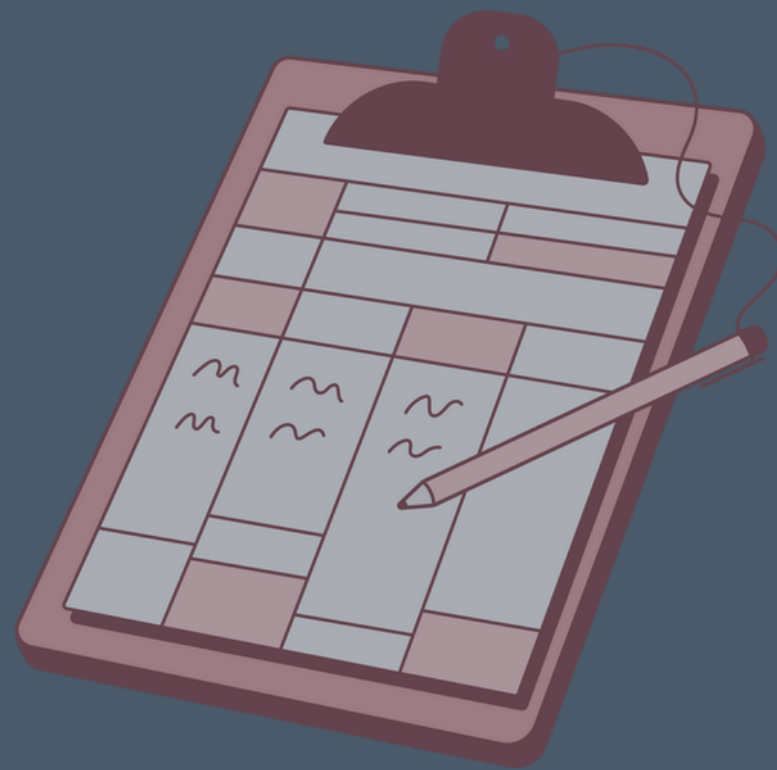


Student Score Prediction

Presented by [Rana Awad Abdel Aziz Saad]



Agenda



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Project Overview

Goal: Predict student scores based on study hours.

