

***In The Name of Allah***  
***Pattern Recognition (Autumn 2018)***  
***Instructor: Mahdi Yazdian***  
***Practical Exercise#2: Bayesian Decision Theory***

Due Date: 97.08.06

Perform this homework using MALTAB or Python.

**Implementation guide**

- Implement a function named as 'CreateBayesModel':

```
[model] = CreateBayesModel (method, Likelihoodfncs, priors, costs)
{
    //insert your code
}
```

inputs:

- *method*: it would '*Bayes*' or '*MAP*' or '*ML*'.
- *Likelihoodfncs*: a vector ( $1 \times n$ ) of handler functions. The  $i^{th}$  element shows the likelihood function of the  $i^{th}$  class ( $n$  shows #class).
- *priors*: a vector ( $1 \times n$ ) that consists of the prior of each class.
- *costs*: a matrix ( $n \times n$ ). The element  $< i, k >$  represents the cost of choosing class  $C_i$  when class  $C_k$  is the true class

output:

- *model*: a structure including all parameters of the specified *method*.

Note: Evidently some inputs are not needed according to the selected *method*.

- Implement a function named as 'ClassifyByBayes':

```
[labels, scores] = ClassifyByBayes(model, samples)
{
    //insert your code
}
```

inputs:

- *model*: the model used for classification (created by CreateBayesModel)
- *samples*: a ( $d \times m$ ) matrix consists of  $m$  d-dim samples when  $m$  shows #samples and  $d$  shows the input dimension, i.e., #features.

outputs:

- *labels*: a ( $1 \times m$ ) vector determines the label of each input sample.
- *scores*: a ( $n \times m$ ) vector determined the score samples at each class when  $n$  shows #class.

1. Describe the input of 'CreateBayesModel' when you want to have (a) Bayes (b) MAP and (c) ML decision rule.

2. **Binary classification**

- Download "tic-tac-toe" dataset from [tic-tac-toe](#).
- Using your above implementations, classify the dataset using Bayes (with cost matrix  $\begin{bmatrix} 0 & 2 \\ 1 & 0 \end{bmatrix}$ ), MAP and ML.

**Guide:** Use the dataset to estimate the prior and likelihood function of each class.

- a) Report the confusion matrix of each method.
- b) For each method, report the accuracy of each class and the averaged accuracy.
- c) Compare and discuss about the results.

3. **Multi-class classification**

- Download dataset from <https://archive.ics.uci.edu/ml/datasets/glass+identification>
- Using your above implementations, classify the dataset using MAP and ML.

**Guide:** Use the datasets to estimate the prior and likelihood function of each class.

- a) Report the confusion matrix of each method.
- b) For each method, report the accuracy of each class and the averaged accuracy.
- c) Compare and discuss about the results.

4. **Hints: In previous sections (2&3):**

- a) Check if the covariance matrixes are singular?
- b) If the covariance matrixes are singular, what problem happens?
- c) Offer solutions for resolve problem in section b.
- d) Implement one of your suggested solutions and evaluate the results. **[optional]**

**Note:**

- Prepare a report in PDF format including the figures, answer to the questions and discussions mentioned in the homework.
- Put your report and your codes within a folder named as "YourNameYourFamily-Practical"+ "Exercise Number".  
(Note that your code is needed to be self-comment)
- Submit all things in a zipped folder.

**Good Luck**