# UNIVERSITY OF ILLINOIS AT URBANA-CHAMPAIGN CS440/ECE448 Artificial Intelligence Exam 1

Spring 2022

February 21, 2022

Your Name:			
Your NetID:			

# **Instructions**

- Please write your name on the top of every page.
- This will be a CLOSED BOOK, CLOSED NOTES exam. You are permitted to bring and use only one 8.5x11 page of hand-written notes, front and back.
- No electronic devices (phones, tablets, calculators, computers etc.) are allowed.
- No calculators are permitted. You need not simplify explicit numerical expressions.

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#### **Possibly Useful Formulas**

**Probability:** 
$$P(B = 1|A = 1) = \frac{P(A = 1, B = 1)}{P(A = 1)}$$

Naïve Bayes: 
$$P(X = x | Y = y) \approx \prod_{i=1}^{n} P(W = w_i | Y = y)$$

**Laplace Smoothing:** 
$$P(w) = \frac{\text{Count}(w) + k}{\sum_{w} \text{Count}(w) + k(1 + \sum_{w} 1)}$$

**Perceptron:** 
$$\vec{w}_y = \vec{w}_y + \eta \vec{x}$$
,  $\vec{w}_{f(\vec{x})} = \vec{w}_{f(\vec{x})} - \eta \vec{x}$ 

**Linear Regression w/SGD:** 
$$\vec{w} \leftarrow \vec{w} - \frac{\eta}{2} \nabla_{\vec{w}} \varepsilon_i^2 = \vec{w} - \eta \varepsilon_i \vec{x}_i$$

$$\textbf{Logistic Regression:} \nabla_{\vec{w}_c} \mathcal{L}_i = \nabla_{\vec{w}_c} \left( -\ln \frac{e^{\vec{w}_{c_i}^T \vec{x}_i}}{\sum_k e^{\vec{w}_k^T \vec{x}_i}} \right) = \left( \frac{e^{\vec{w}_{c_i}^T \vec{x}_i}}{\sum_k e^{\vec{w}_k^T \vec{x}_i}} - y_{i,c} \right) \vec{x}_i$$

$$\textbf{Neural Net:} \ \ \xi_j^{(l)} = b_j^{(l)} + \sum_k w_{j,k}^{(l)} h_k^{(l-1)}, \quad h_j^{(l)} = g^{(l)} \left( \xi_j^{(l)} \right)$$

$$\textbf{Back-Propagation:} \ \, \frac{\partial \mathscr{L}}{\partial h_k^{(l-1)}} = \sum_j \frac{\partial \mathscr{L}}{\partial h_j^{(l)}} \frac{\partial h_j^{(l)}}{\partial h_k^{(l-1)}}$$

Pinhole Camera: 
$$\frac{x'}{f} = -\frac{x}{z}, \quad \frac{y'}{f} = -\frac{y}{z}$$

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# Question 1 (7 points)

Consider two binary random variables, X and Y. Suppose that

$$P(X=1)=a$$

$$P(Y=1)=b$$

$$P(X = 1, Y = 0) = c$$

In terms of a, b, and/or c, what is P(Y = 1|X = 1)?

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Question 2 (7 points)

You've been asked to create a naïve Bayes model of the candy produced by the Santa Claus Candy Company. As your training dataset, you've been given a box containing 80 pieces of candy, of which 8 are strawberry, 48 are raspberry, and 24 are blueberry. In terms of the Laplace smoothing parameter k, estimate the following probabilities:

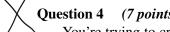
P(flavor=strawberry|Santa Claus Candy Company)=

P(flavor=raspberry|Santa Claus Candy Company)=

P(flavor=blueberry|Santa Claus Candy Company)=

P(flavor=other|Santa Claus Candy Company)=

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Describe, in one sentence each, the purpose of (1) a training set, (2) a development test set, (3) evaluation test set.	an



You're trying to create a multi-class perceptron that will classify animals as being either fish, birds, or reptiles. Your feature vector is  $\vec{x} = [x_1, x_2, x_3, 1]^T$ , where

 $x_1$  = fraction of time the animal spends under water

 $x_2$  = fraction of time the animal spends on land

 $x_3$  = fraction of time the animal spends flying

• Based on your extensive prior knowledge of zoology, you initialize your perceptron with the following weight vectors:  $\vec{w}_{fish} = [1,0,0,0]^T$ ,  $\vec{w}_{reptile} = [0,1,0,0]^T$ , and  $\vec{w}_{bird} = [0,0,1,0]^T$ .

• Your first training token is a crocodile, for which y = reptile, and  $\vec{x} = [0.7, 0.3, 0, 1]^T$ .

After training with this training token, what are the numerical values of  $\vec{w}_{fish}$ ,  $\vec{w}_{reptile}$ , and  $\vec{w}_{bird}$ ? Assume a learning rate of  $\eta = 1$ .

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### Question 5 (7 points)

In stochastic gradient descent, we train using one training token at a time. Suppose

$$\mathscr{L} = (\vec{w}^T \vec{x} - y)^2$$

$$\vec{w} = \begin{bmatrix} w_1 \\ w_2 \\ b \end{bmatrix}, \quad \vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ 1 \end{bmatrix}$$

In terms of  $\vec{x}$ ,  $\vec{w}$ ,  $w_1$ ,  $w_2$ , b,  $x_1$ ,  $x_2$ , and/or y, what is  $\frac{d\mathcal{L}}{dw_2}$ ?

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### Question 6 (7 points)

Suppose that

$$f = w_{1,1}^{(2)} h_1 + w_{1,2}^{(2)} h_2 + b^{(2)}$$

$$h_1 = \text{ReLU}\left(w_{1,1}^{(1)} x_1 + w_{1,2}^{(1)} x_2 + b_1^{(1)}\right)$$

$$h_2 = \text{ReLU}\left(w_{2,1}^{(1)} x_1 + w_{2,2}^{(1)} x_2 + b_2^{(1)}\right)$$

Assume, for a particular training token, that  $h_1 > 0$  and  $h_2 > 0$ . For that particular training token, what is  $\frac{\partial f}{\partial w_{1,1}^{(l)}}$ ? Express your answer in terms of  $x_j$ ,  $h_j$ ,  $w_{j,k}^{(l)}$ , and/or  $b_k^{(l)}$  for any values of j, k, and/or l that may be useful to you.

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Question 7 (7 points	Question	7	(7 poir	nts)
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In the real world, the (x,y,z) coordinates of Joe's face and Mike's face are (14,3,7) and (14,3,17), respectively, where z is distance from the camera. In the image plane (x',y'), which person (Joe or Mike) is closer to the center of the image (the point (x',y')=(0,0)), and why?