Tools Used: Python, Pandas, NumPy, Matplotlib, Scikit-learn

Algorithm: Linear Regression

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

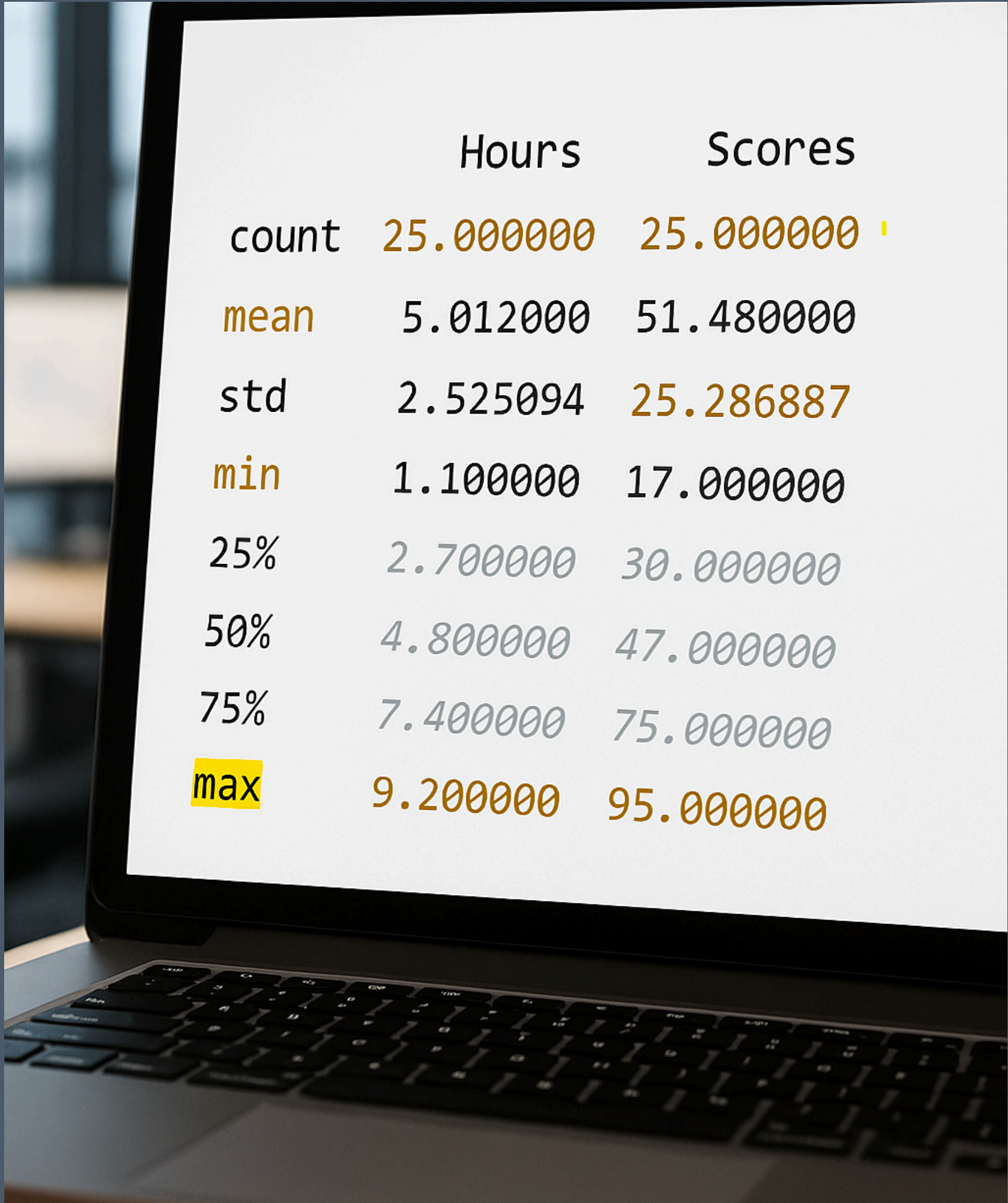
```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
from sklearn.metrics import mean_squared_error, r2_score
```

Overview of the dataset used

The dataset for the Student Score Prediction Project consists of **25 records** detailing the relationship between study hours and exam scores. Each record contains two columns: one representing **study hours** and the other representing **exam scores**. Importantly, the dataset has no missing values, ensuring a complete analysis.

The mean study hours recorded is approximately **5 hours**, while the mean exam score is around **51 points**. This average indicates a moderate level of study time correlated with exam performance. The dataset provides a solid foundation for applying linear regression techniques, allowing us to explore how variations in study hours can predict student performance on exams. The next step involves preprocessing this data to visualize its characteristics effectively.



	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

Data Processing

- We checked for missing values – none were found

```
data.isnull().sum()
```

- we split the data into:
- Training Set (80%) to train the model
- Testing Set (20%) to evaluate its accuracy

```
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
```

Model Training

$$\text{Predicted Score} = 9.78 \times \text{Hours} + 2.48$$

We used the Linear Regression model which

- learns the relationship between hours (input) and scores (output).
- finds a straight line ($y = a \cdot x + b$) that best fits the data.

Here,

x = Hours

y^{\wedge} = predicted score

a = slope (coef)

