

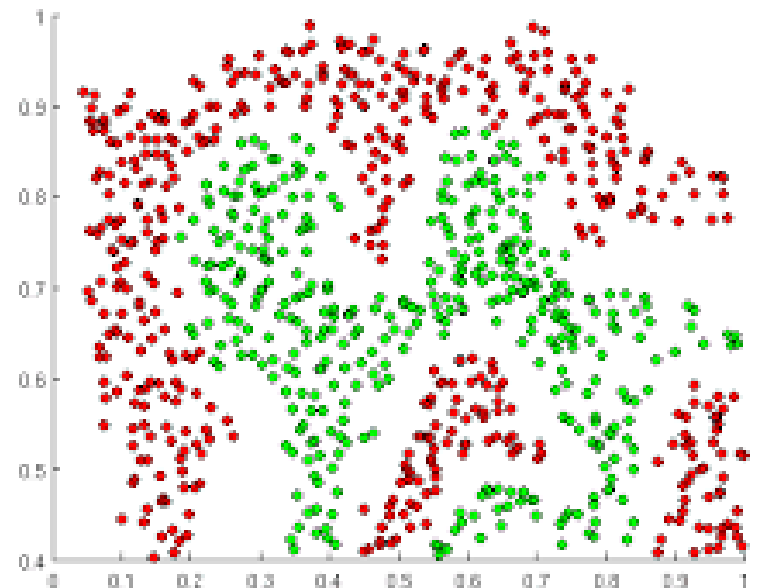
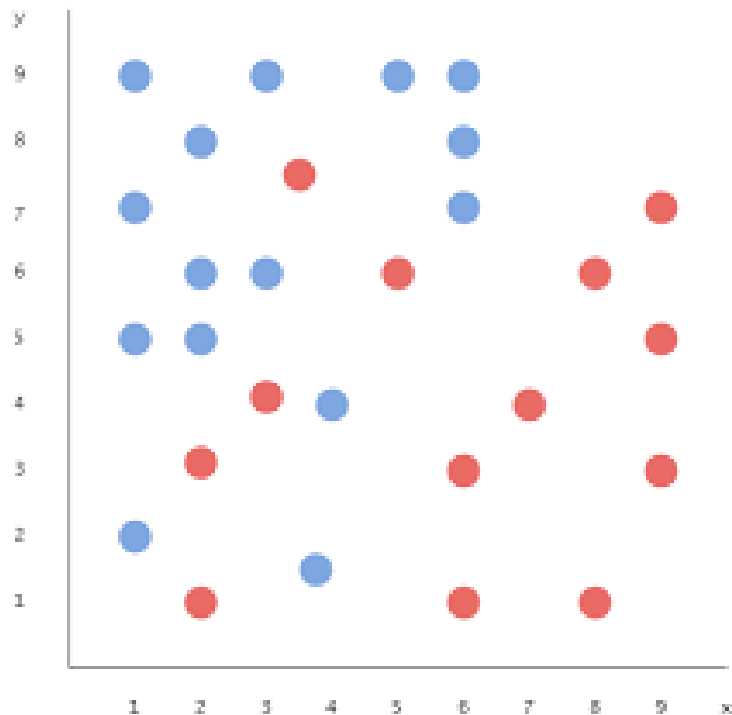
# PERCEPTRON: PROOF OF CONVERGENCE

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Assistant Professor

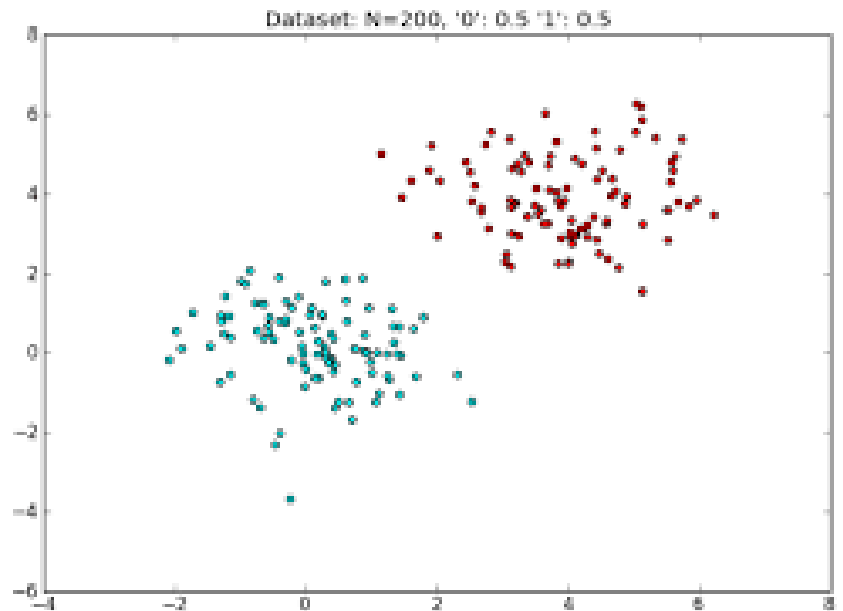
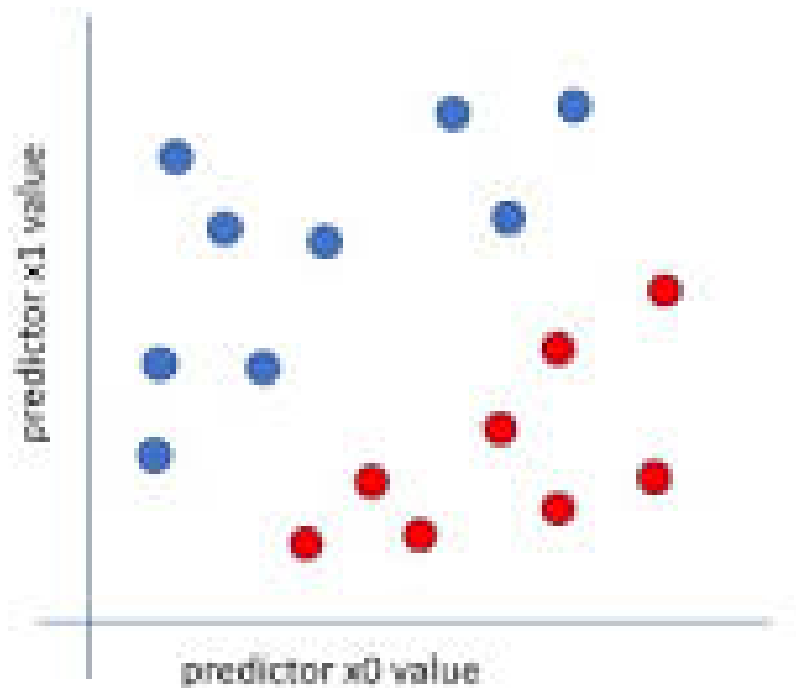


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# What classifier do you use?



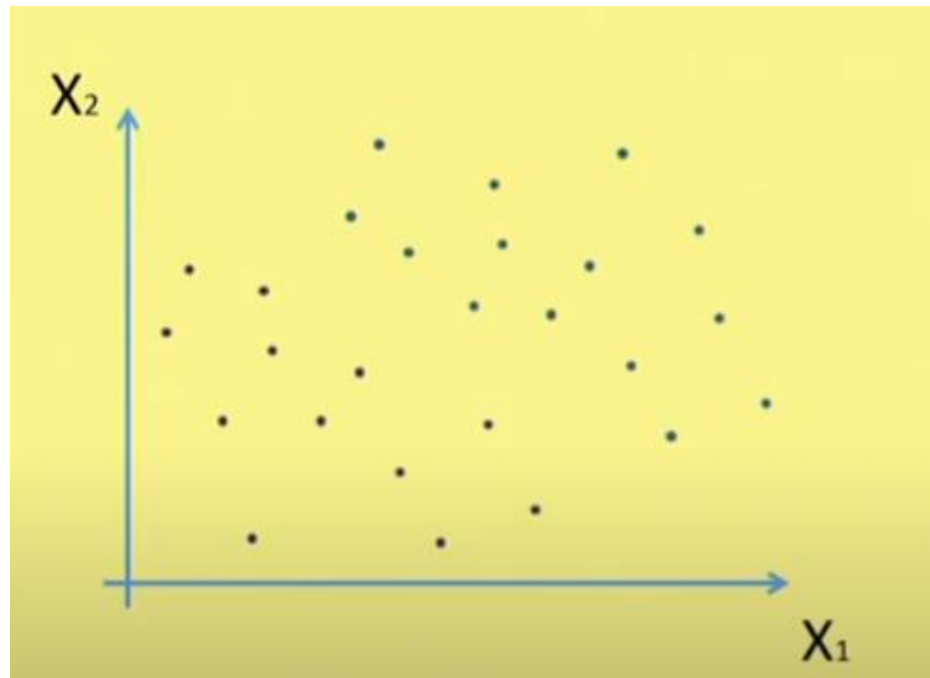
# What classifier do you use?



