

Solutions on attached pages.

In-class Quiz 1 (September 10, 2024)

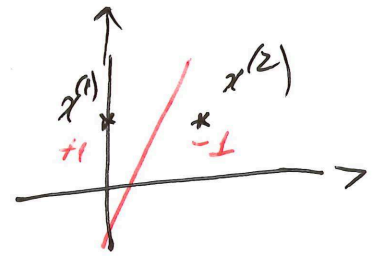
YOUR NAME HERE: _____

1 Linear classifiers (25 + 5 bonus points)

Consider a 2×1 feature vector $\mathbf{x} = [0.3, 0.4]$ that has been labeled with $y = +1$. Answer each of the questions below and **show your work**.

1. Calculate the L2 norm of \mathbf{x} (its length).
2. Calculate the dot product $\mathbf{w}^T \mathbf{x}$, where $\mathbf{w} = [-1, 1]$ is a 2×1 weight vector.
3. Find a weight vector $\mathbf{w} = [w_1, w_2]$ such that $\mathbf{w}^T \mathbf{x} - 2 \geq 0$.
4. Consider a linear model for binary classification with feature-wise parameters $\mathbf{w} = [w_1, w_2] = [-1, 1]$ and bias parameter $w_0 = -0.5$.
 - (a) What is the label \hat{y} predicted by this model for the example \mathbf{x} ?
 - (b) What is the equation for the decision boundary of this linear classifier?
 - (c) [Bonus] What is the distance between example \mathbf{x} and the decision boundary?

2 The Perceptron algorithm (25 points)



Consider a dataset with two training examples (bias feature not shown):

$$\mathbf{x}^{(1)} = [0, 1] \text{ with label } y_1 = +1.$$

$$\mathbf{x}^{(2)} = [1, 1] \text{ with label } y_1 = -1.$$

1. Will the Perceptron algorithm converge on this dataset? Explain your answer.

Yes, bc. they are lin. sep.

Consider the Perceptron algorithm discussed in class, where the parameters are initialized as $\mathbf{w} = [w_0, w_1, w_2] = [-0.5, 0, 0]$. Run one epoch of the algorithm on the two training examples $\mathbf{x}^{(1)}$ and $\mathbf{x}^{(2)}$ above, in this order, and compute the following:

2. The weight vector after processing example $\mathbf{x}^{(1)}$. Show your work.

$$\mathbf{w} = [-0.5, 0, 0] \quad \mathbf{w}^T \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = -0.5 + 0 \times 0 + 0 \times 1 = -0.5 \Rightarrow \hat{y}_1 = -1$$

$$\mathbf{w} \leftarrow \mathbf{w} + [1, 0, 1] \Rightarrow \mathbf{w} = [0.5, 0, 1]$$

3. The weight vector after processing example $\mathbf{x}^{(2)}$. Show your work.

$$\mathbf{w}^T \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \cancel{0.5 \cdot 1 + 0 \cdot 1 + 1 \cdot 1} = 0.5 \times 1 + 0 \times 1 + 1 \times 1 = 1.5 \Rightarrow \hat{y}_2 = +1$$

$$\mathbf{w} \leftarrow \mathbf{w} + [1, 1, 1] \times -1 = [-0.5, -2, 0]$$

Assume examples $\mathbf{x}^{(1)}$ and $\mathbf{x}^{(2)}$ are stored in the rows of a 2D NumPy array named X , and the model parameters are stored in a 1D array named \mathbf{w} . Write Numpy code that:

4. Adds a bias feature to each example in X and stores the results in an array X_{new} .

$$X = \begin{bmatrix} \mathbf{x}^{(1)} \\ \mathbf{x}^{(2)} \end{bmatrix} \quad X_{\text{new}} = \begin{bmatrix} 1 & \vdots & \mathbf{x}^{(1)} \\ 1 & \vdots & \mathbf{x}^{(2)} \end{bmatrix} \quad \text{np.hstack}((\text{np.ones}(2), X))$$

\downarrow column-stack.

