Predicting Student Scores with Machine Learning

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Introduction

Why Predicting Student Scores Matters

In today's data-driven world, educational performance analytics has become a powerful tool to understand and support student success.
One of the key questions in this area is:

Can we predict a student's score based on their study hours?
This project explores that question using machine learning regression techniques.

Why This Topic?

- Helps identify learning patterns and optimize study plans
- Demonstrates real-world application of supervised learning
- A simple yet meaningful case to strengthen my data science foundation

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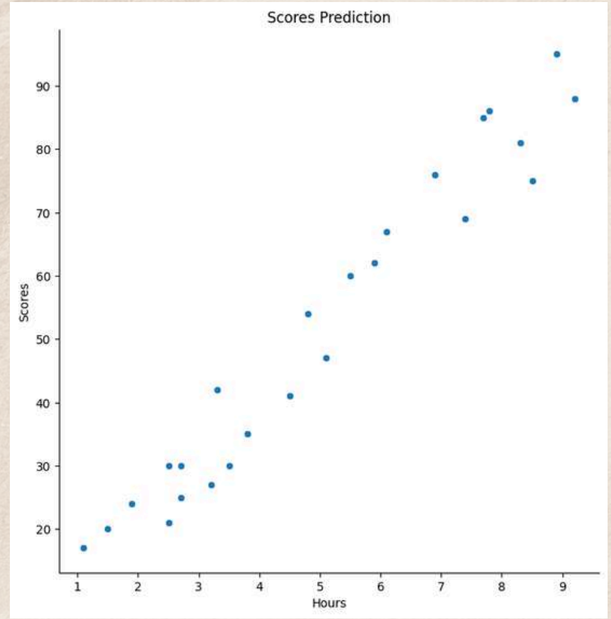
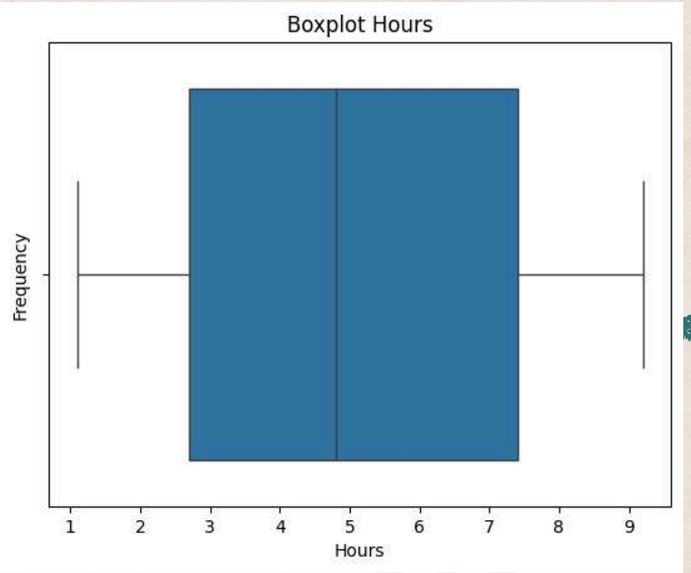
Exploratory Data Analysis (EDA)

- Dataset: Contains 25 data points showing the number of study hours vs corresponding scores.
- Checked for missing values → Not found
- Checked for duplicates → All unique records
- Outlier detection using Boxplot → No extreme outliers
- Relationship between Hours and Scores visualized using scatter plot → Reveals a strong positive linear correlation





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Splitting Data

```
# Import machine learning data from scikit learn
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test= train_test_split(X,y,train_size=0.75,random_state=1)
```

To evaluate model performance objectively, I split the dataset into training and testing sets using an 75-25 split.

- Training Set: 75% of the data
- Testing Set: 25% of the data
- Applied standard data preprocessing techniques before modeling.

