

SVM

- Support vector machine is nothing but supervised machine learning classification technique.
- It helps to classify the classes with the help of the hyperplane.
- The hyperplane is nothing but a line that bisects the two classes or more classes with minimum squared error.
- In the surrounding of the hyperplane, we have the margin which is equidistantly placed on both sides of the plane.
- Whichever datapoint goes through that margin, we suppose to call it a vector.
- By using that vector, we can classify the two classes with least errors.

- But not always we find the classes that can be bisected very easily by using a straight line.
- Suppose if we have the polynomial or the circle datapoint, that a simple straight line will not properly.
- So to get the best classification, we use the kernel tricks.

- Kernel tricks help the algorithm to classify the such complex distribution of the data by enhancing the dimensions of the data.

- Let's us consider we have the linear circle type of the data to classify that we have to use the kernel technique which adds the dimensionality of the model and helps to classify the datapoint with least squared error.

- **Kernel Type:-**

- Linear Kernel.
- Polynomial Kernel.
- RBF Kernel.
- sigmoid kernel.

At the polynomial degree of the polynomial kernel works as linear kernel.

```
In [70]: import numpy as np
import pandas as pd
import seaborn as sns

import warnings
warnings.filterwarnings('ignore')

In [71]: from sklearn.metrics import classification_report, confusion_matrix, mean_squared_error, accuracy_score, recall_score
from sklearn.svm import SVC
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn import datasets

In [72]: cancer = datasets.load_iris()

In [73]: cancer.target_names

Out[73]: array(['setosa', 'versicolor', 'virginica'], dtype='<U10')

In [74]: cancer.feature_names

Out[74]: ['sepal length (cm)',
'sepal width (cm)',
'petal length (cm)',
'petal width (cm)']

In [75]: x = cancer.data
y = cancer.target

In [76]: x.shape, y.shape

Out[76]: ((150, 4), (150,))

In [77]: scaler = StandardScaler()
x = scaler.fit_transform(x)

In [78]: x_train, x_test, y_train, y_test = train_test_split(x, y, train_size=0.80, random_state=42, stratify=y)

In [79]: clf = SVC(kernel='linear')
clf.fit(x_train, y_train)
y_pred = clf.predict(x_test)
print('Accuracy : ', accuracy_score(y_test, y_pred))

Accuracy : 1.0

In [83]: print("Precision Score : ", precision_score(y_test, y_pred,
                                                    pos_label='positive',
                                                    average='micro'))

print("Precision Score : ", recall_score(y_test, y_pred,
                                         pos_label='positive',
                                         average='micro'))

Precision Score : 1.0
Precision Score : 1.0

In [ ]:
```