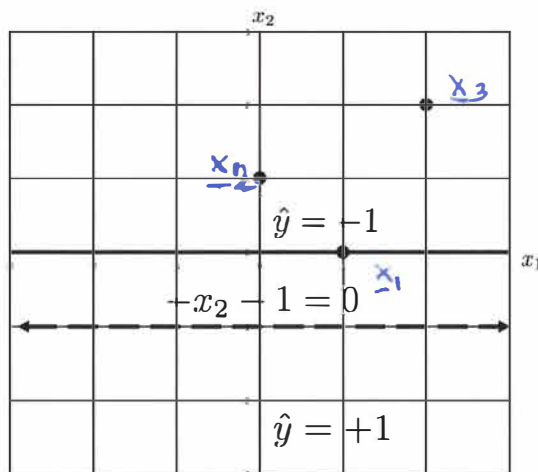


1. (10 MARKS) Consider a binary classification problem where the data points are two dimensional, i.e., $\mathbf{x} = (x_1, x_2) \in \mathbb{R}^2$ and the labels $y \in \{-1, 1\}$. Throughout this problem consider the following three points:

$$\mathbf{x}_1 = (1, 0)^T, \quad \mathbf{x}_2 = (0, 1)^T, \quad \mathbf{x}_3 = (2, 2)^T.$$

4 marks

- (a) Suppose that the hypothesis set \mathcal{H} consists of all linear classifiers whose decision boundary is a horizontal line in the (x_1, x_2) plane. As one example, the classifier $h(\mathbf{x}) = \text{sign}(-x_2 - 1)$ belongs to the set \mathcal{H} . The decision boundary of $h(\mathbf{x})$ is the dashed horizontal line shown in the figure below. Note that in this classifier all points below this line are classified as $\hat{y} = +1$ while all points above the horizontal line are classified as $\hat{y} = -1$ by the hypothesis $h(\mathbf{x})$



List all dichotomies in $\mathcal{H}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$ that can be achieved. (Recall that a dichotomy in this problem will be a vector of length three whose elements are either +1 or -1, and is achieved by applying some hypothesis in \mathcal{H} to the points \mathbf{x}_1 , \mathbf{x}_2 and \mathbf{x}_3 in that order.)

What is the VC dimension of \mathcal{H} ? (No justification is needed for this)

Achievable Dichotomies:

\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3
+1	+1	+1
-1	+1	+1
-1	-1	+1
-1	-1	-1
+1	-1	-1
+1	+1	-1

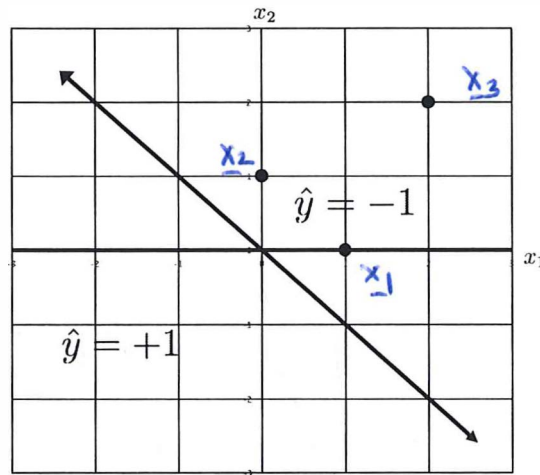
$$d_{VC}(\mathcal{H}) = 1$$

1 pt

0.5 pt for each correct dichotomy
-0.5 for each incorrect

4 marks

- (b) Suppose that the hypothesis set \mathcal{G} consists of all linear classifiers **passing through the origin**. As one example the classifier $g(\mathbf{x}) = \text{sign}(-x_1 - x_2)$ belongs to the set \mathcal{G} . Its decision boundary is shown by the solid line passing through the origin in the figure below. Note that all points below the decision boundary are classified as $\hat{y} = +1$ and all points above this line are classified as $\hat{y} = -1$.



List all dichotomies in $\mathcal{G}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$ that can be achieved.

State without justification the VC dimension of \mathcal{G} .

0.5 for each correct
-0.5 for an incorrect

\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3
+	+	+
-	+	+
+	-	+
-	-	-
+	-	-
-	+	-

$$\text{dvc}(\mathcal{H}) = 2$$

1 pt

2 marks

- (c) Suppose $\mathcal{M} = \mathcal{H} \cup \mathcal{G}$ is the union of the hypothesis classes in parts (a) and (b). What is the number of dichotomies in $\mathcal{M}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$? Provide a **brief** justification for your answer.

$|\mathcal{M}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)| = 8$ — ①
the two dichotomy vector missing
in \mathcal{G} , are found in \mathcal{H} . } ①

10 marks

3. Suppose we are given a sequence of real numbers: $x_1, x_2, x_3, x_4, \dots$ where $x_i \in \mathbb{R}$. We observe the following values: $x_1 = -1, x_2 = 0, x_3 = +1, x_4 = +1$.

We wish to select a prediction function of the form $\hat{x}_i = w_0 + w_1 \cdot x_{i-1}$, for $i \geq 2$ that makes a prediction of x_i from the value of x_{i-1} . Our task is to minimize the following in sample training error:

$$E_{\text{in}}(\mathbf{w}) = \frac{1}{3} \sum_{i=2}^4 (\hat{x}_i - x_i)^2.$$

3 marks

- (a) Rewrite the above problem specifications to get the problem into the standard form of a least squares problem: $\mathbf{w}^* = \arg \min_{\mathbf{w} \in \mathbb{R}^2} \frac{1}{3} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|^2$. Specifically specify the data matrix \mathbf{X} and the target vector \mathbf{y} , where $\mathbf{w} = (w_0, w_1)^T$.

$$\underline{\mathbf{y}} = \begin{bmatrix} x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \rightarrow 1 \text{ pt}$$

$$\bar{\mathbf{X}} = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix} \rightarrow 2 \text{ pt}$$

3 marks

- (b) Find the least squares solution $\mathbf{w}^* = (w_0^*, w_1^*)$ in part (a).

$$\underline{\mathbf{w}}^* = (\bar{\mathbf{X}}^T \bar{\mathbf{X}})^{-1} \bar{\mathbf{X}}^T \underline{\mathbf{y}}$$

$$(\bar{\mathbf{X}}^T \bar{\mathbf{X}})^{-1} = \begin{bmatrix} 1/3 & 0 \\ 0 & 1/2 \end{bmatrix} \rightarrow 1 \text{ pt}$$

$$\bar{\mathbf{X}}^T \underline{\mathbf{y}} = \begin{bmatrix} 1 & 1 & 1 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \rightarrow 1 \text{ pt}$$

$$\underline{\mathbf{w}}^* = \begin{bmatrix} 2/3 \\ 1/2 \end{bmatrix} \rightarrow 1 \text{ pt}$$

(alternative solution accepted)

4 marks

(c) Redo parts (a) and (b) if the prediction function is of the form $\hat{x}_i = w_1 \cdot x_{i-1}$ i.e., we set $w_0 = 0$.

$$\underline{y} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

1 pt

$$\underline{X} = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

1 pt

$$(\underline{X}^T \underline{X})^{-1} = \frac{1}{2}$$

1 pt

$$\underline{X}^T \underline{Y} = \begin{bmatrix} -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = 1$$

1 pt

$$\underline{w}_1^* = \frac{1}{2}$$

1 pt

(alternative soluⁿ to \underline{w}_1^* is accepted)