```
!pip install pandas numpy scikit-learn matplotlib seaborn
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# PRN: 22070521021
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
from sklearn.metrics import mean absolute error, mean squared error,
r2 score
import seaborn as sns
df = pd.read csv('./student scores.csv')
df.columns = ['Hours Studied', 'Test Score']
print("Dataset Loaded Successfully.")
print("\nFirst 5 rows of the data:")
print(df.head())
print("\nData Info:")
print(df.info())
Dataset Loaded Successfully.
First 5 rows of the data:
   Hours Studied Test Score
0
             2.5
                          21
1
             5.1
                          47
2
             3.2
                           27
3
             8.5
                          75
4
             3.5
                          30
Data Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 25 entries, 0 to 24
Data columns (total 2 columns):
#
     Column
                    Non-Null Count
                                     Dtype
0
     Hours Studied 25 non-null
                                     float64
1
     Test Score
                    25 non-null
                                     int64
dtypes: f\overline{loat}64(1), int64(1)
memory usage: 532.0 bytes
None
plt.style.use('seaborn-v0 8-whitegrid')
plt.figure(figsize=(8, 5))
plt.scatter(df['Hours_Studied'], df['Test_Score'], color='darkblue')
plt.title('Study Hours vs. Test Score (EDA) ', fontsize=14)
plt.xlabel('Hours Studied')
plt.ylabel('Test Score')
```

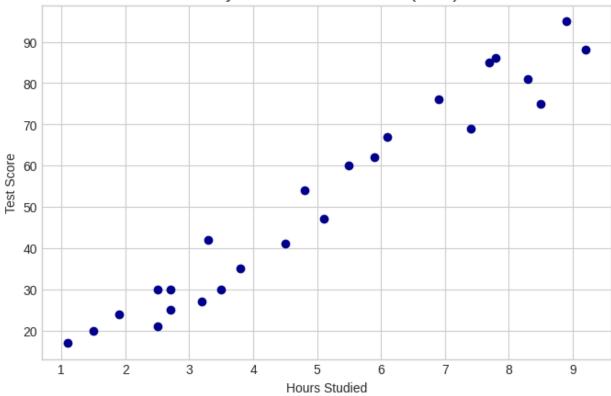
```
plt.show()

X = df[['Hours_Studied']]
y = df['Test_Score']

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)

print(f"Total Samples: {len(df)}")
print(f"Training Samples: {len(X_train)}")
print(f"Testing Samples: {len(X_test)}")
```

Study Hours vs. Test Score (EDA)

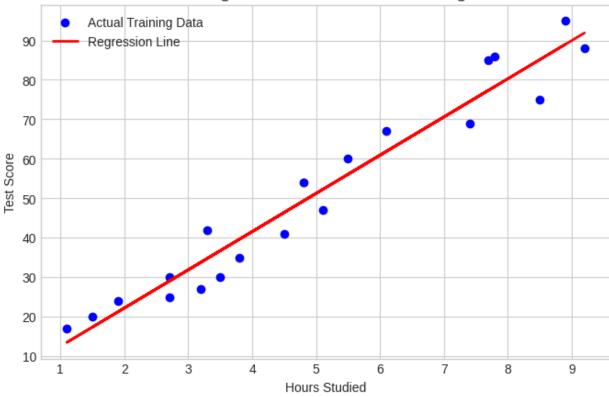


```
Total Samples: 25
Training Samples: 20
Testing Samples: 5

model = LinearRegression()
model.fit(X_train, y_train)
coefficient = model.coef_[0]
intercept = model.intercept_
print("--- Linear Regression Model Parameters ---")
print(f"Coefficient (m, Slope): {coefficient:.2f}")
print(f"Intercept (c): {intercept:.2f}")
```

```
print("\nRegression Line Equation (y = mx + c):")
print(f"Test Score = {coefficient:.2f} * Hours Studied +
{intercept:.2f}")
--- Linear Regression Model Parameters ---
Coefficient (m, Slope): 9.68
Intercept (c): 2.83
Regression Line Equation (y = mx + c):
Test_Score = 9.68 * Hours_Studied + 2.83
plt.figure(figsize=(8, 5))
plt.scatter(X_train, y_train, color='blue', label='Actual Training
Data')
# Plot the predicted line (y predicted = m*X train + c)
plt.plot(X train, model.predict(X train), color='red', linewidth=2,
label='Regression Line')
plt.title('Linear Regression: Model Fit on Training Data',
fontsize=14)
plt.xlabel('Hours Studied')
plt.ylabel('Test Score')
plt.legend()
plt.show()
```

Linear Regression: Model Fit on Training Data



```
y_pred = model.predict(X_test)
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("--- Model Performance Metrics on Test Set ---")
print(f"Mean Absolute Error (MAE): {mae:.2f}")
print(f"Mean Squared Error (MSE): {mse:.2f}")
print(f"Root Mean Squared Error (RMSE): {rmse:.2f}")
print(f"R-squared Score (R2): {r2:.4f}")
residuals = y_test - y_pred
plt.figure(figsize=(3, 3))
plt.scatter(y_pred, residuals, color='purple')
plt.hlines(y=0, xmin=y_pred.min(), xmax=y_pred.max(), colors='red',
linestyles='--', label='Zero Residual Line')
plt.title('Residual Plot', fontsize=14)
plt.xlabel('Predicted Test Score')
plt.ylabel('Residuals (Actual - Predicted)')
plt.legend()
plt.show()
```

--- Model Performance Metrics on Test Set ---

Mean Absolute Error (MAE): 3.92 Mean Squared Error (MSE): 18.94 Root Mean Squared Error (RMSE): 4.35

R-squared Score (R2): 0.9678

