

## "Machine Learning and Computational Statistics"

### 9<sup>th</sup> Homework (Part B)

#### Exercise 4:

Consider the lines (ε1)  $x_2=0$ , (ε2)  $x_1=0$  and (ε3)  $x_1+x_2=2$  in the two-dimensional space that all leave the point (4,4) on their positive side. Consider a two-class classification problem where class 1 contains all the points that lie on the positive side of all lines, as well as all the points that lie on the negative side of all lines. Class 0 contains all points of the remaining (polygonal) regions

- (i) Design the regions on the plane that correspond to each class.
- (ii) Design a multilayer perceptron that solves the above classification problem, where each node is modeled by the relation  $y = f(\mathbf{w}^T \mathbf{x} + w_0)$ , where  $f(z) = 1$ , for  $z > 0$  and  $f(z) = 0$ , otherwise. Give the full architecture along with the weights and thresholds of each node (describe in some detail the steps you followed for designing the network).

Hint: (i) Use the point (4,4) to identify the positive and the negative sides of each line

(ii) Use the theory given in the lecture.

(iii) The equation of a plane that passes through the points  $(x_{11}, x_{12}, x_{13})$ ,  $(x_{21}, x_{22}, x_{23})$ ,  $(x_{31}, x_{32}, x_{33})$  is

$$\begin{vmatrix} x_1 & x_2 & x_3 & 1 \\ x_{11} & x_{12} & x_{13} & 1 \\ x_{21} & x_{22} & x_{23} & 1 \\ x_{31} & x_{32} & x_{33} & 1 \end{vmatrix} = 0$$

#### Exercise 5 (Python code + text):

Consider a two-class, two-dimensional classification problem for which you can find attached two sets: one for training and one for testing (file [HW9a.mat](#)). Each of these sets consists of pairs of the form  $(y_i, \mathbf{x}_i)$ , where  $y_i$  is the class label for vector  $\mathbf{x}_i$ . Let  $N_{train}$  and  $N_{test}$  denote the number of training and test sets, respectively. The data are given via the following arrays/matrices:

- **train\_x** (a  $N_{train} \times 2$  matrix that contains in its rows the training vectors  $\mathbf{x}_i$ )
- **train\_y** (a  $N_{train}$ -dim. column vector containing the class labels (0 or 1) of the corresponding training vectors  $\mathbf{x}_i$  included in **train\_x**).
- **test\_x** (a  $N_{test} \times 2$  matrix that contains in its rows the test vectors  $\mathbf{x}_i$ )

- $\text{test\_y}$  (a  $N_{\text{test}}$ -dim. column **vector** containing the **class labels** (0 or 1) of the corresponding **test** vectors  $\mathbf{x}_i$  included in  $\text{test\_x}$ ).

**Train** the **SVM classifier** using the training set given above and **measure** its **performance** using the test set, **using**: (a) the **linear kernel**, (b) the **polynomial kernel** and (c) **rbf kernel**. Perform **several runs** using the attached code, for **several choices of the parameters** included in each kernel and for **various values of C**.

### Exercise 6 (Python code + text):

Consider a two-class, two-dimensional classification problem for which you can find attached two **sets**: one for **training** and one for **testing** (file [HW9b.mat](#)). Each of these sets consists of pairs of the form  $(y_i, \mathbf{x}_i)$ , where  $y_i$  is the **class label** for vector  $\mathbf{x}_i$ . Let  $N_{\text{train}}$  and  $N_{\text{test}}$  denote the number of training and test sets, respectively. The data are given via the following arrays/matrices:

- $\text{train\_x}$  (a  $N_{\text{train}} \times 2$  **matrix** that contains in its **rows** the **training** vectors  $\mathbf{x}_i$ )
- $\text{train\_y}$  (a  $N_{\text{train}}$ -dim. column **vector** containing the **class labels** (0 or 1) of the corresponding **training** vectors  $\mathbf{x}_i$  included in  $\text{train\_x}$ ).
- $\text{test\_x}$  (a  $N_{\text{test}} \times 2$  **matrix** that contains in its **rows** the **test** vectors  $\mathbf{x}_i$ )
- $\text{test\_y}$  (a  $N_{\text{test}}$ -dim. column **vector** containing the **class labels** (0 or 1) of the corresponding **test** vectors  $\mathbf{x}_i$  included in  $\text{test\_x}$ ).

**Train** a **neural network classifier** with a **single hidden layer** where the nodes have the **hyperbolic tangent output** function, for (a) 3 nodes, (b) 4 nodes, (c) 10 nodes, (d) 50 nodes (use the **MLPClassifier** Python **function** inserting properly the required parameters, see also the attached code), using the training set given above and **measure** the **performance** using the **test set**. Comment on the results.