

# EE527: Machine Learning Laboratory

## Assignment 9

Due Date: 03 April 2023

**Classification of Normal and Shouted Speech using MFCC features.** These features are extracted from speech samples of a number of speakers uttering a few sentences normally or by shouting. The features are divided into train-test splits and are made available in two csv files. You are tasked to learn a discriminative model to classify normal and shouted speech. This example uses discriminative functions. The whole process is described as follows.

Consider the .csv file *“Train\_file.csv”* containing 86060 instances of 61-dimensional arrays. The first 60 dimensions of the array contain the feature values for a particular instance and the last dimension contains its label. The label can be either '0' or '1'.

**[Q1]** Evaluate  $\mu_0$ ,  $\mu_1$ ,  $C_0$ ,  $C_1$  from instances in *“Train\_file.csv”*.

$\mu_0$  : Mean of all instances having label '0'

$\mu_1$  : Mean of all instances having label '1'

$C_0$  : Covariance matrix of all instances having label '0'

$C_1$  : Covariance matrix of all instances having label '1'.

Construct parameters of the discriminant function with Gaussian assumption on instance distribution in classes.

$$g(x) = \ln \left\{ \frac{P(X1)}{P(X0)} \right\} - \frac{1}{2} (\mu_1^T C_1^{-1} \mu_1 - \mu_0^T C_0^{-1} \mu_0) - \frac{1}{2} \ln \left\{ \frac{|C_1|}{|C_0|} \right\} \\ + x^T (C_1^{-1} \mu_1 - C_0^{-1} \mu_0) - \frac{1}{2} x^T (C_1^{-1} - C_0^{-1}) x$$

The decision rule for classification of an unseen instance  $x$  is given by its label  $y(x)$  defined as

$$h(x) = \frac{1}{1 + \exp\{-(\omega^T x + \omega_0)\}}$$

$$y(x) = \begin{cases} 1, & h(x) \geq 0.5 \\ 0, & h(x) < 0.5 \end{cases}$$

**[Q2]** Perform Logistic Regression on the training dataset to predict the logit value  $h(x)$  for input data  $x$  as follows.

$$y(x) = \begin{cases} 1, & g(x) \geq 0 \\ 0, & g(x) < 0 \end{cases}$$

Read “*Test\_file.csv*” consisting of 21516 instances of 61 dimensional arrays. For each array, the first 60 dimensions contain the feature values for the test data and the last dimension contains its **actual label**. Predict the label of each data instance from the testing set using the decision rule mentioned above and compare the **predicted** and **actual** labels. Report the class-wise ( $\rho_0$  and  $\rho_1$ ) and overall ( $\rho$ ) accuracy measures.

$$\rho_0 = \frac{\text{No. of Correctly Classified Instances in Class "0"}}{\text{Total Number of Points in Class "0"}}$$

$$\rho_1 = \frac{\text{No. of Correctly Classified Instances in Class "1"}}{\text{Total Number of Points in Class "1"}}$$

$$\rho = \frac{\text{Total No. of Correctly Classified Instances}}{\text{Total Number of Test Instances}}$$