MCA Sem I

# Practical No. 6

# **Title: Implementation of Support Vector Machine or SVM**

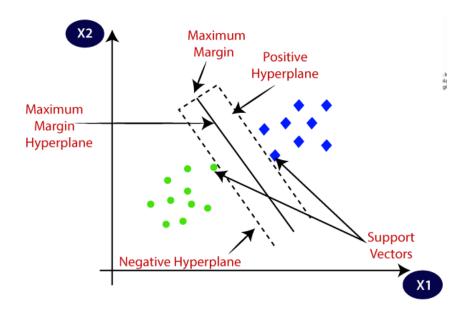
Aim: Understanding basics of Bagging Algorithm: Random Forest

#### **Introduction:**

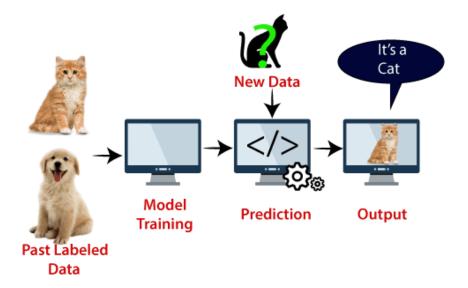
Support Vector Machine or SVM is one of the most popular Supervised Learning algorithms, which is used for Classification as well as Regression problems. However, primarily, it is used for Classification problems in Machine Learning.

The goal of the SVM algorithm is to create the best line or decision boundary that can segregate n-dimensional space into classes so that we can easily put the new data point in the correct category in the future. This best decision boundary is called a hyperplane.

SVM chooses the extreme points/vectors that help in creating the hyperplane. These extreme cases are called support vectors, and hence the algorithm is termed as Support Vector Machine. Consider the below diagram in which there are two different categories that are classified using a decision boundary or hyperplane:



Example: SVM can be understood with the example that we have used in the KNN classifier. Suppose we see a strange cat that also has some features of dogs, so if we want a model that can accurately identify whether it is a cat or dog, so such a model can be created by using the SVM algorithm. We will first train our model with lots of images of cats and dogs so that it can learn about different features of cats and dogs, and then we test it with this strange creature. So as support vector creates a decision boundary between these two data (cat and dog) and choose extreme cases (support vectors), it will see the extreme case of cat and dog. On the basis of the support vectors, it will classify it as a cat. Consider the below diagram:



SVM algorithm can be used for Face detection, image classification, text categorization, etc.

#### Types of SVM

SVM can be of two types:

 Linear SVM: Linear SVM is used for linearly separable data, which means if a dataset can be classified into two classes by using a single straight line, then such data is termed as linearly separable data, and classifier is used called as Linear SVM classifier. Non-linear SVM: Non-Linear SVM is used for non-linearly separated data, which
means if a dataset cannot be classified by using a straight line, then such data is
termed as non-linear data and classifier used is called as Non-linear SVM classifier.

#### Hyperplane and Support Vectors in the SVM algorithm:

Hyperplane: There can be multiple lines/decision boundaries to segregate the classes in n-dimensional space, but we need to find out the best decision boundary that helps to classify the data points. This best boundary is known as the hyperplane of SVM.

The dimensions of the hyperplane depend on the features present in the dataset, which means if there are 2 features (as shown in image), then hyperplane will be a straight line. And if there are 3 features, then hyperplane will be a 2-dimension plane.

We always create a hyperplane that has a maximum margin, which means the maximum distance between the data points.

### Support Vectors:

The data points or vectors that are the closest to the hyperplane and which affect the position of the hyperplane are termed as Support Vector. Since these vectors support the hyperplane, hence called a Support vector.

#### Exercise -

Support Vector Machine Tutorial Using Python Sklearn

## Implementation:

### **Program:**

```
import pandas as pd
import numpy as np
```

```
url = 'http://archive.ics.uci.edu/ml/machine-learning-
databases/iris/iris.data'
iris = pd.read csv(url, names=['sepal length', 'sepal width',
'petal_length', 'petal_width', 'class'])
             iris.head()
               sepal_length sepal_width petal_length petal_width
                                                         class
                      5.1
                                3.5
                                          1.4
                                                   0.2 Iris-setosa
             1
                      4.9
                                3.0
                                          1.4
                                                   0.2 Iris-setosa
             2
                      4.7
                                3.2
                                          1.3
                                                   0.2 Iris-setosa
             3
                      4.6
                                3.1
                                          1.5
                                                   0.2 Iris-setosa
                      5.0
                                3.6
                                          1.4
                                                   0.2 Iris-setosa
from sklearn.model_selection import train_test_split
X train, X test, y train, y test =
train test split(iris.drop(['class'],axis='columns'),iris[['clas
s']],train size=0.8)
len(X train)
from sklearn.svm import SVC
model = SVC()
model.fit(X_train, y_train)
Output:
model.predict([[4.8,3.0,1.5,0.3]])
array(['Iris-setosa'], dtype=object)
```