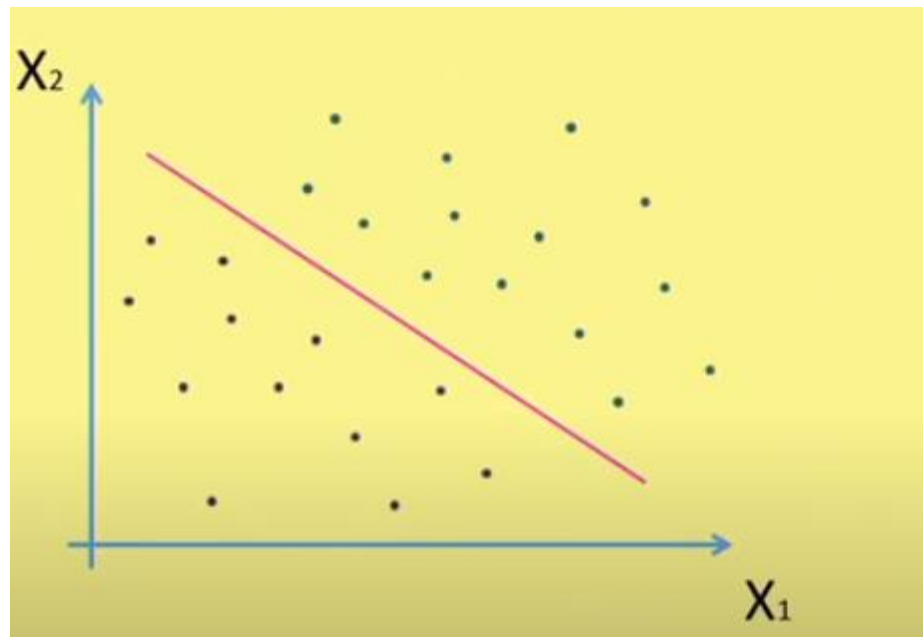
# Possible solution for previous problems

- K-NN search
  - ▣ **Testing:** It is time consuming as it should compare the distance with all the sample
  - ▣ This methods works well if the number of samples are less
- Linear discriminant function (LDF)
  - ▣ It uses sample to estimate the parameter such as  $\mathbf{w}$  and  $\mathbf{w}_0$
  - ▣ **Testing:** It just substitute the value of  $x$  in  $g(x)$  and take the decision
  - ▣ If  $g(x) > 0$  then class 1, otherwise class 2

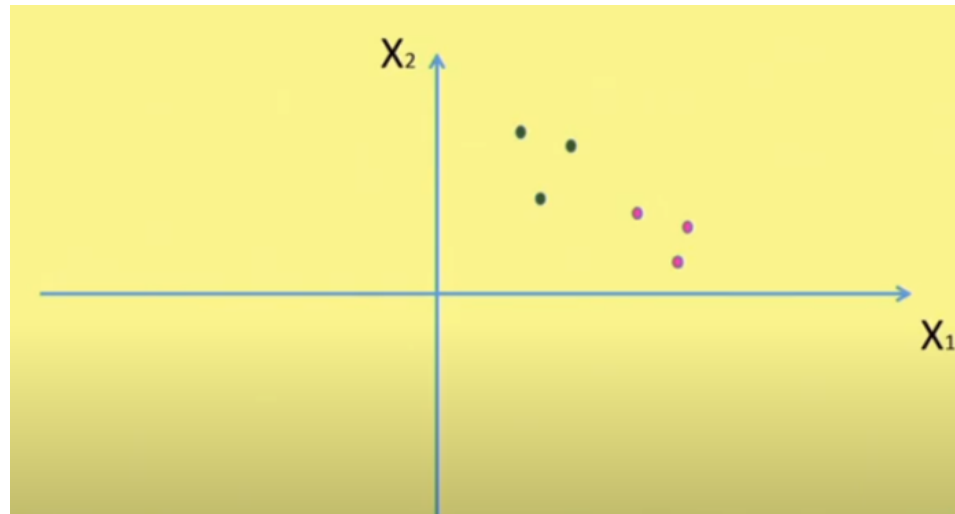
# Perceptron: two class problem



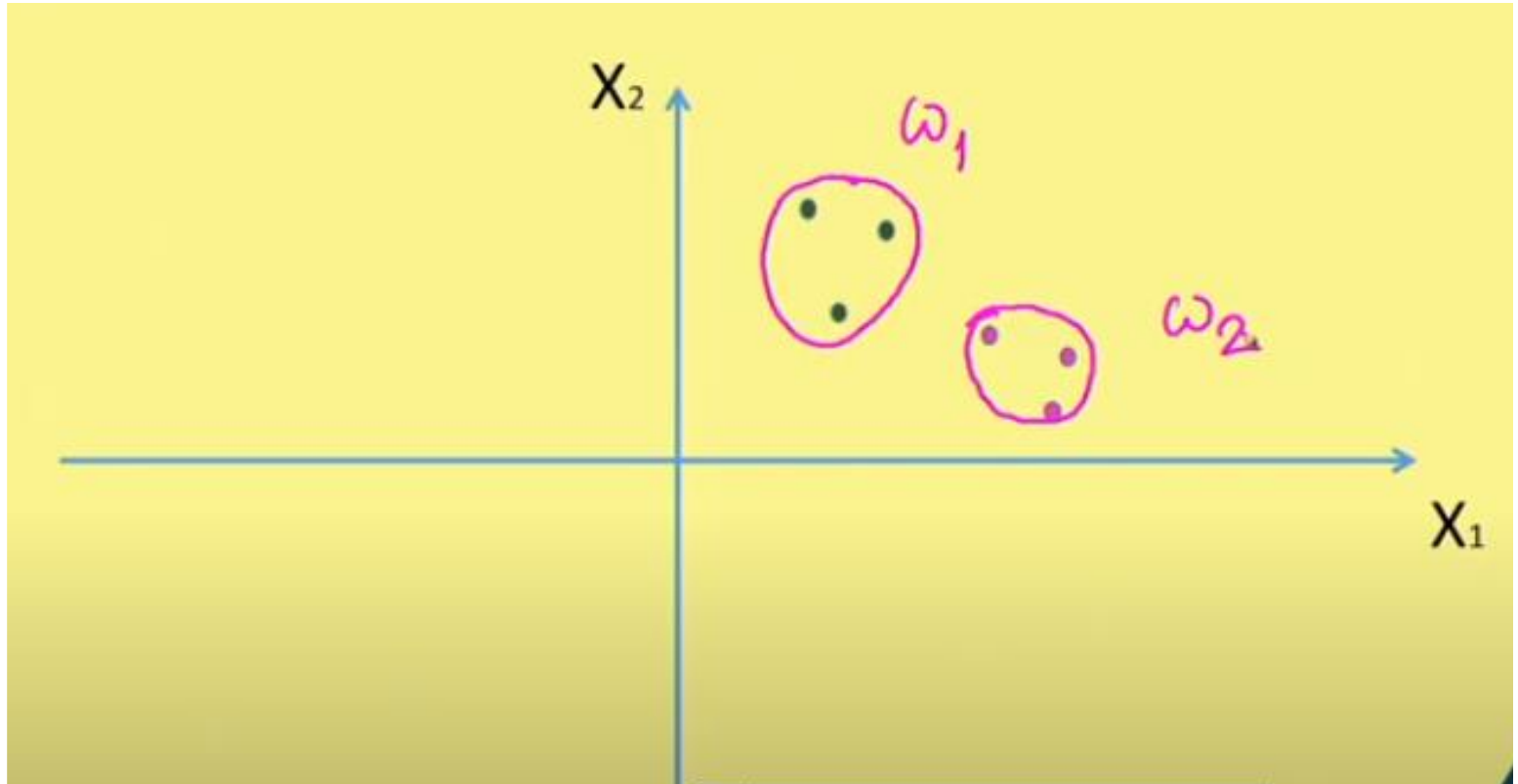
# Perceptron - two class problem



# Perceptron -two class problem



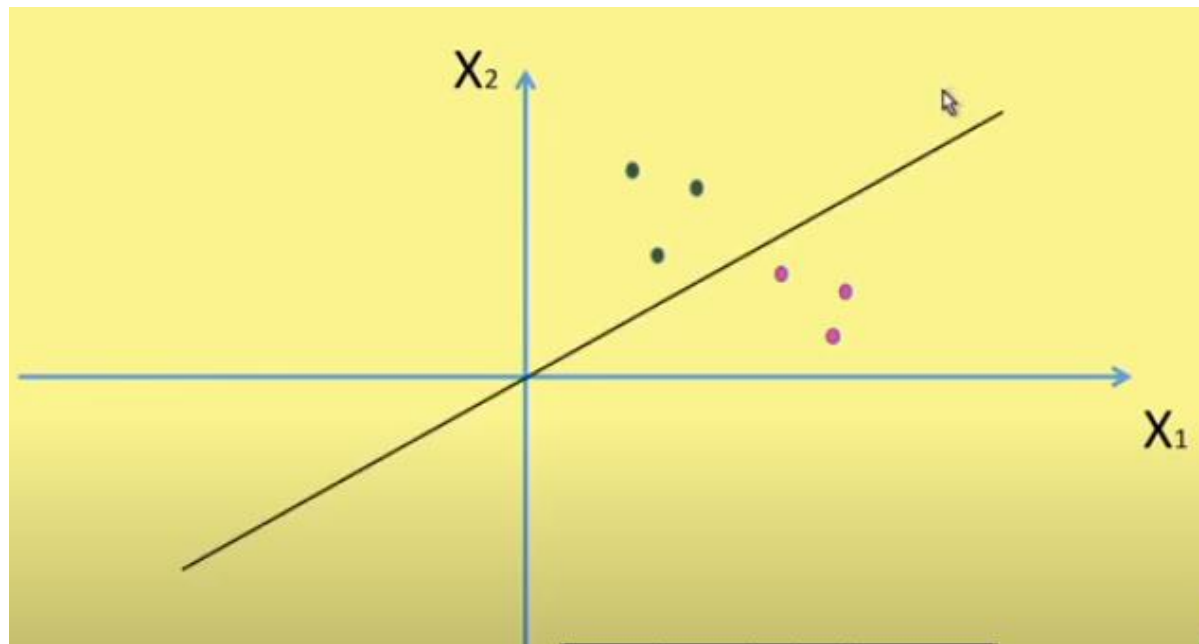
# Linear Classifier: Perceptron





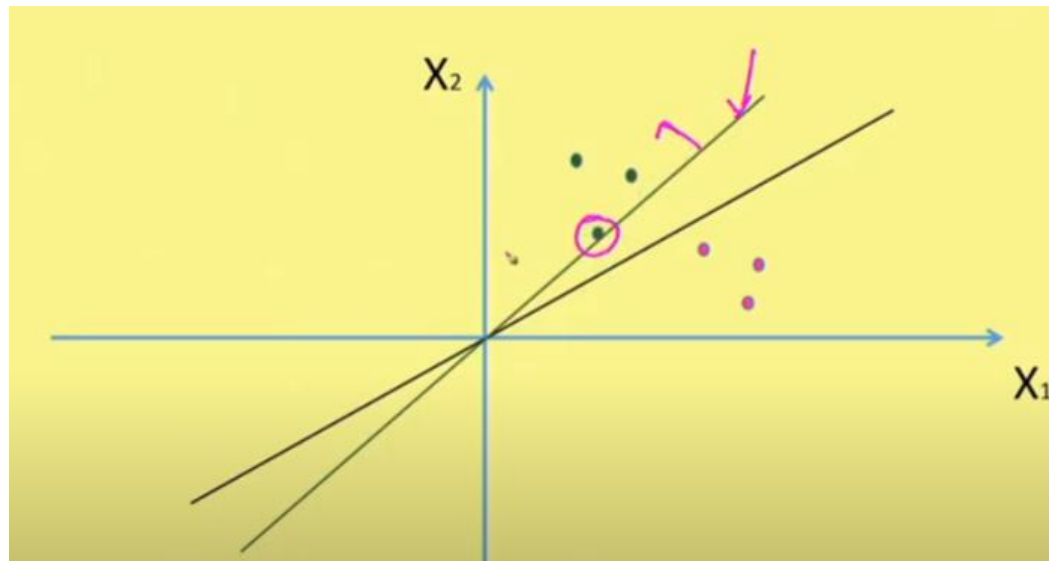
# Linear Classifier: Perceptron

- This linear line clearly separates the two classes
- Also, we can see there is a limit of orientation of this particular line (decision boundary)