5. Computes in `pred` the label predicted by \mathbf{w} on the second example from X_{new} .

$$\text{pred} = \text{np.sign}(\mathbf{w} \cdot \text{dot}(X_{\text{new}}[1])) > 0$$

$$= +1 \text{ if } \downarrow \text{ else } -1.$$

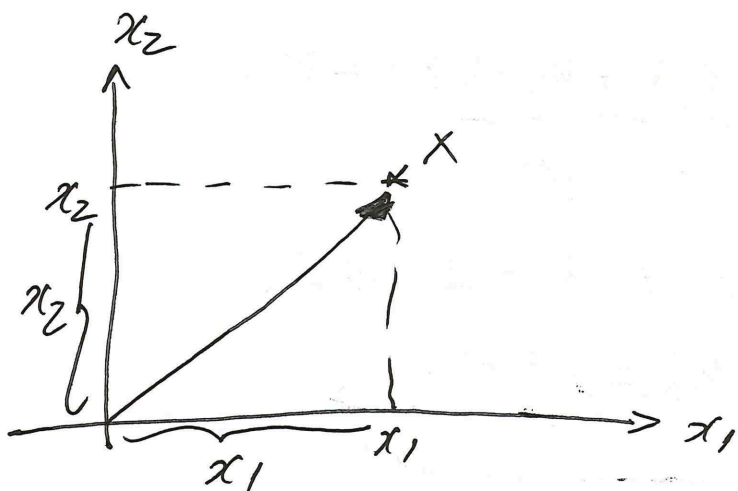
$$\textcircled{1} \quad x = [x_1, x_2, \dots, x_N]$$

L_2 norm \equiv Euclidean norm \equiv length

$$\sqrt{x_1^2 + x_2^2 + \dots + x_N^2}$$

$$= \sqrt{\sum_{n=1}^N x_n^2} = \left(\sum_{n=1}^N x_n^2 \right)^{1/2}$$

$$L_k \text{ norm} = \left(\sum_{n=1}^N |x_n|^k \right)^{1/k}, \quad L_0, L_\infty, \dots$$



$$x = [0.3, 0.4]$$

$$\begin{aligned} \|x\| &= \|x\|_2 = \sqrt{0.3^2 + 0.4^2} = \sqrt{0.09 + 0.16} \\ &= \sqrt{0.25} = 0.5 \end{aligned}$$

② Dot-product of $x = \begin{bmatrix} 0.3 \\ 0.4 \end{bmatrix}$ and $w = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ is $w^T x = x^T w =$

$$= -0.3 + 0.4 = 0.1.$$

$\uparrow \quad \uparrow$
 $-1 \cdot 0.3 \quad 1 \cdot 0.4$

③ $x = \begin{bmatrix} 0.3 \\ 0.4 \end{bmatrix}$. Find $w = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$ s.t. $w^T x - 2 \geq 0$

$$\underbrace{w_1 \cdot 0.3 + w_2 \cdot 0.4 - 2}_{10 \quad 10} \geq 0$$

$$3 + 4 - 2 \geq 0.$$

④ $x = \begin{bmatrix} 0.3 \\ 0.4 \end{bmatrix}$ $w = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} -1 \\ +1 \end{bmatrix}$, bias $w_0 = -0.5$.

4.a $\underbrace{w^T x + w_0}_{\downarrow} \geq 0 ?$

\nearrow Yes $\Rightarrow \hat{y} = +1$
 \searrow No $\Rightarrow \hat{y} = -1$

$0.2 - 0.5 = -0.3 < 0 \Rightarrow \hat{y} = -1$

4.b The decision boundary is $\{ \underline{x} \in \mathbb{R}^2 \mid \underline{w}^T \underline{x} + w_0 = 0 \}$

(9.c) Distance between x & l.d.b is $\frac{|w^T x + w_0|}{\|w\|} = \frac{|-0.4|}{\sqrt{2}} = 0.2\sqrt{2}$

$$w = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \rightarrow \|w\| = \sqrt{2}$$