# Day-10

### **Placement Statement**

#### In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

#### In [2]:

```
d=pd.read_csv(r"C:\Users\user\Downloads\placement.csv")
d
```

#### Out[2]:

	cgpa	placement_exam_marks	placed
0	7.19	26.0	1
1	7.46	38.0	1
2	7.54	40.0	1
3	6.42	8.0	1
4	7.23	17.0	0
995	8.87	44.0	1
996	9.12	65.0	1
997	4.89	34.0	0
998	8.62	46.0	1
999	4.90	10.0	1

1000 rows × 3 columns

#### In [3]:

```
d.columns
```

#### Out[3]:

```
Index(['cgpa', 'placement_exam_marks', 'placed'], dtype='object')
```

```
In [4]:
d.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 3 columns):
     Column
                            Non-Null Count Dtype
 0
     cgpa
                            1000 non-null
                                            float64
 1
     placement_exam_marks 1000 non-null
                                            float64
     placed
                            1000 non-null
                                            int64
dtypes: float64(2), int64(1)
memory usage: 23.6 KB
In [5]:
x=d[['cgpa', 'placement_exam_marks']]
y=d['placed']
In [6]:
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
In [7]:
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
Out[7]:
LinearRegression()
In [8]:
print(lr.intercept_)
0.2573738696988293
In [9]:
print(lr.score(x_test,y_test))
-0.00903257211520403
In [10]:
print(lr.score(x_train,y_train))
```

### **Ridge Regression**

0.004028706285135186

```
In [11]:
from sklearn.linear_model import Ridge,Lasso

In [12]:
rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
rr.score(x_test,y_test)

Out[12]:
-0.008808585617939402
```

## **Lasso Regression**

```
In [13]:
la=Lasso(alpha=10)

In [14]:
la.fit(x_train,y_train)
Out[14]:
Lasso(alpha=10)

In [15]:
la.score(x_test,y_test)
Out[15]:
```

# **Elastic Regreesion**

-0.001677172046648634

ElasticNet()

```
In [16]:

from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)

Out[16]:
```

#### In [17]:

```
predict=(en.predict(x_test))
print(predict)
```

```
[0.48285714 0.48285714 0.48285714 0.48285714 0.48285714 0.48285714
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```

```
In [18]:
```

```
print(en.score(x_test,y_test))
```

-0.001677172046648634

### **Evaluation Method**

```
In [19]:
from sklearn import metrics

In [20]:
print("Mean Absolute Error:",metrics.mean_absolute_error(y_test,predict))
Mean Absolute Error: 0.5001142857142857

In [21]:
print("Root Mean Square Error:",np.sqrt(metrics.mean_squared_error(y_test,predict)))
Root Mean Square Error: 0.500407996803914

In [22]:
print("Mean Square Error:",metrics.mean_squared_error(y_test,predict))
Mean Square Error: 0.25040816326530607

In []:
```