

UNIVERSITY OF ILLINOIS AT URBANA-CHAMPAIGN  
CS440/ECE448 Artificial Intelligence  
**Conflict Exam 1**  
Spring 2026

Main Exam is February 16, 2026

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**Your Name:** \_\_\_\_\_

**Your NetID:** \_\_\_\_\_

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**Instructions**

- Please write your name on the top of every page.
- Have your ID ready; you will need to show it when you turn in your exam.
- This will be a **CLOSED BOOK, CLOSED NOTES** exam. You are permitted to bring and use only one 8.5x11 page of notes, front and back, handwritten or typed in a font size comparable to handwriting.
- No electronic devices (phones, tablets, calculators, computers etc.) are allowed.
- Make sure that your answer includes only the variables that it should include, but **DO NOT** simplify explicit numerical expressions. For example, the answer  $x = \frac{1}{1+\exp(-0.1)}$  is **MUCH** preferred (much easier for us to grade) than the answer  $x = 0.524979$ .

**Possibly Useful Formulas**

$$P(X = x|Y = y)P(Y = y) = P(Y = y|X = x)P(X = x)$$

$$P(X = x) = \sum_y P(X = x, Y = y)$$

$$E[f(X, Y)] = \sum_{x,y} f(x, y)P(X = x, Y = y)$$

$$\text{Precision, Recall} = \frac{TP}{TP + FP}, \frac{TP}{TP + FN}$$

$$\text{MPE=MAP: } f(x) = \arg \max (\log P(Y = y) + \log P(X = x|Y = y))$$

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

$$\text{Laplace Smoothing w/OOV: } P(W = w_i) = \frac{k + \text{Count}(W = w_i)}{k + \sum_v (k + \text{Count}(W = v))}$$

$$\text{Fairness: } P(Y|A) = \frac{P(Y|f(X), A)P(f(X)|A)}{P(f(X)|Y, A)}$$

$$\text{Score: } v_{t+1}(j) = \max_i v_t(i) + \log a_{ij} + \log b_j(x_{t+1})$$

$$\text{Backpointer: } \psi_{t+1}(j) = \arg \max_i v_t(i) + \log a_{ij} + \log b_j(x_{t+1})$$

$$\text{Linear Regression: } \mathcal{L} = \frac{1}{n} \sum_{i=1}^n (\mathbf{w}^T \mathbf{x}_i - y_i)^2$$

**Question 1 (26 points)**

Lochs, lagoons, and lakes are all bodies of water, but they tend to be inhabited by different types of supernatural creatures. Let  $S$  be the type of supernatural creature in a body of water (1 =dinosaur, 2 =merman, 3 =fairy), and let  $L$  be the class label (1 =loch, 2 =lagoon, 3 =lake). Define  $P(L = \ell)$  to be the prior probability of class  $\ell$ , and  $P(S = s|L = \ell)$  to be the likelihood of creature  $s$  in setting  $\ell$ .

- (a) (13 points) A Bayes classifier observes  $m$ , then chooses the value of  $\ell$  with the highest *a posteriori* probability. In terms of the distributions  $P(L)$  and  $P(S|L)$ , what is the error rate of the Bayes classifier? Your answer may contain explicit summations, minimizations, and/or maximizations; please specify the variable(s) over which you are summing, maximizing, and/or minimizing.
- (b) (13 points) The famous explorer Ibn Battuta visited 5 lochs. He found that 3 lochs were inhabited by dinosaurs, while 2 were inhabited by mermen. Assuming that every body of water is inhabited by exactly one supernatural creature of class  $s \in \{1, 2, 3\}$  (assume that exceptions to this rule are impossible), in terms of a Laplacian smoothing parameter, estimate  $P(S = s|L = 1)$  for all  $s \in \{1, 2, 3\}$ .

