#### Question1:

How is Soft Margin Classifier different from Maximum Margin Classifier?

# Answer 1:

- 1. Maximum Margin Classifier separates two classes perfectly but Soft Margin Classifier allows certain points in both classes to be deliberately misclassified.
- 2. Soft Margin Classifier can work for linear as well as non linear data but Maximum Margin Classifier works well only for Linear data.
- 3. Soft Margin Classifier works equally good with train and test data but Maximum Margin Classifier works poorly with test data(unseen data)
- 4. Soft Margin Classifier can handle noise class boundaries but Maximum Margin Classifier fails to do so.

# Question 2:

What does the slack variable Epsilon ( $\epsilon$ ) represent?

## Answer 2:

A slack variable( $\epsilon$ ) tells you where an observation is located relative to the margin and hyperplane. It is used to control **misclassifications**.

Each data point has a slack value associated to it, according to where the point is located. Generally lower values of slack are better than higher values

The value of slack lies between 0 and +infinity where

- 1. slack = 0 implies a correct classification
- 2. slack > 1 implies an incorrect classification
- 3. slack within 0 and 1 classifies correctly but violates the margin

# Question 3:

How do you measure the cost function in SVM? What does the value of C signify?

#### Answer3:

The cost function in SVM is measured by C - the sum of all the values of slack variables (epsilons) of each data point. It is denoted as.

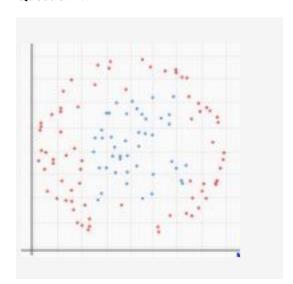
∑ei ≤ C

The parameter cost of misclassification (C) represents the cost of violations to the margin and the hyperplane. Thus, the parameter C is called the tuning parameter.

If C is **large**, the slack variables (epsilons( $\epsilon$ )) can be large, i.e. you allow a larger number of data points to be misclassified or violate the margin

If C is **small**, you force the individual slack variables to be small, i.e. you do not allow many data points to fall on the wrong side of the margin or the hyperplane

## Question 4:



Given the above dataset where red and blue points represent the two classes, how will you use SVM to classify the data?

# Answer 4:

- 1. From graph, it is clear that both classes cannot be separated by linear SVC model.
- 2. So First we should check the accuracy of different non linear kernels like rbf, poly etc. The Non linear model providing highest accuracy should be considered.
- 3. With Help of GridSearch, find the combination of gamma and C which provides highest accuracy
- 4. With Obtained SVC Kernel, Gamma & C we can build a model.
- 5. This model will classify the dataset into two categories

# Question 5:

What do you mean by feature transformation?

#### Answer 5:

The process of transforming the original attributes into a new feature space is called Feature Transformation.

These new features may not have the same interpretation as the original features, but they may have more discriminatory power in a different space than the original space.

Some common techniques for Feature Transformation are:

- 1. Scaling or normalizing features within a range, say between 0 to 1.
- 2. Principle Component Analysis or SVD.
- 3. SVM also transforms features internally.
- 4. Transforming categorical features to numerical.