## **Prediction using Supervised ML**

Prompt

- Predict the percentage of an student based on the no. of study hours.
- This is a simple linear regression task as it involves just 2 variables.
- You can use R, Python, SAS Enterprise Miner or any other tool
- Data can be found at http://bit.ly/w-data
- What will be predicted score if a student studies for 9.25 hrs/ day?
- Sample Solution : <a href="https://bit.ly/2HxiGGI">https://bit.ly/2HxiGGI</a>

#### Social Media

Github - <a href="https://github.com/VarunBhattacharya/TheSparkFoundation\_Supervised\_ML">https://github.com/VarunBhattacharya/TheSparkFoundation\_Supervised\_ML</a>)
Linkedin - <a href="https://www.linkedin.com/in/varunbhattacharya/">https://www.linkedin.com/in/varunbhattacharya/</a> (https://www.linkedin.com/in/varunbhattacharya/)

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### Import necessary libraries

```
In [1]: import numpy as np import pandas as pd import pandas as pd import matplotlib.pyplot as plt

from sklearn.metrics import mean_squared_error, r2_score from sklearn.linear_model import LinearRegression

%matplotlib inline
```

#### Load the Dataset

```
In [2]: df = pd.read_csv('Dataset.csv')
```

### In [3]: df.head()

#### Out[3]:

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

#### Information about dataset

```
In [4]: df.describe()
```

#### Out[4]:

	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

# Training the dataset

```
In [5]: #Differentiating between given train data and outcome data

X = df['Hours'].values.reshape(-1,1)

y = df['Scores'].values.reshape(-1,1)
```

```
In [6]: #Split the data into 70% train and 30% test
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3)
```

## Model - Linear Regression

```
In [7]: lr = LinearRegression()
lr.fit(X_train, y_train)
y_pred = lr.predict(X_test)
In [8]: y_pred
```

```
In [8]: y_pred

Out[8]: array([[69.11660197], [91.47446999], [28.2891908], [26.34502837], [88.5582634], [59.39578979], [28.2891908], [16.62421618]])
```

## Accuracy scores of predicted value

```
In [9]: print('Mean Squarred Error: ', mean_squared_error(y_test,y_pred))
print('Rot Mean Squarred Error: ', np.sqrt(mean_squared_error(y_test,y_pred)))
print('R2 Score: ', r2_score(y_test,y_pred) * 100)

Mean Squarred Error: 18.27898889571674
Rot Mean Squarred Error: 4.275392484506151
R2 Score: 97.7604434156929
```

## Creating a new dataframe with the actual and predicted values

```
In [10]: df_new = pd.DataFrame(('Actual':y_test.flatten(), 'Predicted':y_pred.flatten()))
df_new.head()
```

# Out[10]:

```
        Actual
        Predicted

        0
        76
        69.116602

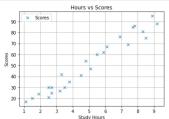
        1
        88
        91.474470

        2
        30
        28.289191

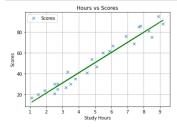
        3
        30
        26.345028
```

95 88.558226





```
In [12]: #Plotting the linear regression line
line = ln.coef_"X + lr.intercept_
df.plot(x = 'Hours', y = 'Scores', style = 'x')
plt.plot(X, line, color = 'green');
plt.title('Hours vs Scores')
plt.xlabel('Stdy Hours')
plt.ylabel('Scores')
plt.grid()
plt.show()
```



### Predicted score

In [13]: lrCoef = list(lr.coef\_)
lrIntr = list(lr.intercept)
regLine = 'scores = ' + str(lrCoef[0][0]) + ' \* hours + ' + str(lrIntr[0])
print(f'The regression line for the above scenario is: {regLine}.')

The regression line for the above scenario is: scores = 9.72081218274112 \* hours + 2.0429979098238107.

In [14]: givenHours = 9.25

In [15]: predScore = lrCoef[0][0] \* givenHours + lrIntr[0] print(f'The predicted score for study of {givenHours} hours is: {predScore}.')

The predicted score for study of 9.25 hours is: 91.96051060017918.

Done By: Varun Bhattacharya