

## HOMEWORK 2: WRITTEN EXERCISE PART

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### 1 Information Theory [25/4 pts]

Suppose  $X, Y$  are two random variables taking values in a discrete finite set  $V$ . Let  $H(Y)$  denote the entropy of  $Y$ , and let  $H(Y|X)$  denote the conditional entropy of  $Y$  conditioned on  $X$ . Prove that if  $X, Y$  are independent, then  $H(Y) = H(Y|X)$ .

As we know,  $H(X) = - \sum_{x \in \{x_1, x_2, \dots, x_k\}} P(X = x) \log_2 P(X = x)$

and,  $H(Y|X) = - \sum_{x \in \{x_1, x_2, \dots, x_k\}} P(X = x) H(Y|X = x)$

now,

$$H(Y|X = x) = - \sum_{y \in \{y_1, y_2, \dots, y_k\}} P(Y = y|X = x) \log_2 P(Y = y|X = x)$$

given,  $X, Y$  are independent,

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

$$\implies H(Y|X = x) = - \sum_{y \in \{y_1, y_2, \dots, y_k\}} P(Y = y) \log_2 P(Y = y)$$

$$\implies H(Y|X = x) = H(Y)$$

$$\implies H(Y|X) = \sum_{x \in \{x_1, x_2, \dots, x_k\}} P(X = x) H(Y)$$

$$\text{as, } \sum_{x \in \{x_1, x_2, \dots, x_k\}} P(X = x) = 1$$

$$H(Y|X) = H(Y)$$

### 2 Standardizing Numeric Features [25/4 pts]

Standardize the data set with four points in 2 dimension:  $(7, 7), (3, 7), (3, 3), (7, 3)$ .

Given,  $(7, 7), (3, 7), (3, 3), (7, 3)$

Let us first calculate the mean and standard deviation for this set.

$$\text{Mean}(x) = \mu(x) = \frac{7+3+3+7}{4}$$

$$\implies = \frac{20}{4} = 5$$

$$\text{Mean}(y) = \mu(y) = \frac{7+7+3+3}{4}$$

$$\implies = \frac{20}{4} = 5$$

$$\text{Std. Deviation}(x) = \sigma(x) = \sqrt{\frac{(7-5)^2 + (3-5)^2 + (3-5)^2 + (7-5)^2}{4}}$$

$$\implies = \sqrt{\frac{4+4+4+4}{4}} = 2$$

$$\text{Std. Deviation (y)} = \sigma(y) = \sqrt{\frac{(7-5)^2 + (3-5)^2 + (3-5)^2 + (7-5)^2}{4}}$$

$$\implies = \sqrt{\frac{4+4+4+4}{4}} = 2$$

For standardization of a set of points:  $(x')_i = (x_i - \mu_i)/\sigma$

$$\begin{aligned}\implies x_1 &= \frac{7-5}{2} = 1, y_1 = \frac{7-5}{2} = 1 \\ \implies x_2 &= \frac{3-5}{2} = -1, y_1 = \frac{7-5}{2} = 1 \\ \implies x_3 &= \frac{3-5}{2} = -1, y_1 = \frac{3-5}{2} = -1 \\ \implies x_4 &= \frac{7-5}{2} = 1, y_1 = \frac{3-5}{2} = -1\end{aligned}$$

Standardized points:

$$(1, 1), (-1, 1), (-1, -1), (1, -1)$$

### 3 $k$ -Nearest Neighbors [25/4 pts]

Consider the training data set  $x_1 = (7, 7), y_1 = 0; x_2 = (3, 7), y_2 = 1; x_3 = (3, 3), y_3 = 1; x_4 = (7, 3), y_4 = 2$ . Suppose the Manhattan distance is used. What is the label for  $x = (0, 0)$  in the following settings? Show the calculation steps.

- 1-nearest neighbors.
- 3-nearest neighbors.
- 3-nearest neighbors, distance weighted. The weight for the  $i$ -th neighbor  $z$  is  $1/d(x, z)^2$ .

