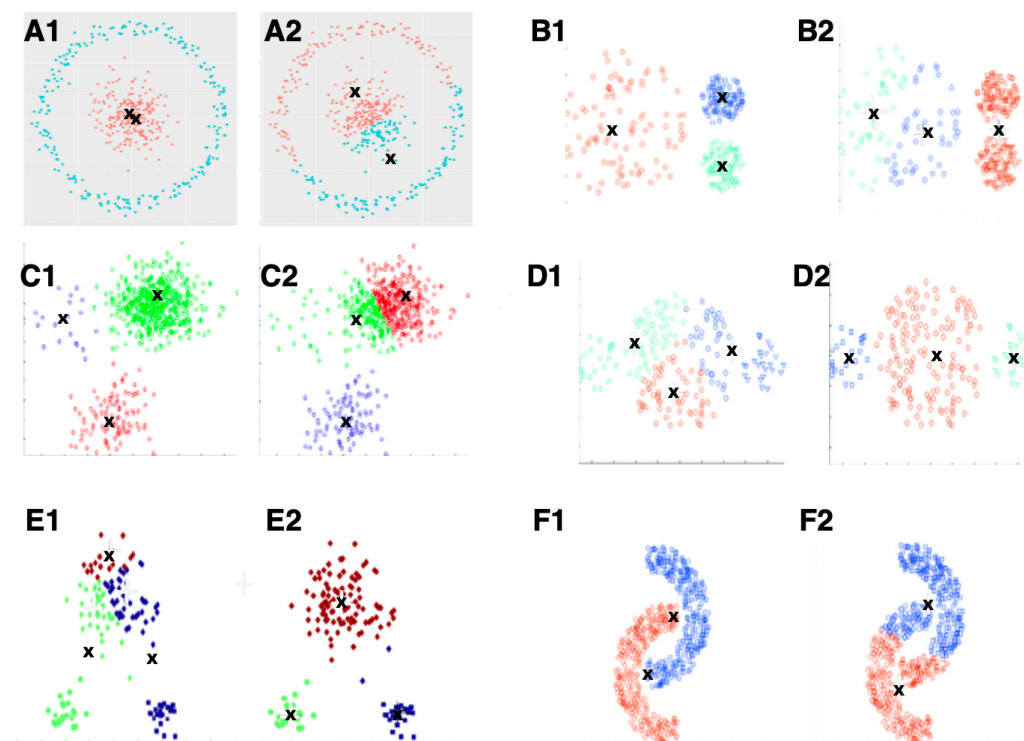
3. Consider a dataset with 3 points in 1-D:

Index	X	Y
1	0	+
2	-1	-
3	+1	-

- (a) Carefully sketch these three training points. Are the classes linearly separable?
- (b) Consider mapping each point to 2-D using new feature vectors $\boldsymbol{\varphi}(x) = [u_1(x), u_2(x)]^T$, in which $u_1(x)$ and $u_2(x)$ are polynomial functions of x . Find a $\boldsymbol{\varphi}(x)$ such that data are linearly separable in the new feature space.
- (c) The maximum margin classifier in the new space has the equation $\mathbf{w}^T \boldsymbol{\varphi}(x) + b = 0$. Find \mathbf{w} and b . Determine the decision boundary in the original 1-D space and the class assigned to $x = \frac{1}{3}$ by the classifier.
- (d) Find a two layer feedforward neural network whose input-output equation is exactly like the above maximum margin classifier, by determining $\mathbf{W}^{(1)}$, $\mathbf{b}^{(1)}$ and $\mathbf{W}^{(2)}$, $\mathbf{b}^{(2)}$, and $\mathbf{f}^{(1)}$ and $\mathbf{f}^{(2)}$.

Hint: in this neural network, elements of $\mathbf{f}^{(1)}$ are *different polynomial functions*.

4. There are 6 different datasets noted as A,B,C,D,E,F. Each dataset is clustered using two different methods, and one of them is K-means. All results are shown in figure below. You are required to determine which result is more likely to be generated by K-means method. (Hint: check the state when K-means converges; Centers for each cluster have been noted as x; Since x and y axis are scaled proportionally, you can determine the distance to centers geometrically). The distance measure used here is the Euclidean distance.

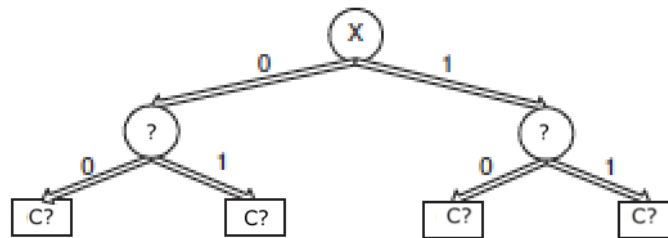


- Dataset A (write A1 or A2)
- Dataset B (write B1 or B2)
- Dataset C (write C1 or C2)
- Dataset D (write D1 or D2)
- Dataset E (write E1 or E2)
- Dataset F (write F1 or F2)

5. Consider the following set of training examples.

X	Y	Z	No. of Class C1 Examples	No. of Class C2 Examples
0	0	0	5	40
0	0	1	0	15
0	1	0	10	5
0	1	1	45	0
1	0	0	10	5
1	0	1	25	0
1	1	0	5	20
1	1	1	0	15

Compute a two-level decision tree, and choose X as the root, as shown in the figure.



Use the classification error rate as the criterion for splitting at the second level and determine whether Y or Z should be used in the second level. To do so, fill in the following tables, assuming that $X = 0$

Y	C1	C2
0		
1		

Z	C1	C2
0		
1		

and the following tables, assuming $X = 1$

Y	C1	C2
0		
1		

Z	C1	C2
0		
1		

The above tables assist in finding the variable splitting which provides better classification error rate for each side of the tree, i.e. for $X = 0$ and $X = 1$. Determine the classes in the leaf nodes. What is the classification error rate of the tree you found on training data?

Solution:

Scratch paper

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