


Pridiction using Supervised ML



Prediction using Supervised ML

(Level - Beginner)

- 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 : <https://bit.ly/2HxiGGI>
- Task submission:
 1. Host the code on GitHub Repository (public). Record the code and output in a video. Post the video on YouTube
 2. Share links of code (GitHub) and video (YouTube) as a post on **YOUR LinkedIn profile**, not TSF Network.
 3. Submit the LinkedIn link in Task Submission Form when shared.

Importing Libraries

```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

Read the Datasets

```
In [ ]: df = pd.read_csv("Datasets.txt")
```

```
In [ ]: df
```

Out[]:

	Hours	Scores
0	2.5	21
1	5.1	47
2	3.2	27
3	8.5	75
4	3.5	30
5	1.5	20
6	9.2	88
7	5.5	60
8	8.3	81
9	2.7	25
10	7.7	85
11	5.9	62
12	4.5	41
13	3.3	42
14	1.1	17
15	8.9	95
16	2.5	30
17	1.9	24
18	6.1	67
19	7.4	69
20	2.7	30
21	4.8	54
22	3.8	35
23	6.9	76
24	7.8	86

Describe()

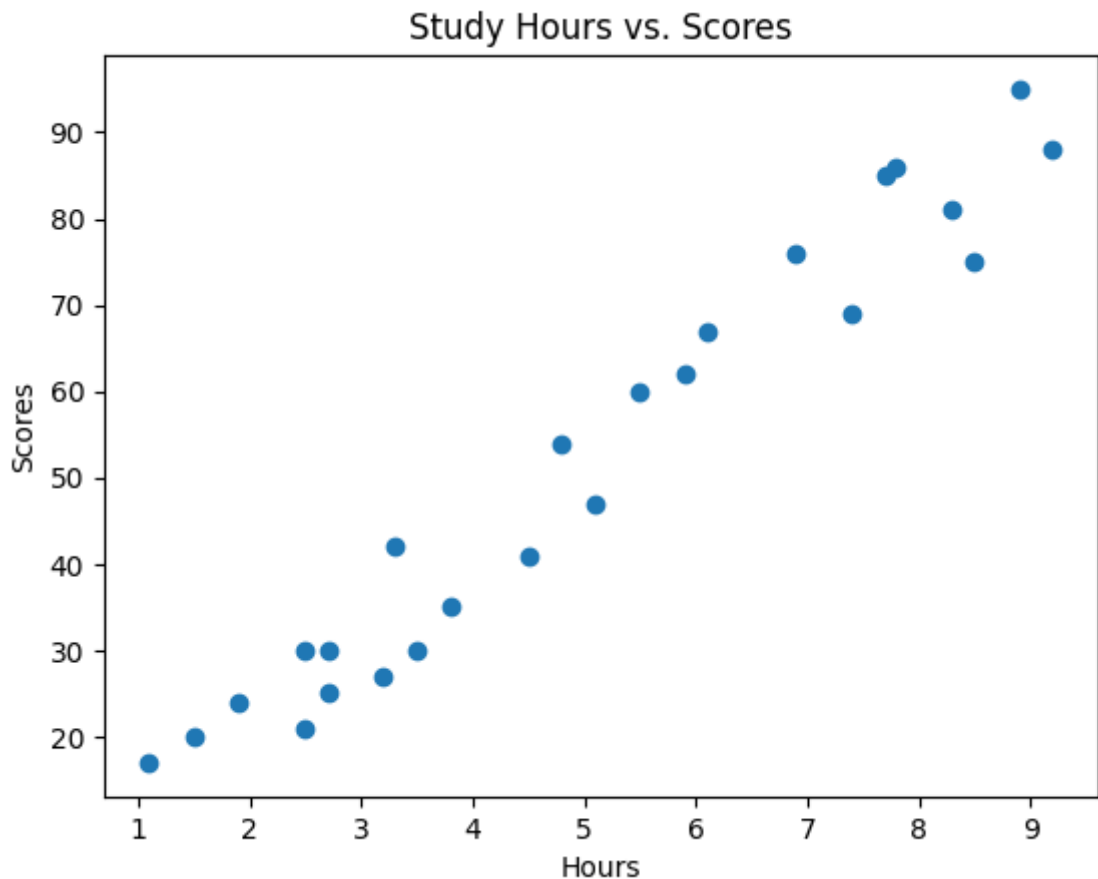
In []: `df.describe().T`

Out[]:

	count	mean	std	min	25%	50%	75%	max
Hours	25.0	5.012	2.525094	1.1	2.7	4.8	7.4	9.2
Scores	25.0	51.480	25.286887	17.0	30.0	47.0	75.0	95.0

Createing the Scatter plot