Manhattan Distance =  $|x - x_i|$  for  $x_1, x_2, x_3, x_4$

$$d(x, x_1) = |0 - 7| + |0 - 7| = 14$$

$$d(x, x_2) = |0 - 3| + |0 - 7| = 10$$

$$d(x, x_3) = |0 - 3| + |0 - 3| = 6$$

$$d(x, x_4) = |0 - 7| + |0 - 3| = 10$$

- 1-nearest neighbors. For this,  $d_3$  is the lowest distance of all.  
 $\implies$  So nearest neighbors =  $x_3$ .  
 $\implies$  Label for  $x = y_3 = 1$
- 3-nearest neighbors. For this,  $d_2, d_3, d_4$  are the lowest distance of all.  
 $\implies$  So 3nearest neighbors  $x_2, x_3, x_4$ .  
 $\implies$  Label for  $x = \sum_{i=2}^4 \gamma(v, y_i); v \in \{1, 2, 3\}$   
 $\implies$  Label for  $x = 1$
- 3-nearest neighbors, distance weighted.

$$\text{As we know, } w_i = \frac{1}{d(x, x_i)^2}$$

$$w_2 = \frac{1}{d(x, x_2)^2} = \frac{1}{10^2} = \frac{1}{100}$$

$$w_3 = \frac{1}{d(x, x_3)^2} = \frac{1}{6^2} = \frac{1}{36}$$

$$w_4 = \frac{1}{d(x, x_4)^2} = \frac{1}{10^2} = \frac{1}{100}$$

$$\begin{aligned}\implies \text{label} &= \begin{bmatrix} \frac{1}{100} \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} \frac{1}{36} \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{1}{100} \\ 0 \end{bmatrix} \\ &= 1\end{aligned}$$

## 4 Performance Measurements [25/4 pts]

Consider the following confusion matrix for 2 classes.

	actual positive	actual negative
predict positive	76	18
predict negative	24	82

Compute the accuracy, error, true positive rate, false positive rate, precision, and recall.

given,

True Positive(TP) = 76

False Positive(FP) = 18

True Negative(TN) = 82

False Negative(FN) = 24

$$\text{Accuracy} = \frac{TP+TN}{TP+FP+FN+TN} = \frac{76+82}{200} = 0.79$$

$$\text{Error} = 1 - \text{Accuracy} = 1 - 0.79 = 0.21$$

$$\text{True positive rate} = \frac{TP}{\text{actual positive}} = \frac{TP}{TP+FN} = \frac{76}{76+24} = 0.76$$

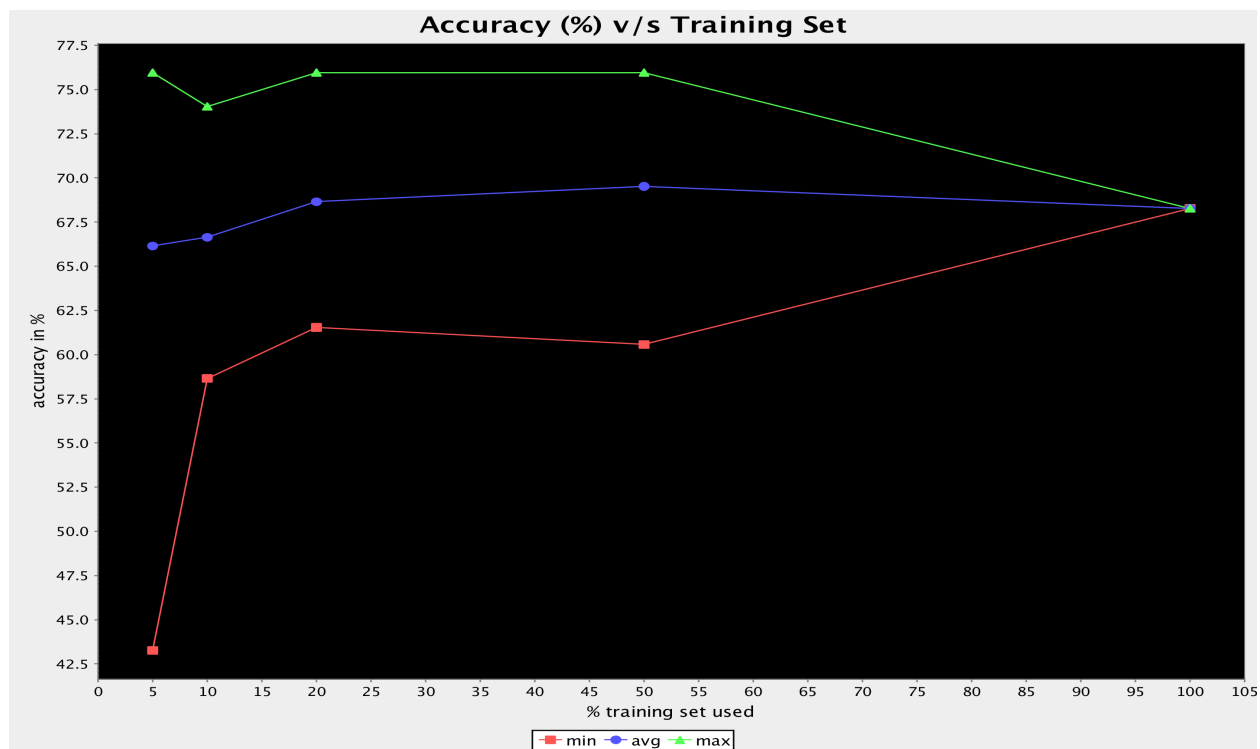
$$\text{False positive rate} = \frac{FP}{\text{actual negative}} = \frac{FP}{FP+TN} = \frac{18}{18+82} = 0.18$$

