## **Linear Regression - Student Percentage Prediction**

```
# Libraries
In [143...
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
In [144... df = pd.read csv('student scores.csv')
          df.head()
Out[144...
             Hours Scores
           0
                2.5
                        21
                5.1
                        47
                3.2
                        27
                8.5
                        75
                3.5
                        30
```

## **Data Splitting**

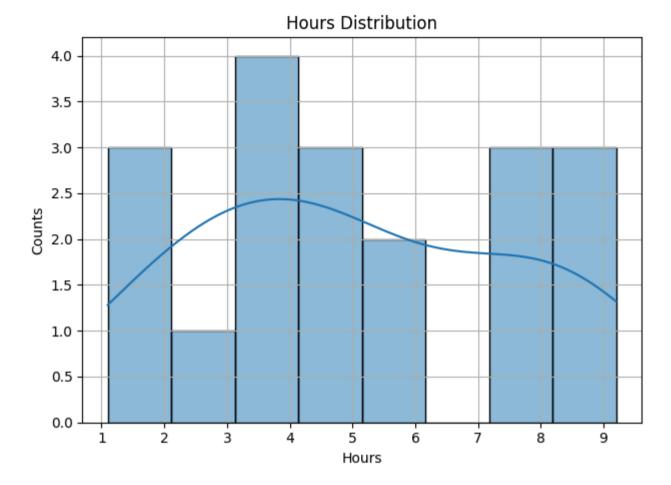
```
In [145... df.shape
Out[145... (25, 2)
In [146... X,y = df['Hours'],df['Scores']
In [147... from sklearn.model_selection import train_test_split
In [148... X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

## **Exploratory Data Analysis**

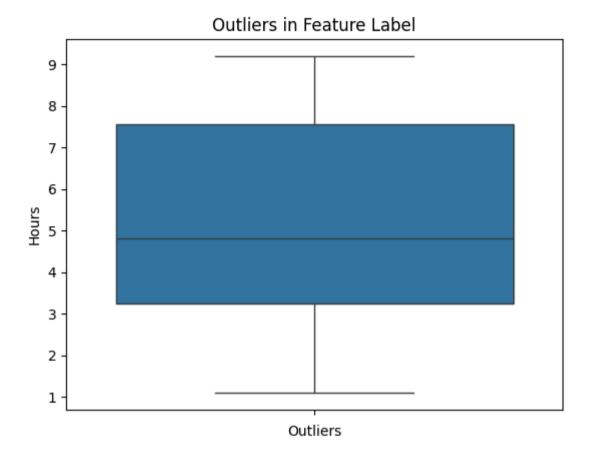
```
In [149...
          X train.shape
Out[149...
          (20,)
          X train.dtypes
In [150...
Out[150... dtype('float64')
          X_train.info()
In [151...
         <class 'pandas.core.series.Series'>
         Index: 20 entries, 9 to 6
         Series name: Hours
         Non-Null Count Dtype
         20 non-null
                         float64
         dtypes: float64(1)
         memory usage: 320.0 bytes
In [152...
          X_train.describe()
Out[152...
                    20.000000
           count
           mean
                     4.960000
           std
                     2.566874
           min
                     1.100000
           25%
                     3.075000
           50%
                     4.650000
           75%
                     7.475000
                     9.200000
           max
           Name: Hours, dtype: float64
In [153...
          X_train.isnull().sum()
Out[153... 0
          X_train[X_train.duplicated()]
In [154...
```

```
Out[154...
                2.7
          20
           Name: Hours, dtype: float64
          y train.drop(labels=[20],inplace=True)
In [159...
In [160... X train.drop duplicates(inplace=True)
In [161...
         X train.shape
Out[161... (19,)
          lower,upper = np.percentile(X train, 25),np.percentile(X train, 75)
In [162...
          iqr = upper-lower
          lb,ub = iqr-1.5*lower,iqr+1.5*upper
In [163...
          lower, upper, iqr
Out[163... (3.25, 7.55000000000001, 4.30000000000001)
In [164...
          1b,ub
Out[164... (-0.57499999999999, 15.625000000000000)
         X_train[X_train<lb],X_train[X_train>ub]
In [165...
Out[165... (Series([], Name: Hours, dtype: float64),
            Series([], Name: Hours, dtype: float64))
            • No Outliers Found
In [166...
          X train.head()
```