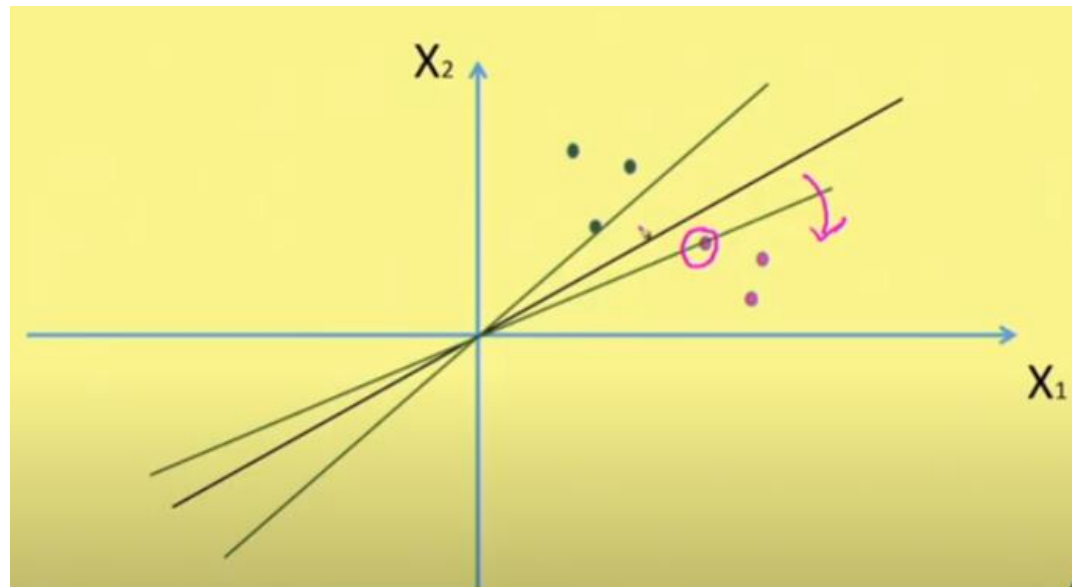
# Linear Classifier: Perceptron

- If we rotate further in anti-clock wise direction, the misclassified sample is the one marked as circle.



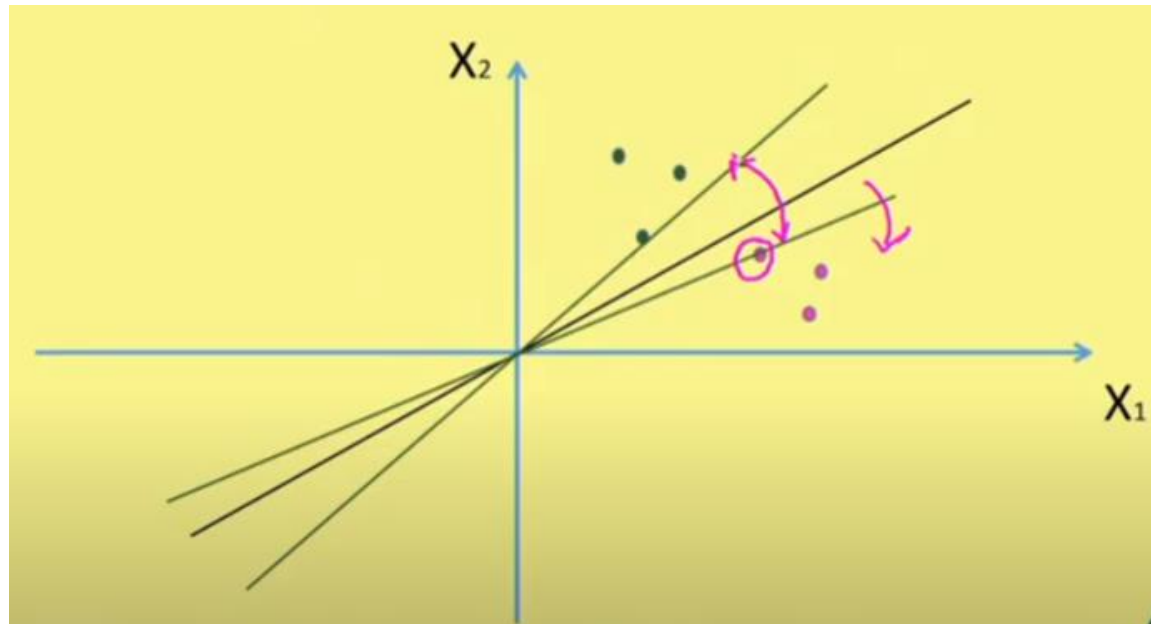
# Linear Classifier: Perceptron

- Similarly, if we come to the other side, this is the limit
- If we rotate it further in clock wise direction; the misclassified sample is shown in circle



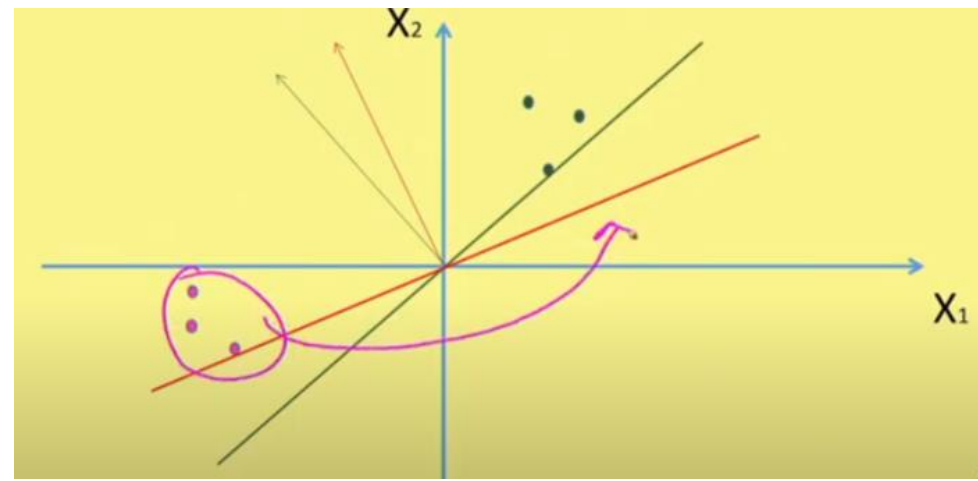
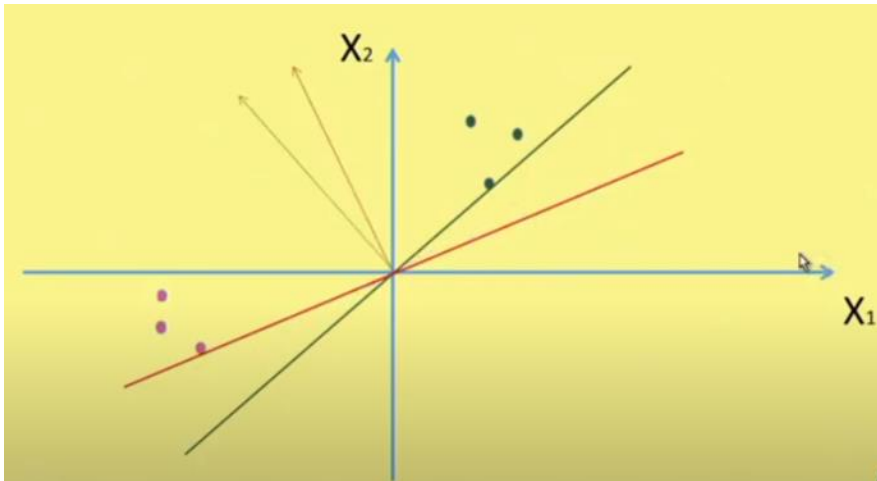
# Linear Classifier: Perceptron

- So the range, that we could have is shown here.