$$\text{Precision} = \frac{TP}{\text{predicted negative}} = \frac{TP}{TP+FP} = \frac{76}{76+18} = 0.80$$

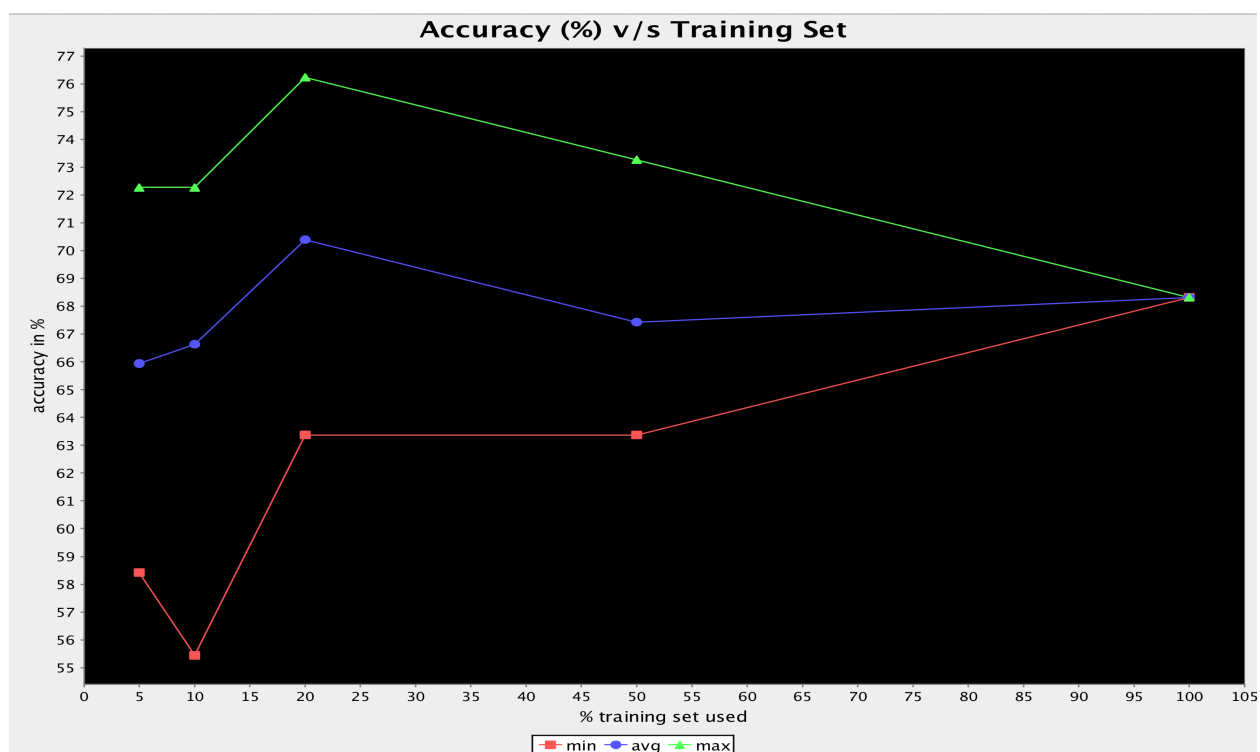
$$\text{Recall} = \frac{TP}{\text{actual positive}} = \frac{TP}{TP+FN} = \frac{76}{76+24} = 0.76$$

## 5 Graphs for Part 2 and 3

In both the figures (on next page) for part 2, the accuracy converges as the % of training data used is 100. The same is because, with 100% data, the model is always constant and hence will provide the same set of results for the given testing data.



for HEART data



for DIABETES data