```
In [ ]: # Scatter plot of study hours vs. scores
plt.scatter(df['Hours'], df['Scores'])
plt.title('Study Hours vs. Scores')
plt.xlabel('Hours')
plt.ylabel('Scores')
plt.show()
```



Data Preparation

Split the data into training and testing sets:

```
In [ ]: X = df['Hours'].values.reshape(-1, 1) # Independent variable (study hours)
y = df['Scores'].values # Dependent variable (scores)
```

```
In [ ]: X
```

```
Out[ ]: array([[2.5],
               [5.1],
               [3.2],
               [8.5],
               [3.5],
               [1.5],
               [9.2],
               [5.5],
               [8.3],
               [2.7],
               [7.7],
               [5.9],
               [4.5],
               [3.3],
               [1.1],
               [8.9],
               [2.5],
               [1.9],
               [6.1],
               [7.4],
               [2.7],
               [4.8],
               [3.8],
               [6.9],
               [7.8]])
```

```
In [ ]: y
```

```
Out[ ]: array([21, 47, 27, 75, 30, 20, 88, 60, 81, 25, 85, 62, 41, 42, 17, 95, 30,
               24, 67, 69, 30, 54, 35, 76, 86], dtype=int64)
```

```
In [ ]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_
```

```
In [ ]: X_train
```

```
Out[ ]: array([[2.7],
               [3.3],
               [5.1],
               [3.8],
               [1.5],
               [3.2],
               [4.5],
               [8.9],
               [8.5],
               [3.5],
               [2.7],
               [1.9],
               [4.8],
               [6.1],
               [7.8],
               [5.5],
               [7.7],
               [1.1],
               [7.4],
               [9.2]])
```

```
In [ ]: X_test
```

```
Out[ ]: array([[8.3],
               [2.5],
               [2.5],
               [6.9],
               [5.9]])
```

```
In [ ]: y_train
```

```
Out[ ]: array([25, 42, 47, 35, 20, 27, 41, 95, 75, 30, 30, 24, 54, 67, 86, 60, 85,
              17, 69, 88], dtype=int64)
```

```
In [ ]: y_test
```

```
Out[ ]: array([81, 30, 21, 76, 62], dtype=int64)
```

Train the Linear Regression Model

Create and train a simple linear regression model:

```
In [ ]: # Create a Linear regression model
model = LinearRegression()
```

```
In [ ]: # Train the model on the training data
model.fit(X_train, y_train)
```

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

Make Predictions

Use the trained model to make predictions on the test data and also predict the score for a student who studies for 9.25 hours:

```
In [ ]: # Predict scores on the test data
y_pred = model.predict(X_test)
y_pred
```

```
Out[ ]: array([83.18814104, 27.03208774, 27.03208774, 69.63323162, 59.95115347])
```

```
In [ ]: # Predict the score for a student who studies for 9.25 hours
hours_to_predict = np.array([[9.25]])
predicted_score = model.predict(hours_to_predict)
predicted_score
```

```
Out[ ]: array([92.38611528])
```

```
In [ ]: print("Predicted Score for 9.25 hours of study:", predicted_score[0])
```

Predicted Score for 9.25 hours of study: 92.38611528261494

Evaluate the Model

You should evaluate the model's performance using appropriate metrics (e.g., Mean Absolute Error, Mean Squared Error, R-squared):

```
In [ ]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score

# Calculate evaluation metrics
mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
```

```
In [ ]: print("Mean Absolute Error:", mae)
        print("Mean Squared Error:", mse)
        print("R-squared:", r2)
```

Mean Absolute Error: 3.9207511902099244

Mean Squared Error: 18.943211722315272

R-squared: 0.9678055545167994