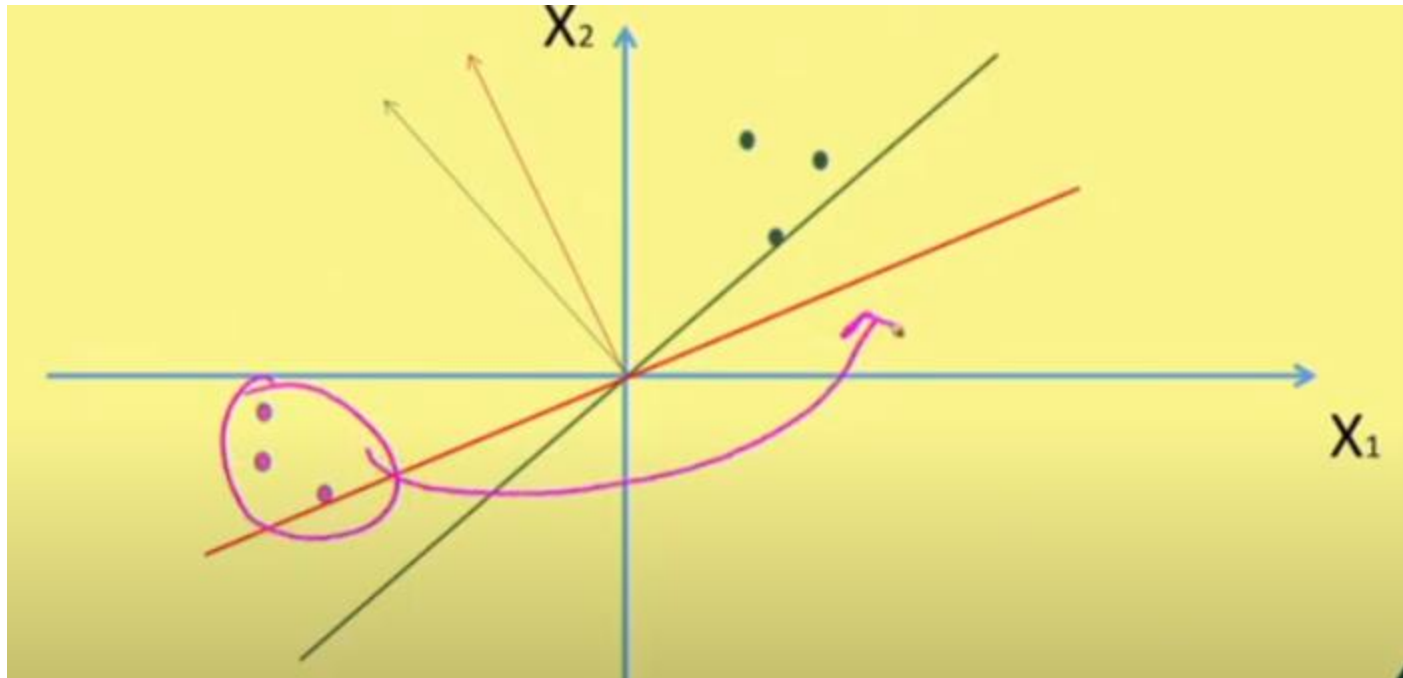


# Linear Classifier: Perceptron

- After taking negation of all sample for class 2

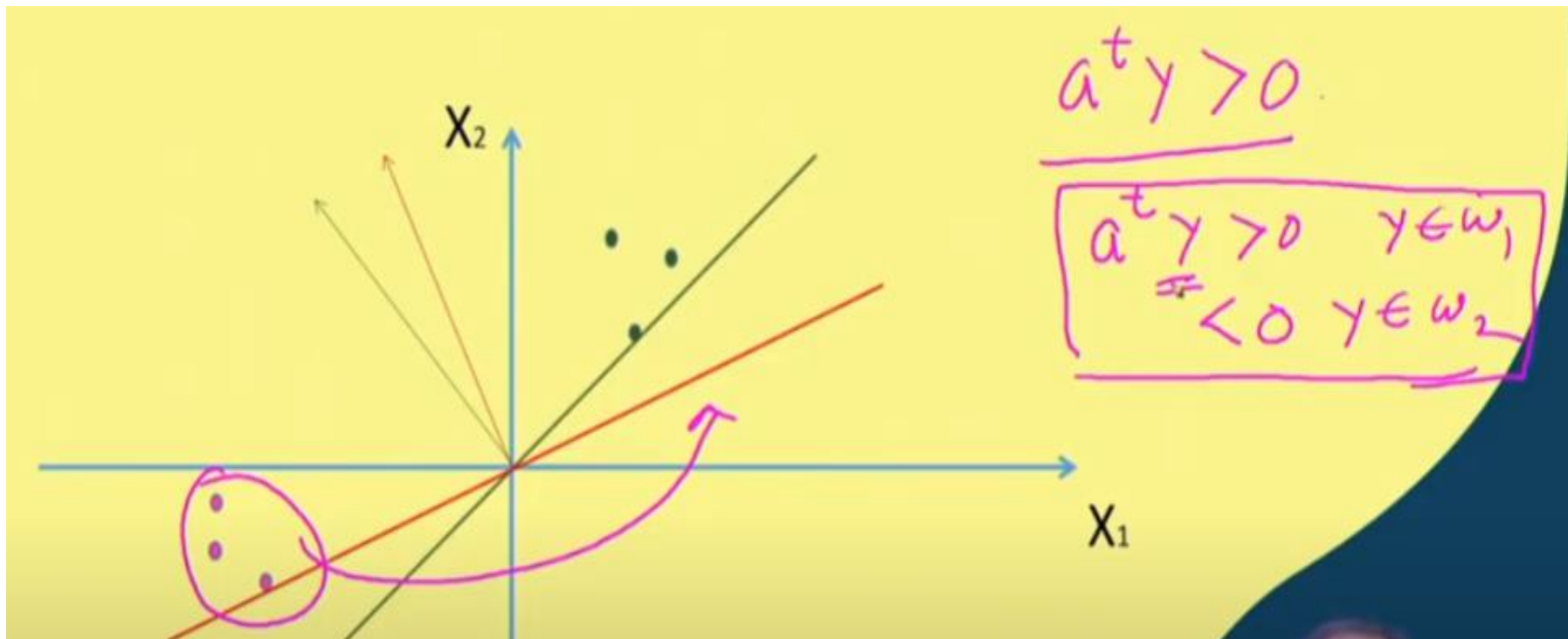


# Linear Classifier: Perceptron

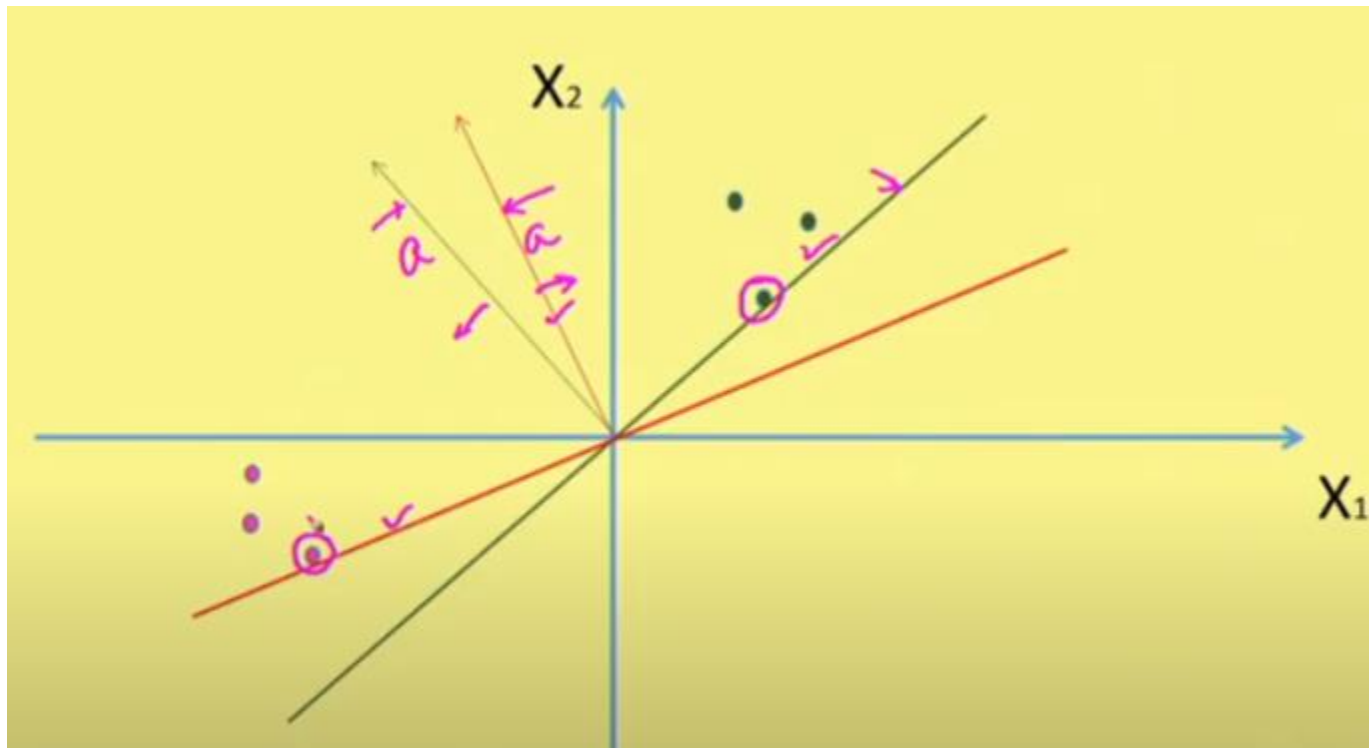


# Linear Classifier: Perceptron

- For design purpose alone we use  $a^t y > 0$
- For testing we use if  $a^t y > 0$ , then  $y$  is class 1
- If  $a^t y < 0$ , then  $y$  is class 2



