#### **Assignment 4- Support Vector Machine**

#### **Problem Statement-**

Assignment on Classification technique Download Data Set Available on kaggle (The last column of the dataset needs to be changed to 0 or 1)Data Set:

https://www.kaggle.com/mohansacharya/graduate-admissions The counselor of the firm is supposed check whether the student will get an admission or not based on his/her GRE score and Academic Score. So to help the counselor to take appropriate decisions build a machine learning model classifier using SVM to predict whether a student will get admission or not.

A.Apply Data pre-processing (Label Encoding, Data Transformation....) techniques if necessary. B.Perform data-preparation (Train-Test Split) C. Apply Machine Learning Algorithm D. Evaluate Model.

#### Importing python libraries

```
In [1]: import seaborn as sns
  import pandas as pd
  import matplotlib.pyplot as plt
  from sklearn import *
```

#### loading the csv file into a dataframe

```
In [2]: A=pd.read_csv(r"C:\Users\HP\Downloads\Admission_Predict.csv")
A
```

Out[2]:		Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
	0	1	337	118	4	4.5	4.5	9.65	1	0.92
	1	2	324	107	4	4.0	4.5	8.87	1	0.76
	2	3	316	104	3	3.0	3.5	8.00	1	0.72
	3	4	322	110	3	3.5	2.5	8.67	1	0.80
	4	5	314	103	2	2.0	3.0	8.21	0	0.65
	•••									
	395	396	324	110	3	3.5	3.5	9.04	1	0.82
	396	397	325	107	3	3.0	3.5	9.11	1	0.84
	397	398	330	116	4	5.0	4.5	9.45	1	0.91
	398	399	312	103	3	3.5	4.0	8.78	0	0.67
	399	400	333	117	4	5.0	4.0	9.66	1	0.95

400 rows × 9 columns

## head() function used to access the first n rows of a dataframe

[3]:	A.h	ead(5)								
[3]:		Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
	0	1	337	118	4	4.5	4.5	9.65	1	0.92
	1	2	324	107	4	4.0	4.5	8.87	1	0.76
	2	3	316	104	3	3.0	3.5	8.00	1	0.72
	3	4	322	110	3	3.5	2.5	8.67	1	0.80
	4	5	314	103	2	2.0	3.0	8.21	0	0.65

## describe() function returns the description of data in dataframe

In [4]: A.describe()

Out[4]:		Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Rese
	count	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.000000	400.00
	mean	200.500000	316.807500	107.410000	3.087500	3.400000	3.452500	8.598925	0.54
	std	115.614301	11.473646	6.069514	1.143728	1.006869	0.898478	0.596317	0.49
	min	1.000000	290.000000	92.000000	1.000000	1.000000	1.000000	6.800000	0.00
	25%	100.750000	308.000000	103.000000	2.000000	2.500000	3.000000	8.170000	0.00
	50%	200.500000	317.000000	107.000000	3.000000	3.500000	3.500000	8.610000	1.00
	75%	300.250000	325.000000	112.000000	4.000000	4.000000	4.000000	9.062500	1.00
	max	400.000000	340.000000	120.000000	5.000000	5.000000	5.000000	9.920000	1.00
4									<b>•</b>

## dtype function returns the data type of each column

```
In [5]: A.dtypes
        Serial No.
                                 int64
Out[5]:
        GRE Score
                                int64
        TOEFL Score
                                int64
        University Rating
                                int64
        SOP
                              float64
        LOR
                              float64
        CGPA
                              float64
        Research
                                int64
        Chance of Admit
                              float64
        dtype: object
```

## counting the total number of null values in each column

```
In [6]: A.isnull().sum()
        Serial No.
                               0
Out[6]:
        GRE Score
                               0
        TOEFL Score
        University Rating
                               0
        SOP
                               0
        LOR
                              0
        CGPA
                               0
        Research
                               0
        Chance of Admit
        dtype: int64
```

#### Dropping the column "Serial No"

```
In [7]: A.drop('Serial No.',axis=1)
```

Out[7]:		GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
	0	337	118	4	4.5	4.5	9.65	1	0.92
	1	324	107	4	4.0	4.5	8.87	1	0.76
	2	316	104	3	3.0	3.5	8.00	1	0.72
	3	322	110	3	3.5	2.5	8.67	1	0.80
	4	314	103	2	2.0	3.0	8.21	0	0.65
	•••								
	395	324	110	3	3.5	3.5	9.04	1	0.82
	396	325	107	3	3.0	3.5	9.11	1	0.84
	397	330	116	4	5.0	4.5	9.45	1	0.91
	398	312	103	3	3.5	4.0	8.78	0	0.67
	399	333	117	4	5.0	4.0	9.66	1	0.95

400 rows × 8 columns

## Changing the values of column "Chance of Admit" to 0 and 1

## Assigning the values of independent and dependent variables

```
In [9]: x=A[["GRE Score","TOEFL Score", "University Rating","SOP","CGPA","Research"]]
y=A["Chance of Admit"]
```

## Dividing the dataset into training and testing data

```
In [10]: from sklearn.model_selection import train_test_split
```

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
```

#### feature transformation

```
In [11]: from sklearn.preprocessing import StandardScaler
    sc=StandardScaler()

    x_train=sc.fit_transform(x_train)
    x_test=sc.fit_transform(x_test)
```

# creating the object of kernel="linear" using SVC class along with fitting the model and calculating the accuracy of model

```
In [12]: from sklearn.svm import SVC
    model=SVC(kernel="linear",random_state=0)
    model.fit(x_train,y_train)

    x_pred=model.predict(x_test)

    print("Accuracy: ",metrics.accuracy_score(y_test, x_pred))

Accuracy: 0.9
```

## creating the object of kernel= "rbf" using SVC class along with fitting the model and calculating the accuracy of model

```
In [13]: from sklearn.svm import SVC
   model_rbf=SVC(kernel="rbf",gamma=20,C=7.0,random_state=0)
   model_rbf.fit(x_train,y_train)

x_pred=model_rbf.predict(x_test)

print("Accuracy: ",metrics.accuracy_score(y_test, x_pred))
```

Accuracy: 0.75

## creating the object of kernel = "poly" using SVC class along with fitting the model and calculating the accuracy of model

```
In [14]: from sklearn.svm import SVC
model_poly=SVC(kernel="poly",degree=4)
model_poly.fit(x_train,y_train)

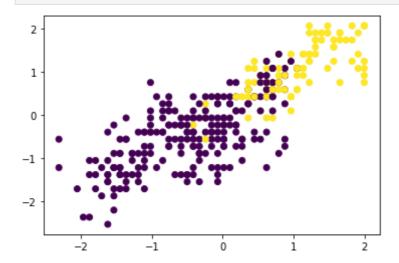
x_pred=model_poly.predict(x_test)
```

```
print("Accuracy: ",metrics.accuracy_score(y_test, x_pred))
```

Accuracy: 0.8

### Plotting the x\_train values

```
In [15]: plt.scatter(x_train[:, 0],x_train[:, 1],c=y_train)
    plt.show()
```



#### Plotting the x\_test values

```
In [16]: plt.scatter(x_test[:, 0],x_test[:, 1],c=y_test)
    plt.show()
```