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Regression Models Used

```
lr_model= LinearRegression() # I
lr_model.fit(X_train, y_train) #
```

```
LinearRegression
LinearRegression()
```

1.Linear Regression

```
dt_model= DecisionTreeRegressor()
dt_model.fit(X_train, y_train) # I
```

```
DecisionTreeRegressor
DecisionTreeRegressor()
```

2. Decision Tree Regressor

```
[ ]  rf_model= RandomForestRegressor()
     rf_model.fit(X_train, y_train) # I
```

```
RandomForestRegressor
RandomForestRegressor()
```

3. Random Forest Regressor

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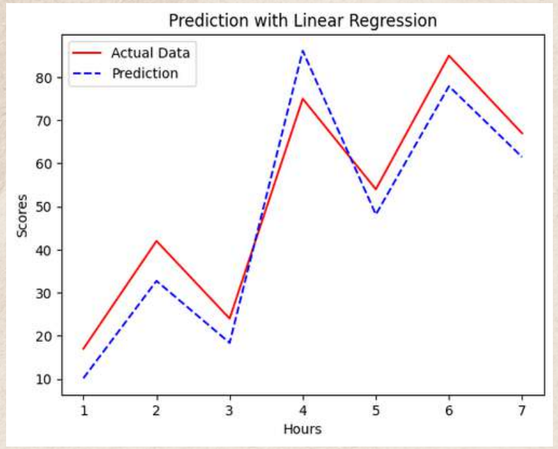
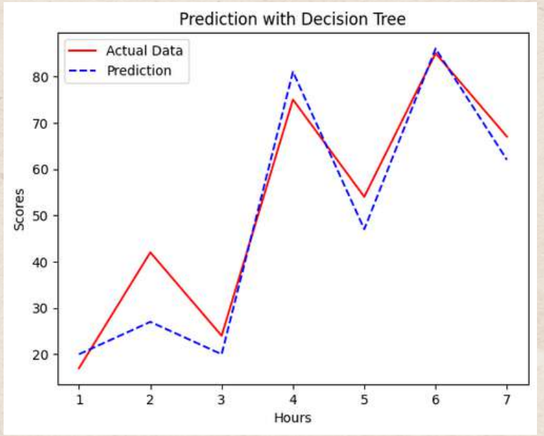
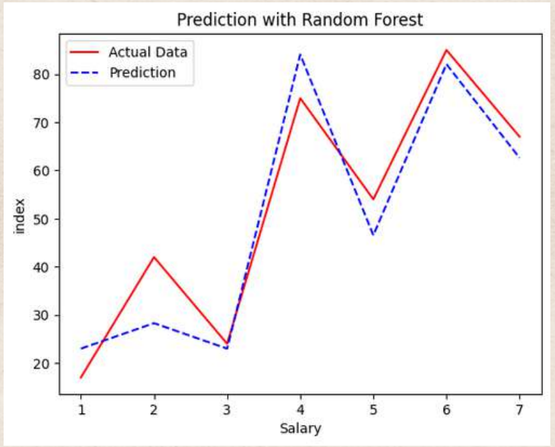
Dataset Visualization

Three regression models were implemented to predict Scores:

1. Linear Regression
   - Simple and interpretable baseline model
2. Decision Tree Regressor
   - Captures non-linear patterns and feature interactions
3. Random Forest Regressor
   - Ensemble method to improve accuracy and reduce overfitting







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Model Evaluation

To compare model performance, I used Mean Absolute Error (MAE) and R^2 Score.

Model	MAE	R^2 Score
Linear Regression	5.52	0.94
Decision Tree Regressor	4.85	0.96
Random Forest Regressor	4.3	0.97

Best Performing Model: Random Forest Regressor

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Conclusion & Takeaways

- There is a strong linear relationship between Hours Studied and Scores.
- Among all models tested, Random Forest Regressor achieved the best performance.
- This project strengthened my skills in EDA, regression modeling, and data visualization using Python.
- I'm excited to apply these skills in real-world data projects during my upcoming data science internship.

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Certificate of Completion

I am proud to have successfully completed Data Series Fair 18.0 – Data & AI Intensive Bootcamp, hosted by dibimbing.id, held on March 4–7, 2025.

This 4-day program deepened my understanding of Machine Learning, Data Science, and AI through hands-on classes and practical portfolio building.

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