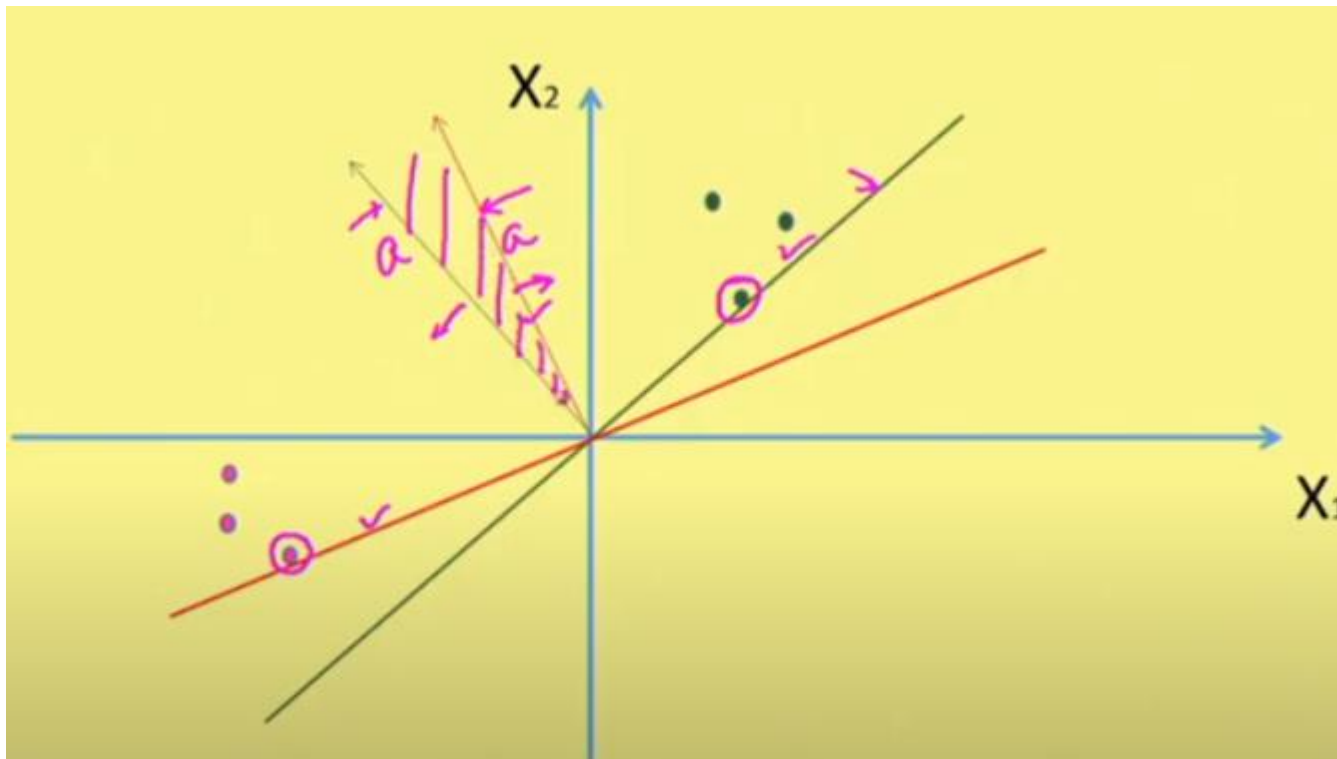
# Linear Classifier: Perceptron





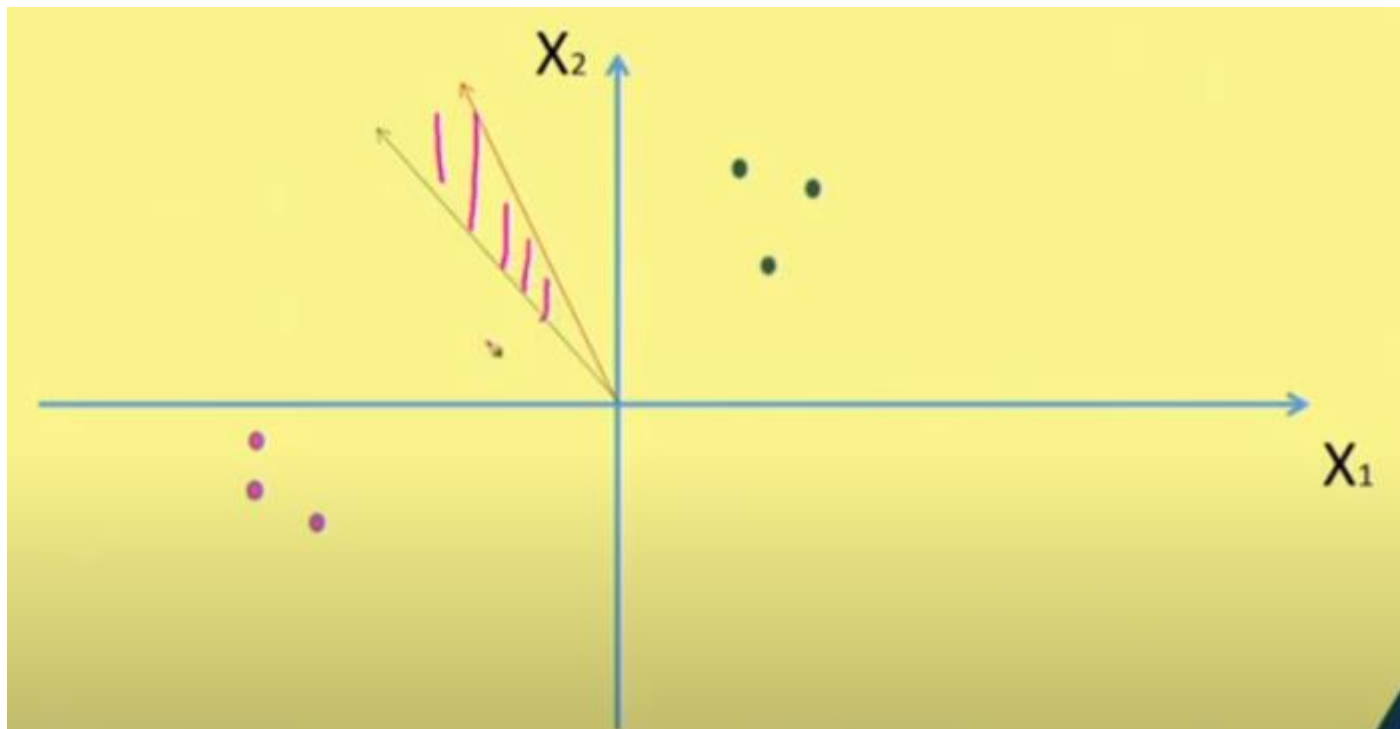
# Linear Classifier: Perceptron

- So the solution vector ' $a$ ' must lie on this solution region.



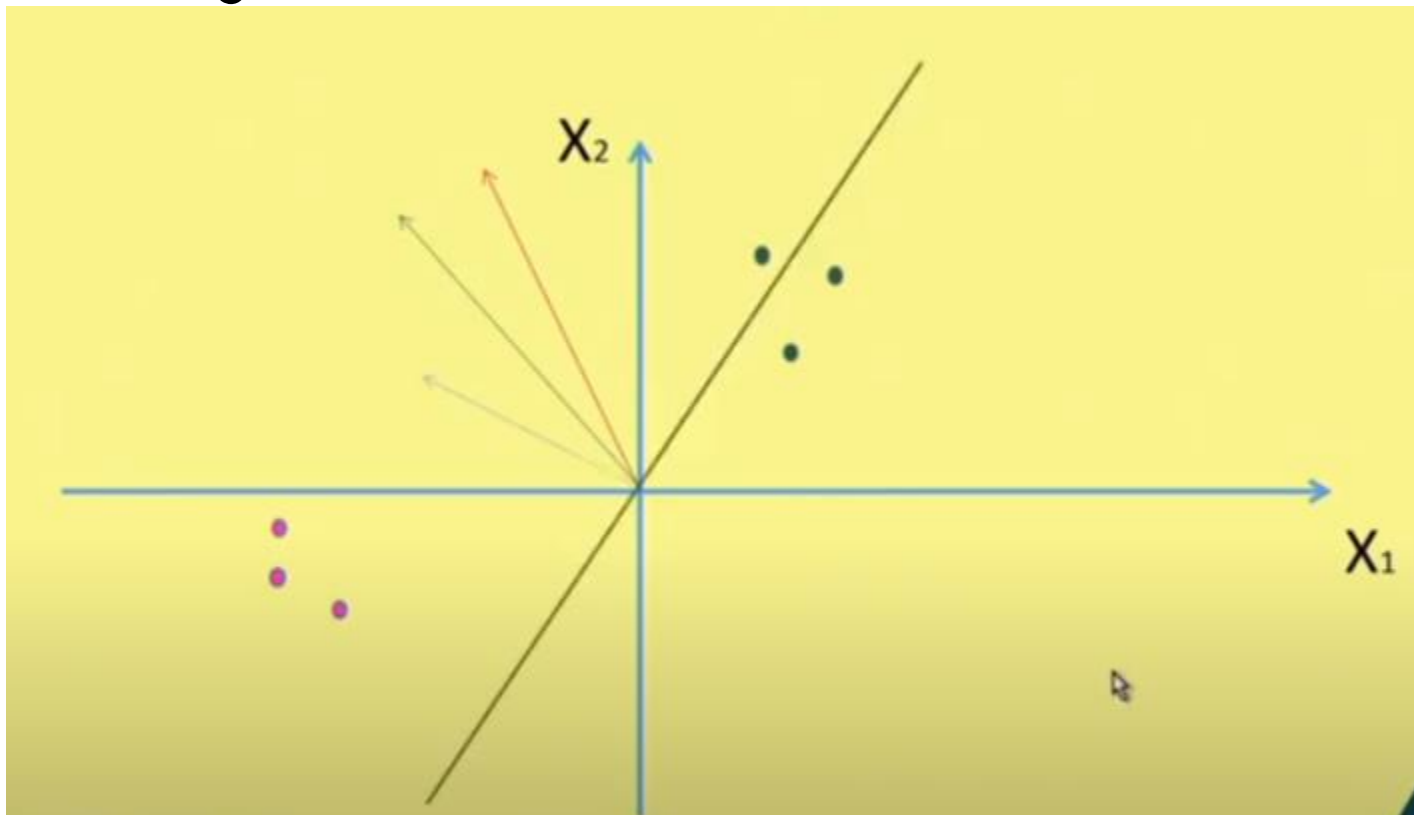
# Linear Classifier: Perceptron

- How do we make the solution vector ' $a$ ' should land in solution region?