**Question 2 (30 points)**

Consider three binary events  $A$ ,  $B$ , and  $C$ , related by the Bayes network shown here:



The parameters of this Bayes network are given by the unknown constants  $v$  through  $z$ , as follows:

$$\begin{aligned} P(A) &= v \\ P(C|A) &= w, \quad P(C|\neg A) = x \\ P(B|C) &= y, \quad P(B|\neg C) = z \end{aligned}$$

- (a) (10 points) Are events  $A$  and  $B$  independent, conditionally independent given knowledge of  $C$ , both, or neither? Explain your answer.
- (b) (10 points) Suppose  $C$  is unknown. Find the four probabilities  $P(A, B)$ ,  $P(A, \neg B)$ ,  $P(\neg A, B)$ , and  $P(\neg A, \neg B)$ .

- (c) (10 points) Suppose  $C$  is known to be true. Find  $P(C)$ , and in terms of  $P(C)$  find  $P(A, B|C)$ ,  $P(A, \neg B|C)$ ,  $P(\neg A, B|C)$ , and  $P(\neg A, \neg B|C)$ .

**Question 3 (20 points)**

Consider an HMM with binary state variable  $Y_t$ , binary observation variable  $X_t$ , transition probabilities  $a_{ij} = P(Y_{t+1} = j | Y_t = i)$ , and observation probabilities  $b_{ij} = P(X_t = j | Y_t = i)$ , and suppose you know that  $Y_0 = 0$ . Suppose there is some normalizer,  $z$ , and base,  $c$ , such that  $\log_c(a_{ij}/z) = \log_c(b_{ij}/z) = 0$  for  $i = j$ , but for  $i \neq j$ ,  $\log_c(b_{ij}/z) = -2$  and  $\log_c(a_{ij}/z) = -3$ . Suppose that  $\{X_1, \dots, X_9\} = \{1, 0, 0, 1, 1, 1, 0, 0, 1\}$ . Draw a trellis specifying the numerical values of  $v_t(i) = \max_{Y_{\tau}, \tau < t} \log_c P(Y_1, \dots, Y_t = i | X_1, \dots, X_t) / z$  for  $i \in \{0, 1\}$  and  $1 \leq t \leq 9$ , and specify a sequence of state variables  $\{Y_1, \dots, Y_9\}$  that maximizes the score. If there are more than one state sequences with the same maximum score, you only need to provide one of them.

**Question 4 (24 points)**

Consider a linear regression model  $f(\mathbf{x}_i) = \mathbf{w}^T \mathbf{x}_i$  where  $\mathbf{x}_i$  is an  $m$ -dimensional vector drawn from a size- $n$  training corpus,  $1 \leq i \leq n$ . Suppose you want to choose the vector  $\mathbf{w}$  to minimize  $\mathcal{L}_{\text{train}}$ , defined as

$$\mathcal{L}_{\text{train}} = \frac{1}{n} \sum_{i=1}^n |f(\mathbf{x}_i) - y_i|^5,$$

where  $y_i$  is a real-valued scalar for  $1 \leq i \leq n$ .

(a) (12 points) Find the gradient  $\frac{\partial \mathcal{L}_{\text{train}}}{\partial \mathbf{w}}$ .

(b) (12 points) Suppose you also have a development corpus, with tokens numbered  $n+1 \leq i \leq n+j$ , and a test corpus, with tokens numbered  $n+j+1 \leq i \leq n+j+k$ . The losses in these corpora are

$$\mathcal{L}_{\text{dev}} = \frac{1}{j} \sum_{i=n+1}^{n+j} |f(\mathbf{x}_i) - y_i|^5,$$

$$\mathcal{L}_{\text{test}} = \frac{1}{k} \sum_{i=n+j+1}^{n+j+k} |f(\mathbf{x}_i) - y_i|^5.$$

Your client, a hardware manufacturer, is very clever about finding things to measure. They propose the following experiment: (1) increase  $m$  by finding something else to measure about each product, (2) re-train the new  $\mathbf{w}$ , finding the value that minimizes  $\mathcal{L}_{\text{train}}$ , (3) measure  $\mathcal{L}_{\text{dev}}$  and  $\mathcal{L}_{\text{test}}$ , (4) repeat.

Do you expect  $\mathcal{L}_{\text{test}}$  to be smallest when  $m = 1$ , when  $m = \frac{n}{2}$ , or when  $m = \frac{3n}{2}$ ? Why?

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