

# ECE/CS/ME 539: Intro to Artificial Neural Networks

## Homework 2

### Problem 1: Maximum A Posteriori Classification

**(4pts)** Suppose you want to classify marathon runners participating on the next Madison marathon into two categories – professionals (pro), and amateurs (amateur) – based on their marathon times. After some research, you found that marathon times among pros follow a Gaussian distribution with mean  $\mu_{\text{pro}} = 2.5$  and standard deviation  $\sigma_{\text{pro}} = 0.2$  and amateurs times follow a Gaussian distribution with mean  $\mu_{\text{amateur}} = 4$  and standard deviation  $\sigma_{\text{amateur}} = 0.5$ . Furthermore, you know that in the Madison marathon, only about 10% of the runners are professional. You may find `scipy.stats.norm` useful to solve the problem computationally.

- (a) **(1pt)** Compute the likelihood of three participant’s times provided that they were either amateurs or pros.

Participant’s time (t)	2.5	3	3.5
$P(t \text{pro})$			
$P(t \text{amateur})$			

- (b) **(1pt)** For each of the three participants in the table above, compute the posterior probabilities  $P(\text{pro}|t)$  and  $P(\text{amateur}|t)$ .

Participant’s time (t)	2.5	3	3.5
$P(\text{pro} t)$			
$P(\text{amateur} t)$			

- (c) **(1pt)** Apply the MAP decision rule to decide which participants are pros and which are amateurs.

Participant’s time (t)	2.5	3	3.5
MAP Decision			

- (d) **(1pt)** Estimate the probability of misclassification among amateurs and among pro runners through simulation. To accomplish this, begin by sampling  $N = 1000$  marathon times for each class from the underlying Gaussian distributions (sampling function is already provided in file “hw2\_starter\_code.ipynb”). After classification, count the number of mistakes made by the MAP classifier.

## Problem 2: Confusion Matrix

**(3pts)** A new test is developed to detect spam emails. After an experiment, the confusion matrix of this test is reported below:

Actual / Predicted	P	N
P	15	5
N	4	16

- (a) **(1pt)** How many spam (P) results does this new test report? \_\_\_\_\_
- (b) **(1pt)** What percentage of actual spam emails are correctly identified as spam in this test?  
\_\_\_\_\_
- (c) **(1pt)** What is the false positive rate, defined as the fraction of products that are reported as defective but are actually non-defective, among all negative tests? \_\_\_\_\_

## Problem 3: Performance Metrics

**(3pts)** Consider 10 feature-label pairs  $(x(k), y(k); 1 \leq k \leq 10)$ . Assume that  $y(k) \in \{0, 1\}$ . The posterior probability  $P(y(k) = 1|x(k))$  is given in the 2nd row of the table below, and the corresponding ground truth label is given in the third row. Here, a class label = 0 means Negative, and a class label = 1 means Positive.

Index (k)	1	2	3	4	5	6	7	8	9	10
$P(y(k) = 1 x(k))$	0.05	0.15	0.40	0.55	0.25	0.45	0.48	0.62	0.67	0.75
True Label	0	0	0	0	1	1	1	1	1	1
Predicted label $y(k)$										

Given a threshold  $b$ , we set the predicted label  $y(k) = 0$  if  $P(y(k) = 1|x(k)) \leq b$ ; and  $= 1$  otherwise.

- (a) **(0.5pt)** If  $b = 0.3$ , fill in the predicted label in the 4th row of the above table.
- (b) **(0.5pt)** Compute the confusion matrix  $C$  with  $b = 0.3$ .
- (c) **(1pt)** With  $b = 0.3$ , compute the following quantities: sensitivity (sen), specificity (spe), Pr. False Alarm (pfa), Pr. Miss (pmis), precision (pre), recall, and accuracy.
- (d) **(1pt)** For the value of threshold  $b$  varying from 0 to 1, compute the list of distinct pairs of (TPR, FPR) and then plot the ROC curve and calculate the area under the ROC curve (AUC).