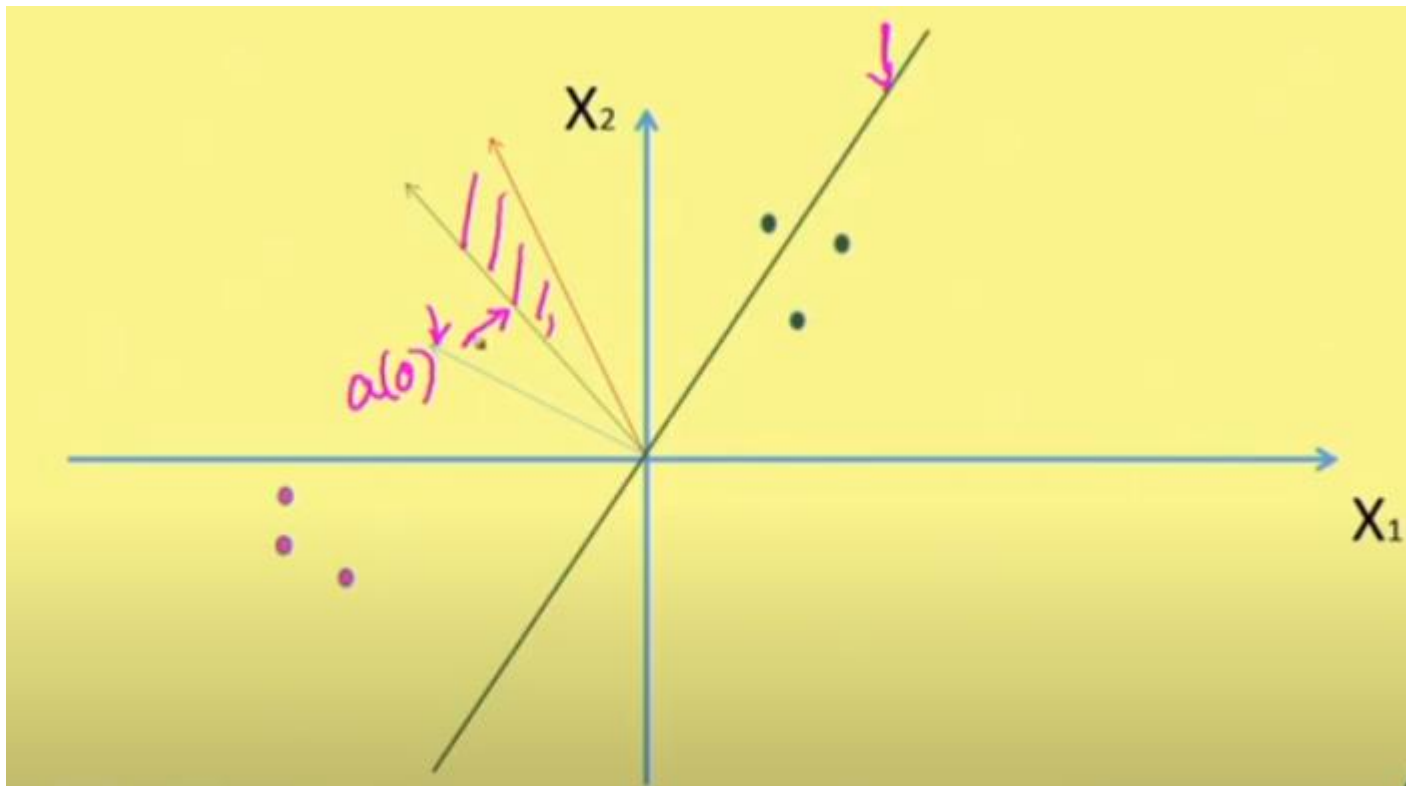


# Linear Classifier: Perceptron

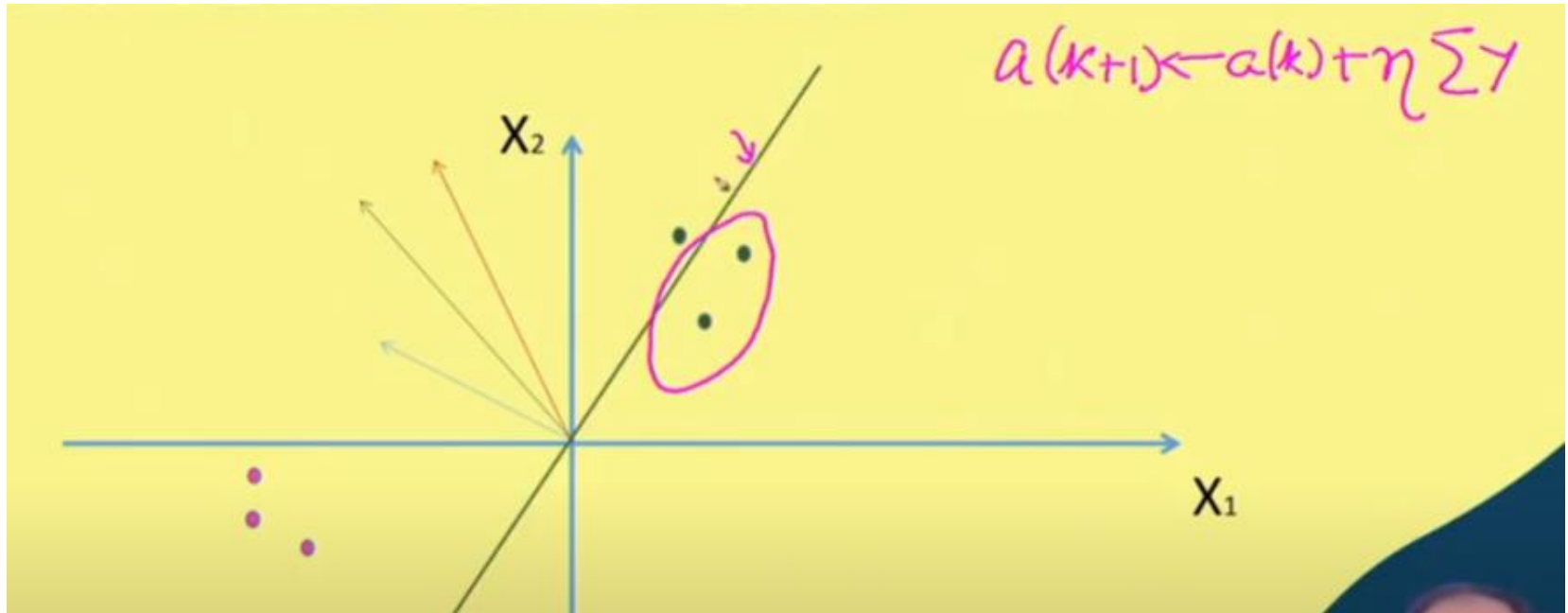
- Suppose, initial weight vector  $a(0)$  is chosen something like this



