

GRIPJUNE21 @ The Sparks Foundation

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Pediction using Supervised Learning

Importing the Packages

```
In [1]: ▶ import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

Reading the Data Set

```
In [2]: ▶ url = "https://raw.githubusercontent.com/AdiPersonalWorks/Random/master/student_scores%20-%20student_scores."
```

```
In [3]: ▶ df = pd.read_csv(url)
```

```
In [4]: ▶ df.head()
```

Out[4]:

	Hours	Scores
0	2.5	21
1	5.1	47
2	3.2	27
3	8.5	75
4	3.5	30

```
In [5]: ▶ # Exploratory Data Analysis
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 25 entries, 0 to 24
Data columns (total 2 columns):
#   Column  Non-Null Count  Dtype  
---  -
0   Hours    25 non-null      float64
1   Scores   25 non-null      int64   
dtypes: float64(1), int64(1)
memory usage: 528.0 bytes
```

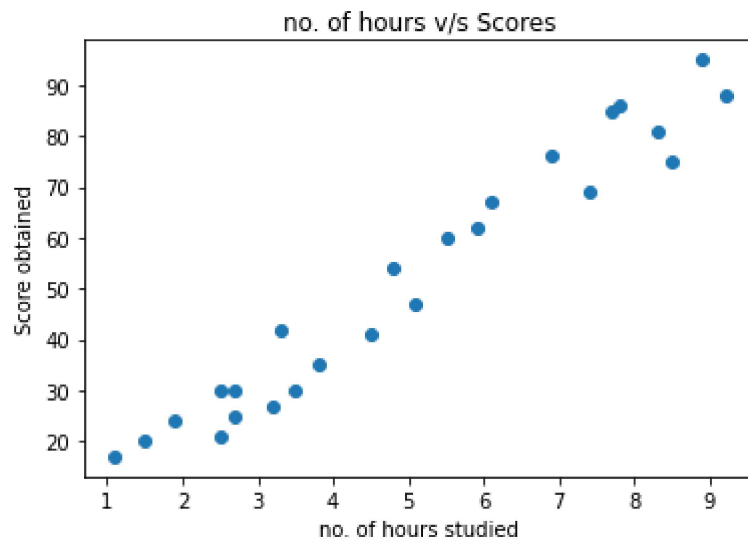
```
In [6]: ▶ df.describe()
```

Out[6]:

	Hours	Scores
count	25.000000	25.000000
mean	5.012000	51.480000
std	2.525094	25.286887
min	1.100000	17.000000
25%	2.700000	30.000000
50%	4.800000	47.000000
75%	7.400000	75.000000
max	9.200000	95.000000

Data Visualization

```
In [8]: ▶ # visualizing the data using scatter plot
plt.scatter(x="Hours",y="Scores",data=df)
plt.title('no. of hours v/s Scores')
plt.xlabel('no. of hours studied')
plt.ylabel('Score obtained')
plt.show()
```



Feature Selection

```
In [9]: ▶ from sklearn.model_selection import train_test_split
```

```
In [10]: ▶ x = df[["Hours"]]
          ▶ y = df["Scores"]
```

Training and Testing the Data

```
In [11]: x_train,x_test,y_train,y_test = train_test_split(x, y,test_size=0.25, random_state=0 )
```

```
In [12]: x_train.head()
```

```
Out[12]:
```

	Hours
24	7.8
23	6.9
14	1.1
1	5.1
10	7.7

Training the Model

```
In [13]: from sklearn.linear_model import LinearRegression
```

```
In [14]: lm = LinearRegression()
```

Fitting the Data

```
In [15]: lm.fit(x_train,y_train)
```

```
Out[15]: LinearRegression()
```

```
In [16]: lm.intercept_
```

```
Out[16]: 1.932204253151646
```

```
In [17]: lm.coef_
```

```
Out[17]: array([9.94167834])
```

Predicting the Data

```
In [18]: ▶ predict = lm.predict([[9.25]])
```

```
In [19]: ▶ predict
```

```
Out[19]: array([93.89272889])
```

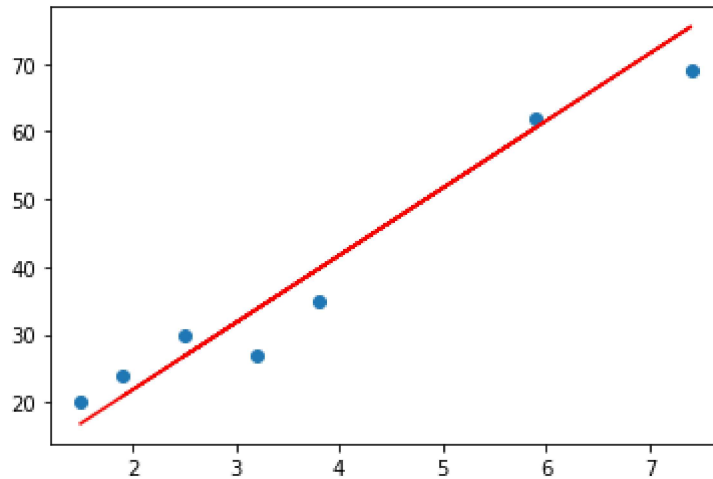
```
In [20]: ▶ p = lm.predict(x_test)
```

```
In [21]: ▶ # Table for predicted values and tested values  
df1=pd.DataFrame({'Actual':y_test, 'Predicted':p})  
df1
```

```
Out[21]:
```

	Actual	Predicted
5	20	16.844722
2	27	33.745575
19	69	75.500624
16	30	26.786400
11	62	60.588106
22	35	39.710582
17	24	20.821393

```
In [22]: ▶ # Plotting the best fit line  
plt.scatter(x_test, y_test)  
plt.plot(x_test, p, color='red')  
plt.show()
```



Evaluating the Data

```
In [23]: ▶ from sklearn import metrics
```

```
In [24]: ▶ r1=metrics.mean_absolute_error(y_test,p)
print("Mean Absolute Error      : ",r1)

Mean Absolute Error      :  4.130879918502486
```

```
In [25]: ▶ r2=metrics.mean_squared_error(y_test,p)
print("Mean Squared Error       : ",r2)

Mean Squared Error       :  20.33292367497997
```

```
In [26]: ▶ r3=np.sqrt(r2)
print("Root Mean Squared Error : ", r3)

Root Mean Squared Error :  4.5092043283688055
```

```
In [27]: ▶ #answer for the question
hours=9.35
print(f"Number of hours of study are {hours}")
print(f"Predicted score {predict}")

Number of hours of study are 9.35
Predicted score [93.89272889]
```

Conclusion

```
In [ ]: ▶ I was successf
```