

## Assignment 1 (Due: Oct. 22, 2023)

(Send your solutions to our TA, Tianjun Zhang, 1911036@tongji.edu.cn)

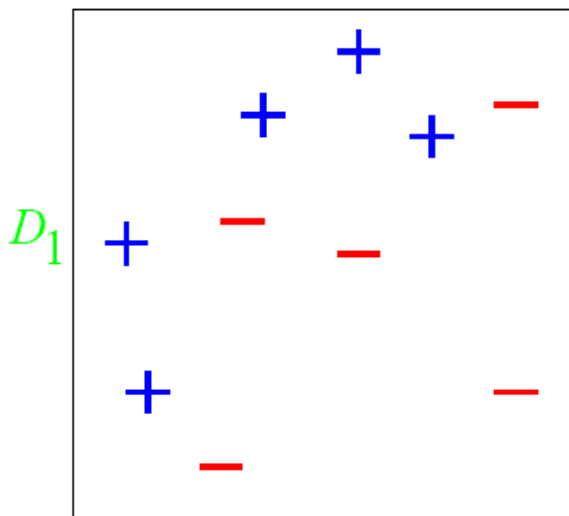
1. **(Programming)** AdaBoost is a powerful classification tool, with which a strong classifier can be learned by composing a set of weak classifiers. In our lecture, we use a vivid example to demonstrate the basic idea of AdaBoost. Now, your task is to implement this demo.

Training:

There are 10 samples on a 2-D plane and information of the  $i$ th sample is given as  $(x_i, y_i, l_i)$ , where  $(x_i, y_i)$  is its coordinate and  $l_i$  is its label. 10 samples are (80, 144, +1), (93, 232, +1), (136, 275, -1), (147, 131, -1), (159, 69, +1), (214, 31, +1), (214, 152, -1), (257, 83, +1), (307, 62, -1), (307, 231, -1). Weak classifiers are vertical or horizontal lines as described in our lecture. The final trained strong classifier actually is a function having the form,

$$\text{label} = \text{strongClassifier}(x, y)$$

Finally, test your strong classifier to verify whether it can correctly classify all the training samples.



2. **(Math)** There are  $n$   $p$ -dimensional data points and we can stack them into a data matrix,  $\mathbf{X} = \{\mathbf{x}_i\}_{i=1}^n, \mathbf{x}_i \in \mathbb{R}^{p \times 1}, \mathbf{X} \in \mathbb{R}^{p \times n}$ . The covariance matrix of  $\mathbf{X}$  is  $\mathbf{C} = \frac{1}{n-1} \sum_{i=1}^n (\mathbf{x}_i - \boldsymbol{\mu})(\mathbf{x}_i - \boldsymbol{\mu})^T$ , where  $\boldsymbol{\mu} = \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i$  (actually, it is the mean of the data points).
  - 1) Please prove that  $\mathbf{C}$  is positive semi-definite.
  - 2) Based on discussions in our lecture, we know that if  $\boldsymbol{\alpha}_1$  is the eigen-vector

associated with the largest eigen-value of  $\mathbf{C}$ , the data projections along  $\boldsymbol{\alpha}_1$  will have the largest variance. Now let's consider such an orientation  $\boldsymbol{\alpha}_2$ . It is orthogonal to  $\boldsymbol{\alpha}_1$ ; and among all the orientations orthogonal to  $\boldsymbol{\alpha}_1$ , the variance of data projections to  $\boldsymbol{\alpha}_2$  is the largest one. Please prove that:  $\boldsymbol{\alpha}_2$  actually is the eigen-vector associated to  $\mathbf{C}$ 's second largest eigen-value. (we can assume that  $\boldsymbol{\alpha}_2$  is a unit-vector)