# Linear Classifier: Perceptron

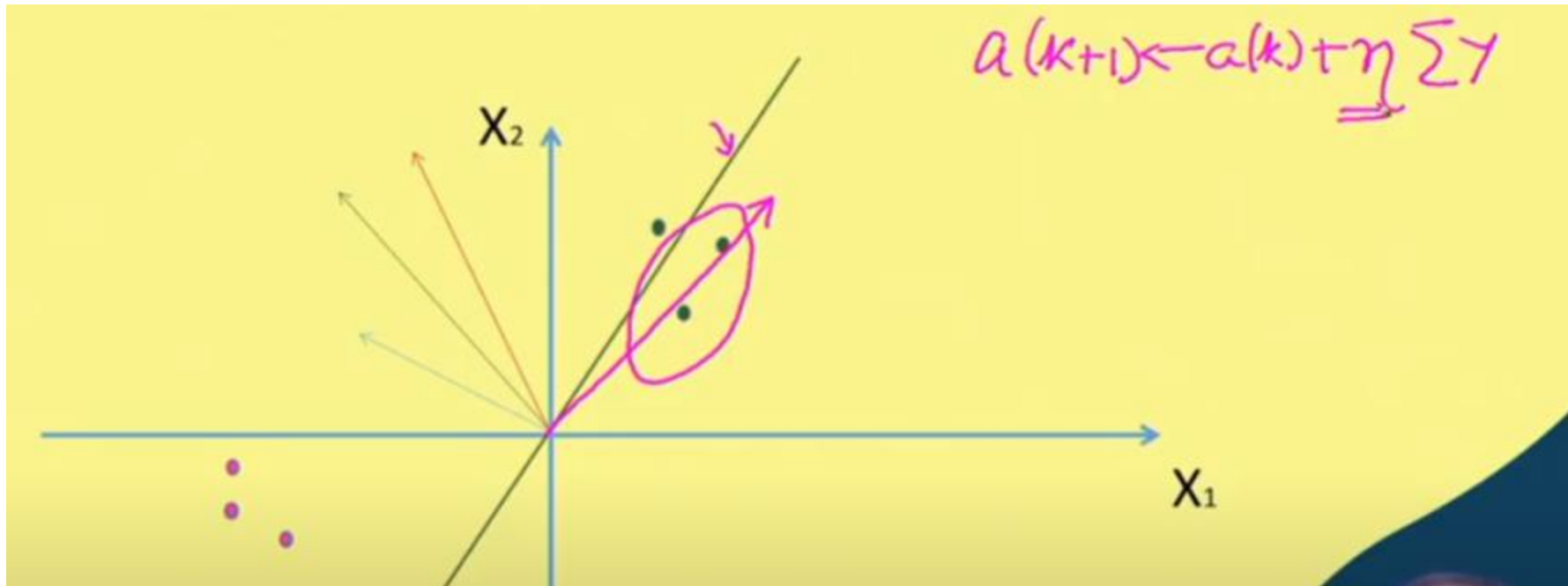


# Linear Classifier: Perceptron



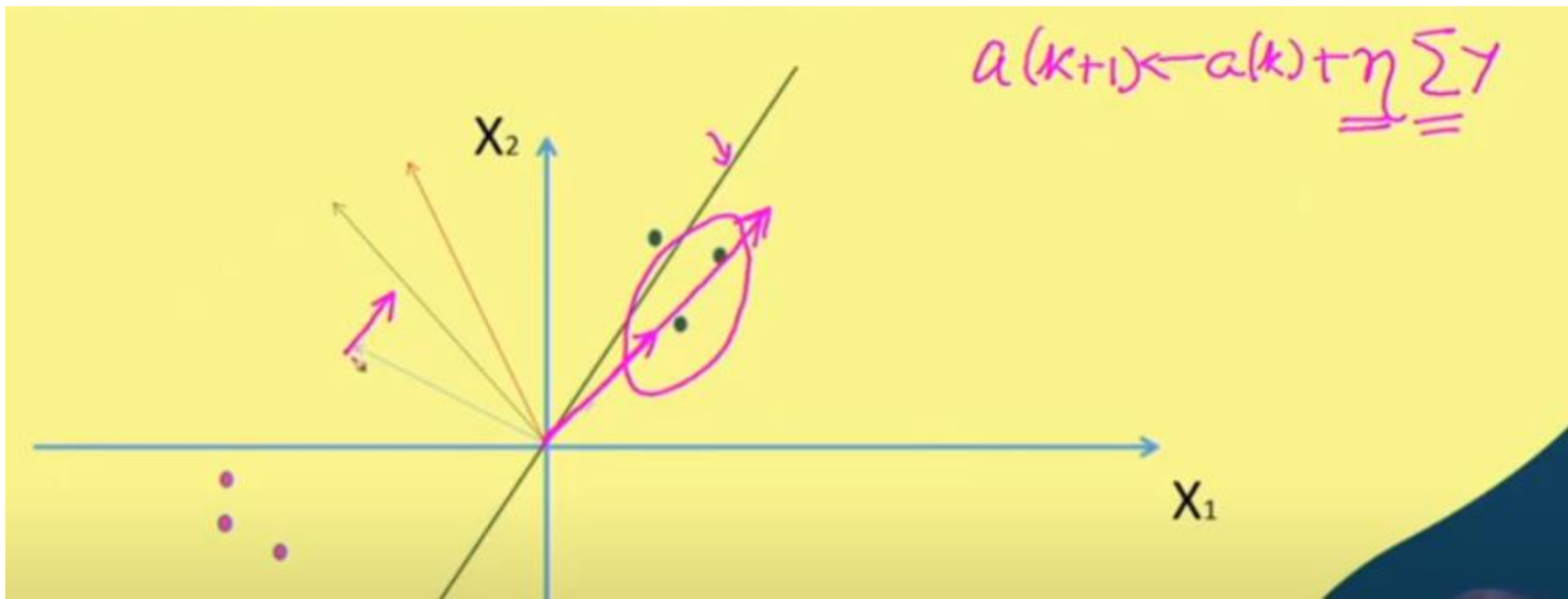
# Linear Classifier: Perceptron

- All mis-classified samples are in the direction shown



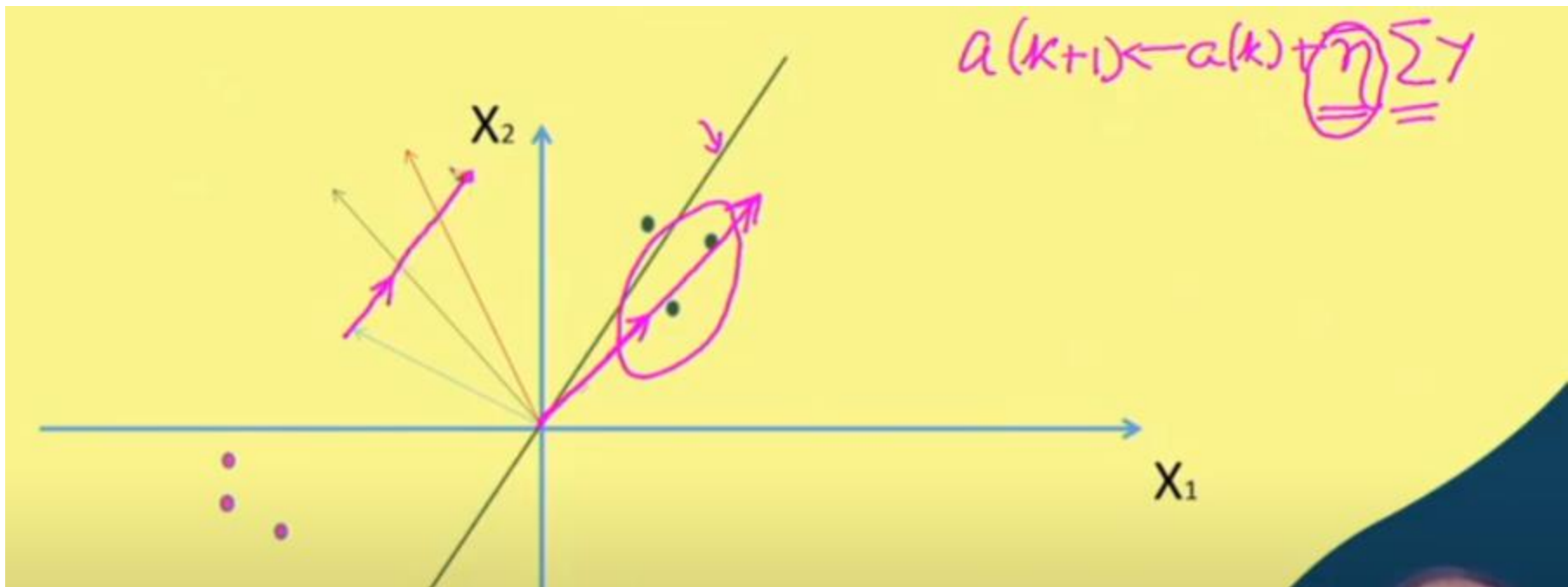
# Linear Classifier: Perceptron

- So,  $a(0)$  is moved in the direction as shown and slowly it land in the solution region.



# Linear Classifier: Perceptron

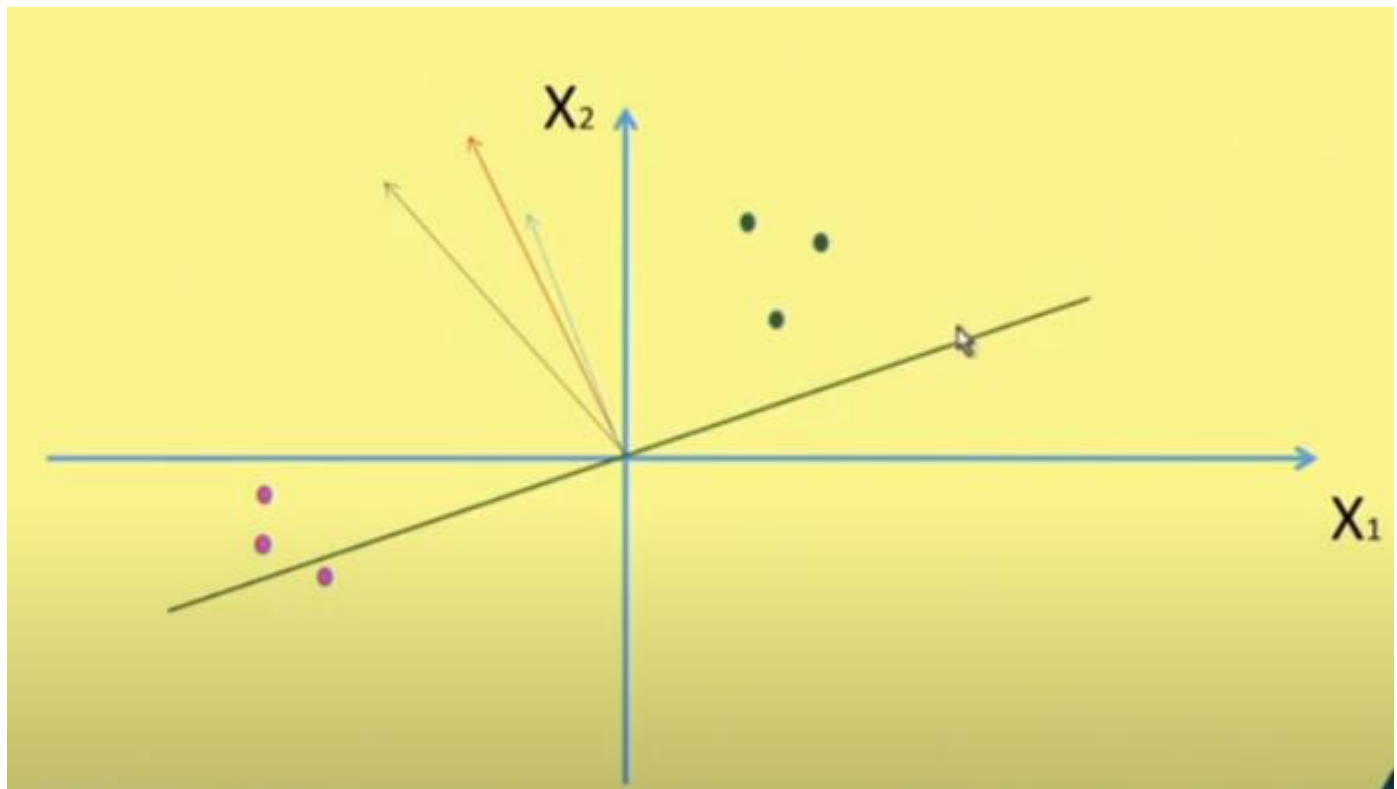
- If the learning rate  $\eta$  is very large then solution vector is over shot.



