Short Assignment 3

This is an individual assignment.

Due: Friday, November 4 @ 11:59PM

Problem 1

Consider the two classes represented by Gaussians distributions P1 and P2 in Figures 1 and 2. Calculate Fisher's univariate separation indices to answer the following questions.

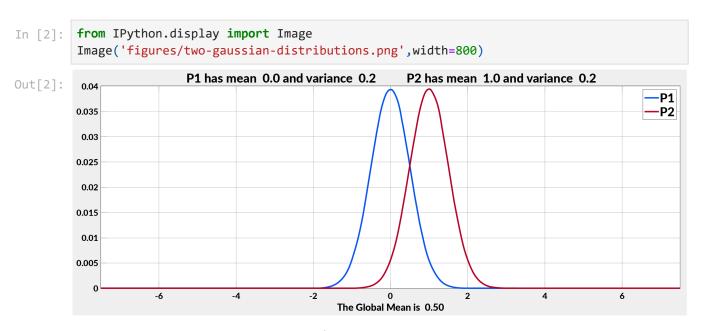


Figure 1: Two Gaussian Classes

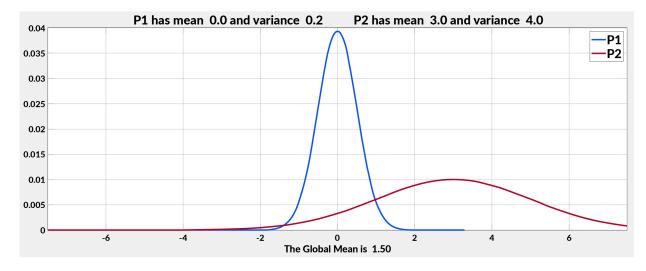


Figure 2: Two Gaussian Classes

1. What is the Within-Class Separation in Figure 1?

- 2. What is the Within-Class Separation in Figure 2?
- 3. What is the Between-Class Separation in Figure 1?
- 4. What is the Between-Class Separation in Figure 2?
- 5. Which Distributions are more separated: (A) P1,P2 in Figure 1 or (B) P1,P2 in Figure 2? Justify your answer.

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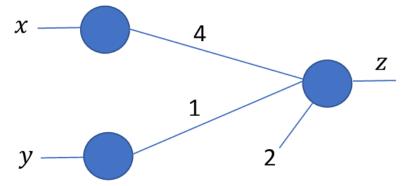
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Problem 2

Consider the following perceptron:

In [3]: Image('figures/Perceptron.png', width=400)

Out[3]:



Recall that the perceptron uses the activation function:

$$\phi(x) = \begin{cases} -1 & x \le 0 \\ 1 & x > 0 \end{cases}$$

And the cost function is:

$$E_p(\mathbf{w},b) = -\sum_{m \in \mathcal{M}} (\mathbf{w}^T \mathbf{x}_n + b)^T t_n$$

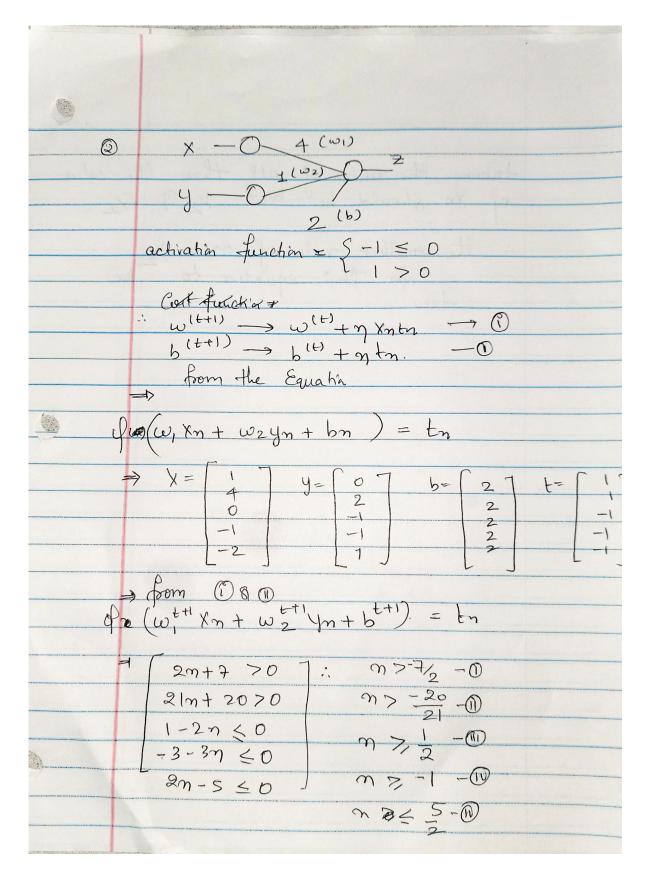
where is the set of all misclassified points. The update equations for the weights and bias term are:

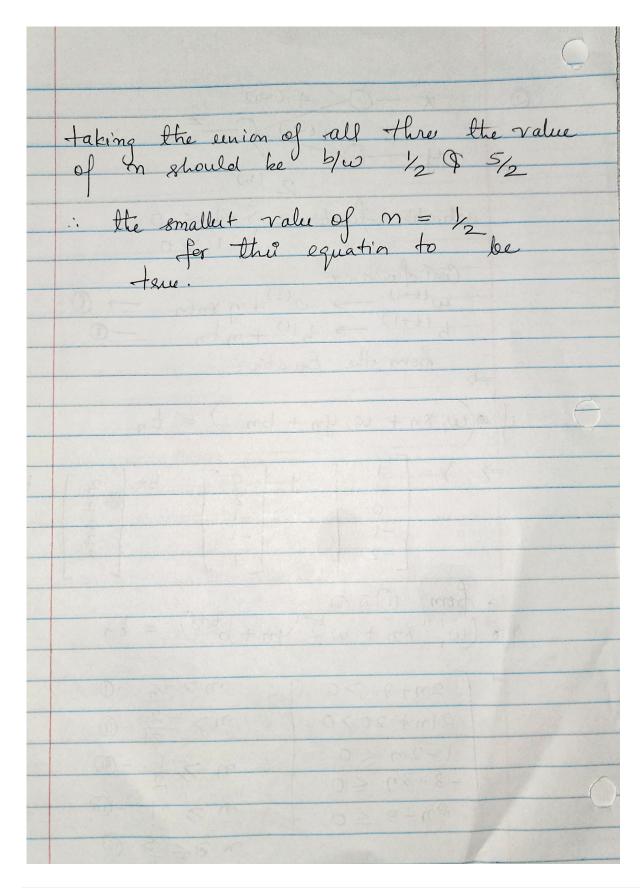
$$egin{aligned} \mathbf{w}^{(t+1)} \leftarrow & \mathbf{w}^{(t)} - \eta rac{\partial E_p(\mathbf{w}, b)}{\partial \mathbf{w}} = \mathbf{w}^{(t)} + \eta \mathbf{x}_n t_n \ b^{(t+1)} \leftarrow & b^{(t)} - \eta rac{\partial E_p(\mathbf{w}, b)}{\partial b} = b^{(t)} + \eta t_n \end{aligned}$$

Suppose you have the following 5 data samples (x, y) and their corresponding labels t:

$$(x_1,y_1)=(1,0) ext{ with } t_1=1 \ (x_2,y_2)=(4,2) ext{ with } t_2=1 \ (x_3,y_3)=(0,-1) ext{ with } t_3=-1 \ (x_4,y_4)=(-1,-1) ext{ with } t_4=-1 \ (x_5,y_5)=(-2,1) ext{ with } t_5=-1$$

What is the smallest value for the learning rate η such that the updated network will result in zero misclassified points using only one iteration?





In []:

Confirm that you've successfully completed the assignment.

Along with the Notebook, include a PDF of the notebook with your solutions.

add and commit the final version of your work, and push your code to your GitHub repository.

Submit the URL of your GitHub Repository as your assignment submission on Canvas.