

VE 492 Homework8

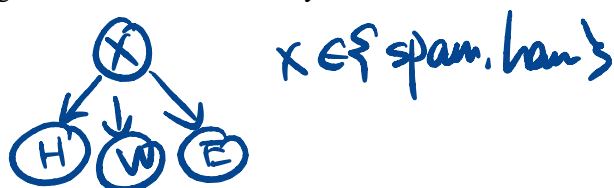
Due: 23:59, July 28

Q1. Naive Bayes

Your friend claims that he can write an effective Naive Bayes spam detector with only three features: the hour of the day that the email was received ($H \in \{1, 2, \dots, 24\}$), whether it contains the word 'viagra' ($W \in \{yes, no\}$), and whether the email address of the sender is Known in his address book, Seen before in his inbox, or Unseen before ($E \in \{K, S, U\}$).

(a) Flesh out the following information about this Bayes net:

Graph structure:



Parameters:

Size of the set of parameters: 53

Suppose now that you labeled three of the emails in your mailbox to test this idea:

spam or ham?	H	W	E
spam	3	yes	S
ham	14	no	K
ham	15	no	K

$$P(E=S | X=\text{spam})=1$$

$$P(W=\text{yes} | X=\text{spam})=1$$

$$P(E=K | X=\text{ham})=1$$

(b) Use the three instances to estimate the maximum likelihood parameters.

(c) Using the maximum likelihood parameters, find the predicted class of a new datapoint with $H = 3, W = no, E = U$. **No predicted class**

(d) Now use the three to estimate the parameters using Laplace smoothing and $k = 2$. Do not forget to smooth both the class prior parameters and the feature values parameters.

(e) Using the parameters obtained with Laplace smoothing, find the predicted class of a new datapoint with $H = 3, W = no, E = U$. **ham**

(f) You observe that you tend to receive spam emails in batches. In particular, if you receive one spam message, the next message is more likely to be a spam message as well. Explain a new graphical model which most naturally captures this phenomena.

Graph structure:



Parameters:

$$P(X_0=\text{spam}), P(X_t=\text{spam} | X_{t-1}), P(H_t | X_t), P(W_t=\text{yes} | X_t), P(E_t | X_t).$$

Size of the set of parameters:

Q2. Perceptron

- (a) Suppose you have a binary perceptron in 2D with weight vector $\mathbf{w} = r [w_1, w_2]^T$. You are given w_1 and w_2 , and are given that $r > 0$, but otherwise not told what r is. Assume that ties are broken as positive. Can you determine the perceptron's classification of a new example x with known feature vector $f(x)$?

- ☒ A. Always
B. Sometimes
C. Never

- (b) Now you are learning a multi-class perceptron between 4 classes. The weight vectors are currently $[1, 0]^T, [0, 1]^T, [-1, 0]^T, [0, -1]^T$ for the classes A, B, C, and D. The next training example x has a **label of A** and feature vector $f(x)$.

For the following questions, *do not make any assumptions about tie-breaking*. (Do not write down a solution that creates a tie.)

If the answer does not exist, write down **Not possible**

$f(x) = \begin{bmatrix} \\ \end{bmatrix}$ ☐ Not possible

- (i) Write down a feature vector in which no weight vectors will be updated.

- (ii) Write down a feature vector in which **only** \mathbf{w}_A will be updated by the perceptron.

- (iii) Write down a feature vector in which **only** \mathbf{w}_A and \mathbf{w}_B will be updated by the perceptron.

- (iv) Write down a feature vector in which **only** \mathbf{w}_A and \mathbf{w}_C will be updated by the perceptron.

$f(x) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$
Not possible
 $f(x) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$
 $f(x) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$

The weight vectors are the same as before, but now there is a bias feature with value of 1 for all x and the weight of this bias feature is 0, -2, 1, -1 for classes A, B, C, and D respectively. As before, the next training example x has a **label of A** and a feature vector $f(x)$. The always "1" bias feature is the first entry in $f(x)$.

If the answer does not exist, write down **Not possible**

$f(x) = \begin{bmatrix} 1 \\ \end{bmatrix}$ ☐ Not possible

- (v) Write down a feature vector in which **only** \mathbf{w}_B and \mathbf{w}_C will be updated by the perceptron.

- (vi) Write down a feature vector in which **only** \mathbf{w}_A and \mathbf{w}_C will be updated by the perceptron.

Not possible
 $f(x) = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$