THE SPARKS FOUNDATION

TASK 1 - Prediction using Supervised ML

To Predict the percentage of marks of the students based on the number of hours they studied

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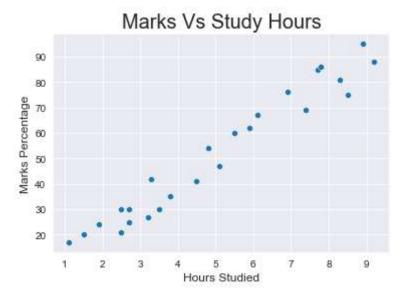
```
In [1]:
         # importing the required libraries
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.model selection import train test split
         from sklearn.linear_model import LinearRegression
         from sklearn.metrics import mean absolute error
In [2]:
         # Reading the Data
         data = pd.read csv('http://bit.ly/w-data')
         data.head(5)
           Hours Scores
Out[2]:
        0
              2.5
                     21
              5.1
                     47
```

```
3.2
                27
3
                75
      8.5
4
      3.5
                30
```

```
In [3]:
         # Check if there any null value in the Dataset
         data.isnull == True
        False
Out[3]:
```

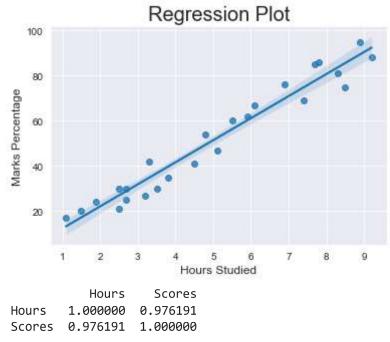
There is no null value in the Dataset so, we can now visualize our Data.

```
In [4]:
         sns.set style('darkgrid')
         sns.scatterplot(y= data['Scores'], x= data['Hours'])
         plt.title('Marks Vs Study Hours', size=20)
         plt.ylabel('Marks Percentage', size=12)
         plt.xlabel('Hours Studied', size=12)
         plt.show()
```



From the above scatter plot there looks to be correlation between the 'Marks Percentage' and 'Hours Studied', Lets plot a regression line to confirm the correlation.

```
In [5]:
    sns.regplot(x= data['Hours'], y= data['Scores'])
    plt.title('Regression Plot',size=20)
    plt.ylabel('Marks Percentage', size=12)
    plt.xlabel('Hours Studied', size=12)
    plt.show()
    print(data.corr())
```



It is confirmed that the variables are positively correlated.

Training the Model

1) Splitting the Data

```
In [6]: # Defining X and y from the Data
```

```
X = data.iloc[:, :-1].values
y = data.iloc[:, 1].values

# Spliting the Data in two
train_X, val_X, train_y, val_y = train_test_split(X, y, random_state = 0)
```

2) Fitting the Data into the model

```
In [7]:
    regression = LinearRegression()
    regression.fit(train_X, train_y)
    print("------Model Trained------")
------Model Trained-------
```

Predicting the Percentage of Marks

```
pred_y = regression.predict(val_X)
prediction = pd.DataFrame({'Hours': [i[0] for i in val_X], 'Predicted Marks': [k for k prediction
```

Out[8]:		Hours	Predicted Marks
	0	1.5	16.844722
	1	3.2	33.745575
	2	7.4	75.500624
	3	2.5	26.786400
	4	5.9	60.588106
	5	3.8	39.710582
	6	1.9	20.821393

Comparing the Predicted Marks with the Actual Marks

```
In [9]:
    compare_scores = pd.DataFrame({'Actual Marks': val_y, 'Predicted Marks': pred_y})
    compare_scores
```

Out[9]:		Actual Marks	Predicted Marks
	0	20	16.844722
	1	27	33.745575
	2	69	75.500624
	3	30	26.786400
	4	62	60.588106
	5	35	39.710582
	6	24	20.821393

Visually Comparing the Predicted Marks with the Actual Marks

```
In [10]:
    plt.scatter(x=val_X, y=val_y, color='blue')
    plt.plot(val_X, pred_y, color='Black')
    plt.title('Actual vs Predicted', size=20)
    plt.ylabel('Marks Percentage', size=12)
    plt.xlabel('Hours Studied', size=12)
    plt.show()
```

Actual vs Predicted Actual vs Predicted Actual vs Predicted Actual vs Predicted Actual vs Predicted

Evaluating the Model

Score = 93.893

```
# Calculating the accuracy of the model
print('Mean absolute error: ',mean_absolute_error(val_y,pred_y))
```

Mean absolute error: 4.130879918502486

Small value of Mean absolute error states that the chances of error or wrong forecasting through the model are very less.

What will be the predicted score of a student if he/she studies for 9.25 hrs/ day?

```
In [12]:
    hours = [9.25]
    answer = regression.predict([hours])
    print("Score = {}".format(round(answer[0],3)))
```

According to the regression model if a student studies for 9.25 hours a day he/she is likely to score 93.89 marks.