```
Out[166... 9
               2.7
         13
             3.3
         1
               5.1
          22 3.8
               1.5
          5
          Name: Hours, dtype: float64
In [167... sns.histplot(X_train,bins=8,kde=True)
         plt.xlabel('Hours')
         plt.ylabel('Counts')
         plt.title('Hours Distribution')
         plt.tight_layout()
         plt.grid()
         plt.show()
```



```
In [168... sns.boxplot(data=X_train)
    plt.xlabel('Outliers')
    plt.ylabel('Hours')
    plt.title('Outliers in Feature Label')
    plt.show()
```



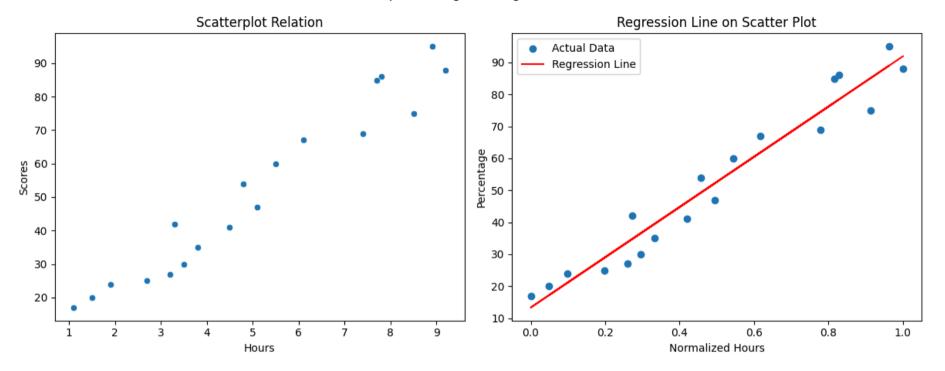
No Outliers

# **Building Model**

```
In [170... # Normalization
In [174... X_train = pd.DataFrame(X_train)
In [183... X_test = pd.DataFrame(X_test)
```

```
In [184... | from sklearn.preprocessing import MinMaxScaler
          scaler = MinMaxScaler()
          X train normalized = scaler.fit transform(X train)
          X test normalized = scaler.transform(X test)
         from sklearn.linear model import LinearRegression
In [185...
          model = LinearRegression()
          model.fit(X train normalized, y train)
          print("Coefficients:", model.coef )
          print("Intercept:", model.intercept )
         Coefficients: [78.59073018]
         Intercept: 13.341395700266219
In [210... fig, axes = plt.subplots(1, 2, figsize=(12, 5))
          # 1st Scatter plot
          sns.scatterplot(x=X train['Hours'], y=y train, ax=axes[0])
          axes[0].set xlabel('Hours')
          axes[0].set ylabel('Scores')
          axes[0].set title('Scatterplot Relation')
          # 2nd Scatter plot with Regression Line
          axes[1].scatter(X train normalized[:, 0], y train, label='Actual Data')
          axes[1].plot(X train normalized[:, 0], model.predict(X train normalized), color='red', label='Regression Line')
          # Add labels and title
          axes[1].set_xlabel('Normalized Hours')
          axes[1].set_ylabel('Percentage')
          axes[1].set_title('Regression Line on Scatter Plot')
          axes[1].legend()
          plt.suptitle('Scatterplots Along with Regression Line')
          plt.tight_layout()
          plt.show()
```

#### Scatterplots Along with Regression Line



### **Evaluation**

```
In [211.... from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
    y_pred = model.predict(X_test_normalized)
    mae = mean_absolute_error(y_test, y_pred)
    mse = mean_squared_error(y_test, y_pred)
    rmse = np.sqrt(mse)
    r2 = r2_score(y_test, y_pred)

print("Mean Absolute Error (MAE):", mae)
    print("Mean Squared Error (MSE):", mse)
    print("Root Mean Squared Error (RMSE):", rmse)
    print("R-squared (R2):", r2)
```

```
Mean Absolute Error (MAE): 3.933980548404949
Mean Squared Error (MSE): 18.901095098644035
Root Mean Squared Error (RMSE): 4.347538970342191
R-squared (R2): 0.9678771327351393
```

#### **Prediction**

```
import warnings
warnings.filterwarnings("ignore")
hours_to_predict = float(input('Enter Hours to predict Percentage..'))
normalized_hours_to_predict = scaler.transform([[hours_to_predict]])

predicted_percentage = model.predict(normalized_hours_to_predict)

print(f"Predicted Percentage for {hours_to_predict} hours / day of study is : {round(predicted_percentage[0],2)}")
```

Predicted Percentage for 5.0 hours / day of study is : 51.18