

## CptS 570 Machine Learning: Sample Questions

### 1. Short Questions

Please keep your answers short (at most two sentences). For the True / False questions, please also provide a short justification.

- a) The ID3 decision tree learning algorithm is guaranteed to find the optimal decision tree (True / False)
- b) We would expect the support vectors to remain the same in general as we move from a linear kernel to higher order polynomial kernels (True / False)
- c) Why does the kernel trick allow us to solve SVMs with high dimensional feature spaces, without significantly increasing the running time?
- d) Your professor wants to automatically classify PhD applications into good/bad categories, and also wants to detect outliers (applications not in his research area) using density estimation. Will you recommend him a discriminative or generative classifier? Why?
- e) Your professor who is new to WSU wants to automatically classify PhD applications who are likely to become great researchers after they graduate. However, he has a small amount of training data. To create the most accurate classifier, will you recommend him a discriminative or generative classifier? Why?

### 2. Decision Trees

Suppose we want to learn a decision tree for predicting whether a person is *happy* or *sad* based on the color of their shirt, whether they wear a tie, and the number of eyes they have.

**Table 1:** Training data for decision tree learning.

COLOR	TIE	NO. OF EYES	(OUTPUT) EMOTION
Green	Yes	2	Sad
Green	No	2	Sad
Green	No	2	Sad
Blue	No	2	Sad
Blue	No	2	Happy
Red	No	2	Happy
Red	No	2	Happy
Red	No	2	Happy
Red	Yes	3	Happy

- a) What is the conditional entropy  $H(\textit{Emotion}|\textit{Tie} = \textit{yes})$ ?
- b) What is the conditional entropy  $H(\textit{Emotion}|\textit{Eyes} = 3)$ ?
- c) Which feature would be selected first (root of the tree) by the decision-tree learning algorithm?
- d) Draw the full decision tree (without any pruning) that would be learned from this data?

### 3. Bayes Rule

- a) I give you the following fact:  $P(A|B) = 2/3$ . Do you have enough information to compute  $P(B|A)$ ? If not, write “not enough information.” If yes, compute the value of  $P(B|A)$ .
- b) I give you the following facts:  $P(A|B) = 2/3$ ,  $P(A|\neg B) = 1/3$ . Do you have enough information to compute  $P(B|A)$ ? If not, write “not enough information.” If yes, compute the value of  $P(B|A)$ .
- c) I give you the following facts:  $P(A|B) = 2/3$ ,  $P(A|\neg B) = 1/3$ ,  $P(B) = 1/3$ . Do you have enough information to compute  $P(B|A)$ ? If not, write “not enough information.” If yes, compute the value of  $P(B|A)$ .
- d) I give you the following facts:  $P(A|B) = 2/3$ ,  $P(A|\neg B) = 1/3$ ,  $P(B) = 1/3$ ,  $P(A) = 4/9$ . Do you have enough information to compute  $P(B|A)$ ? If not, write “not enough information.” If yes, compute the value of  $P(B|A)$ .

### 4. Parameter Estimation

- a) SVM classifier tend to be robust to irrelevant features. Suppose we run SVM classifier with features  $X_1, X_2, \dots, X_n$  and then add a irrelevant feature  $X_{n+1}$  that cannot help increase the margin. How will SVM classifier automatically ignore this feature? Justify your answer.