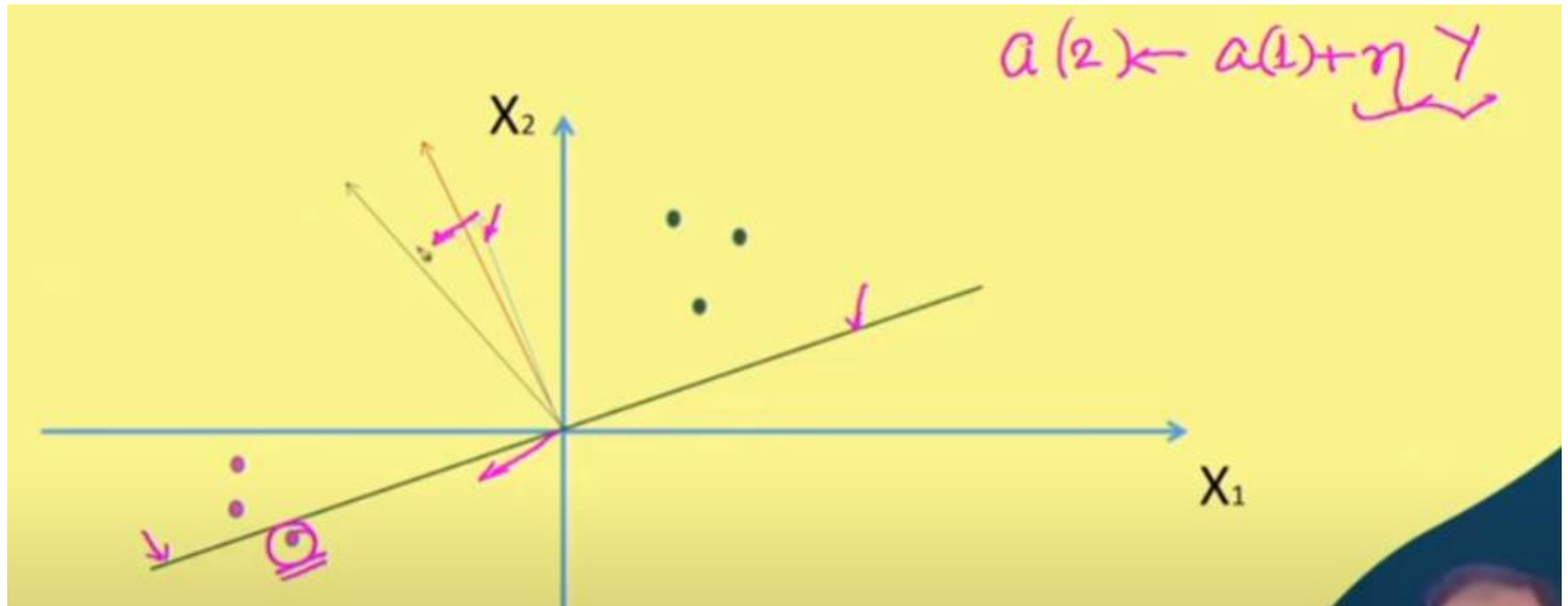


# Linear Classifier: Perceptron

- If the learning rate  $\eta$  is very large then solution vector is crossed the solution region.

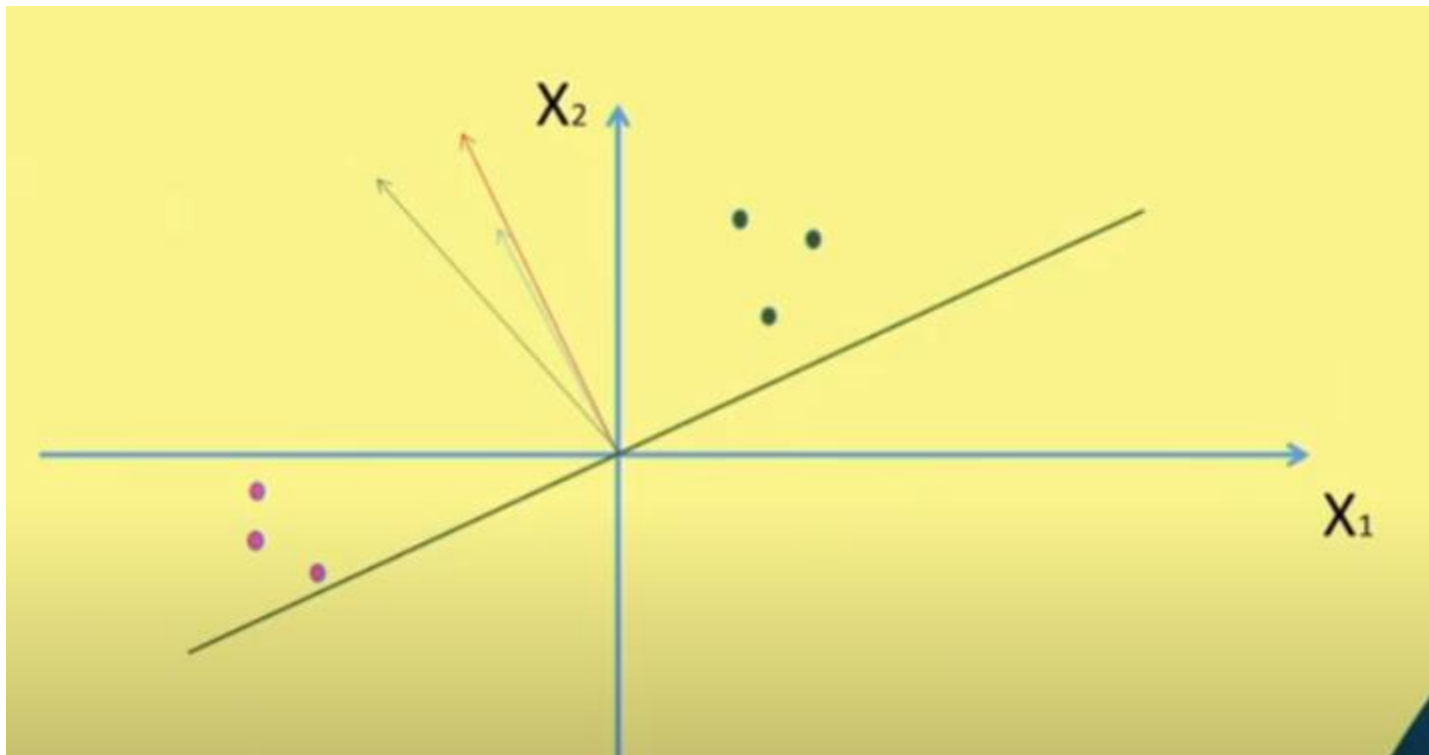


# Linear Classifier: Perceptron



# Linear Classifier: Perceptron

- This approach clearly says that by taking gradient descent approach it is possible the weight vector converge in the solution region.



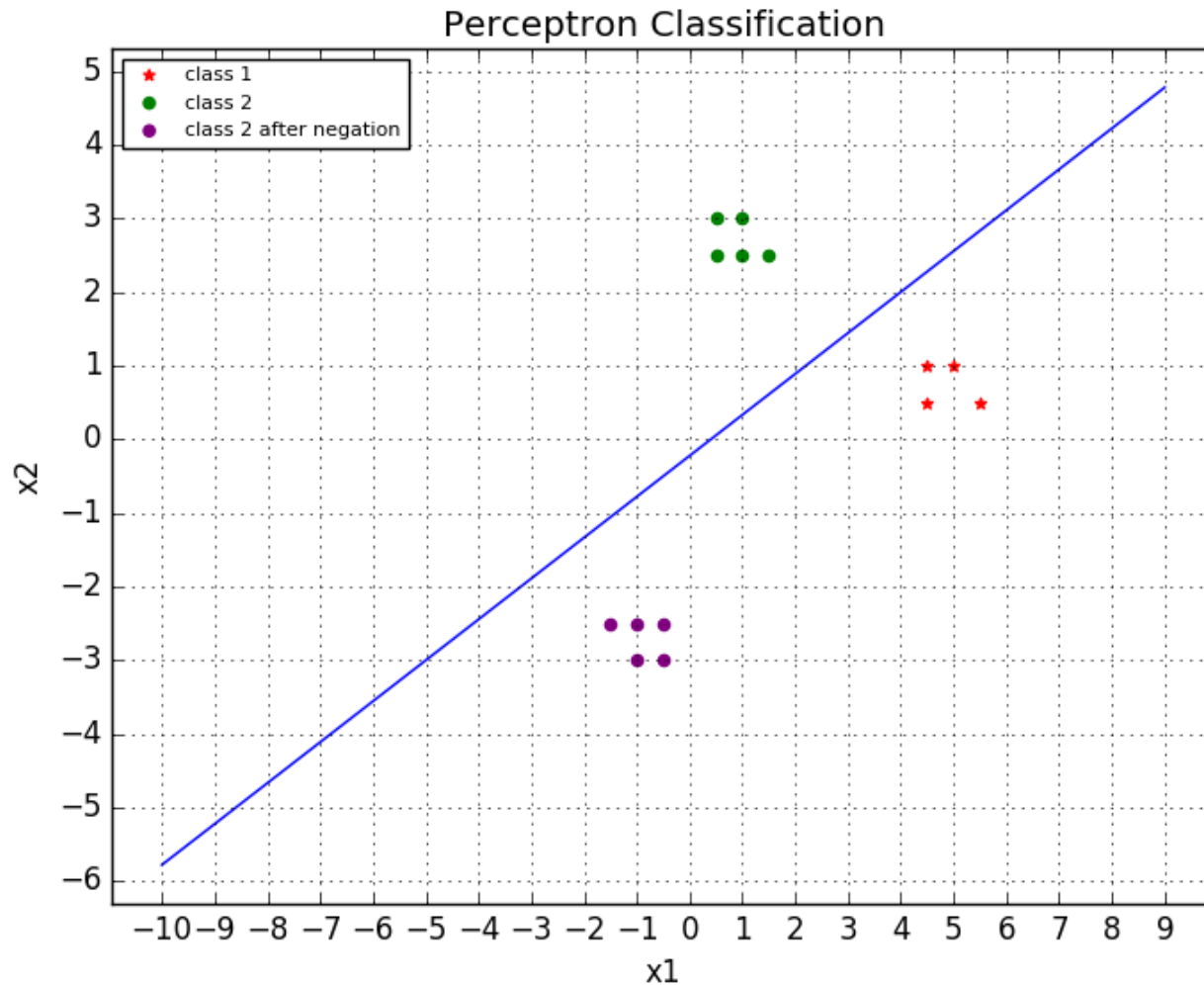
# Perceptron criteria

- The approach can be  $J(a) = \sum - (a^T y)$
- Using **gradient decent algorithm**
- It works well if the **classes are linearly separable**

# MSE criteria

- If the classes are linearly separable, then it is possible to have linear decision boundary like this
- If the classes are not linearly separable then the approach is based on **minimum error criteria (Mean Square Error criteria)**

# Linear Classifier: Perceptron



# Summary

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- Convergence Procedure of Perceptron Algorithm



**Thank you**