b = intercept

```
model = LinearRegression()  
model.fit(X_train, y_train)
```

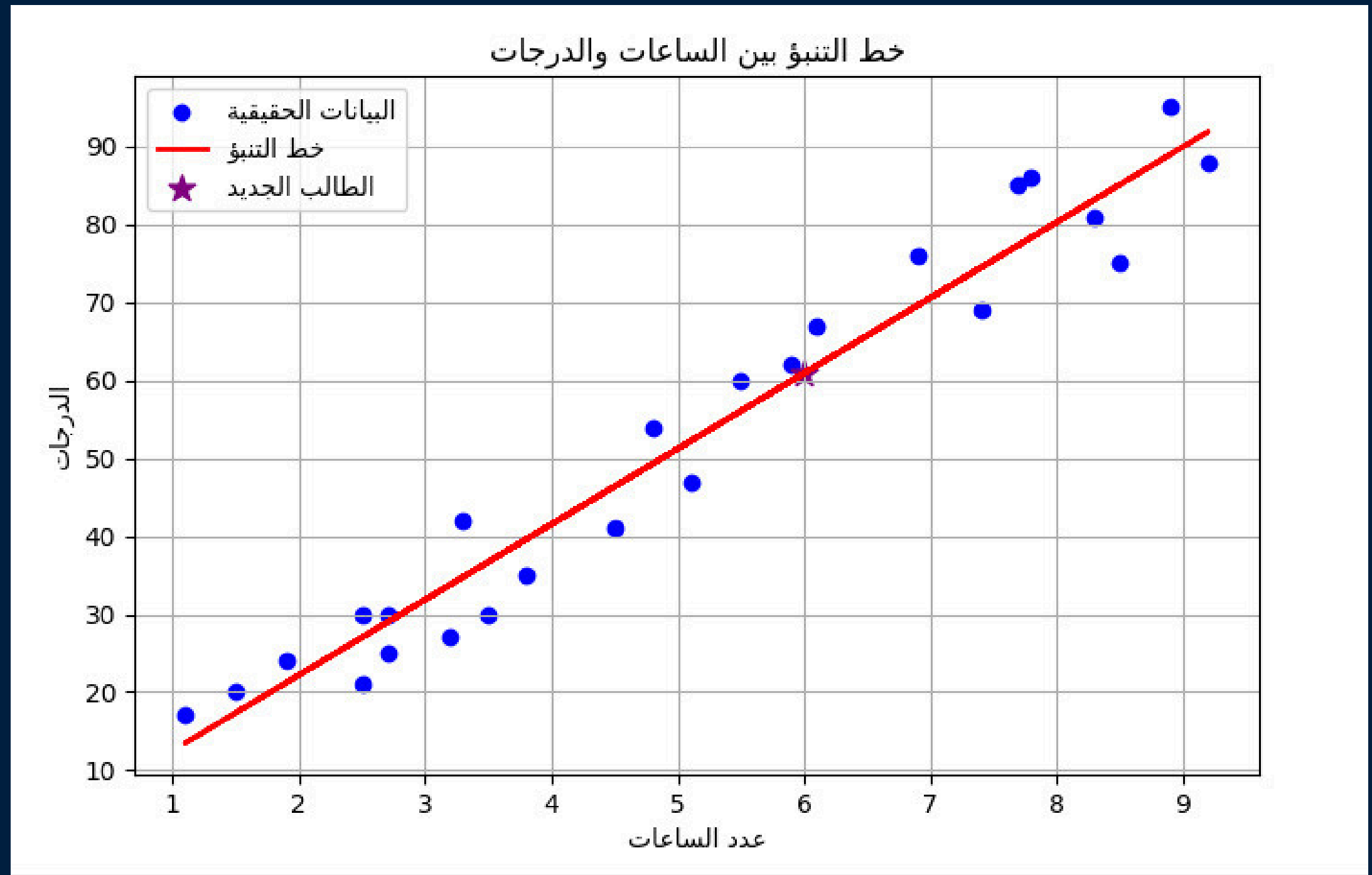
```
print("coef =", model.coef_[0])  
print("intercept =", model.intercept_)
```

Data Visualization of Scores

Hours vs. Scores Scatter Plot

This plot shows a **strong positive correlation**.

```
plt.scatter(data['Hours'], data['Scores'])
```



Blue = Real data

Red = Prediction line

Purple = New student entered manually

Model Evaluation

- We used two metrics:
- MSE (Mean Squared Error) = 18.94 →
- R^2 (R-squared) = 0.967 → means the model explains 96.7% of the relationship between hours and scores.



18.94

Mean Squared Error (MSE)
measures average prediction error
lower is the better

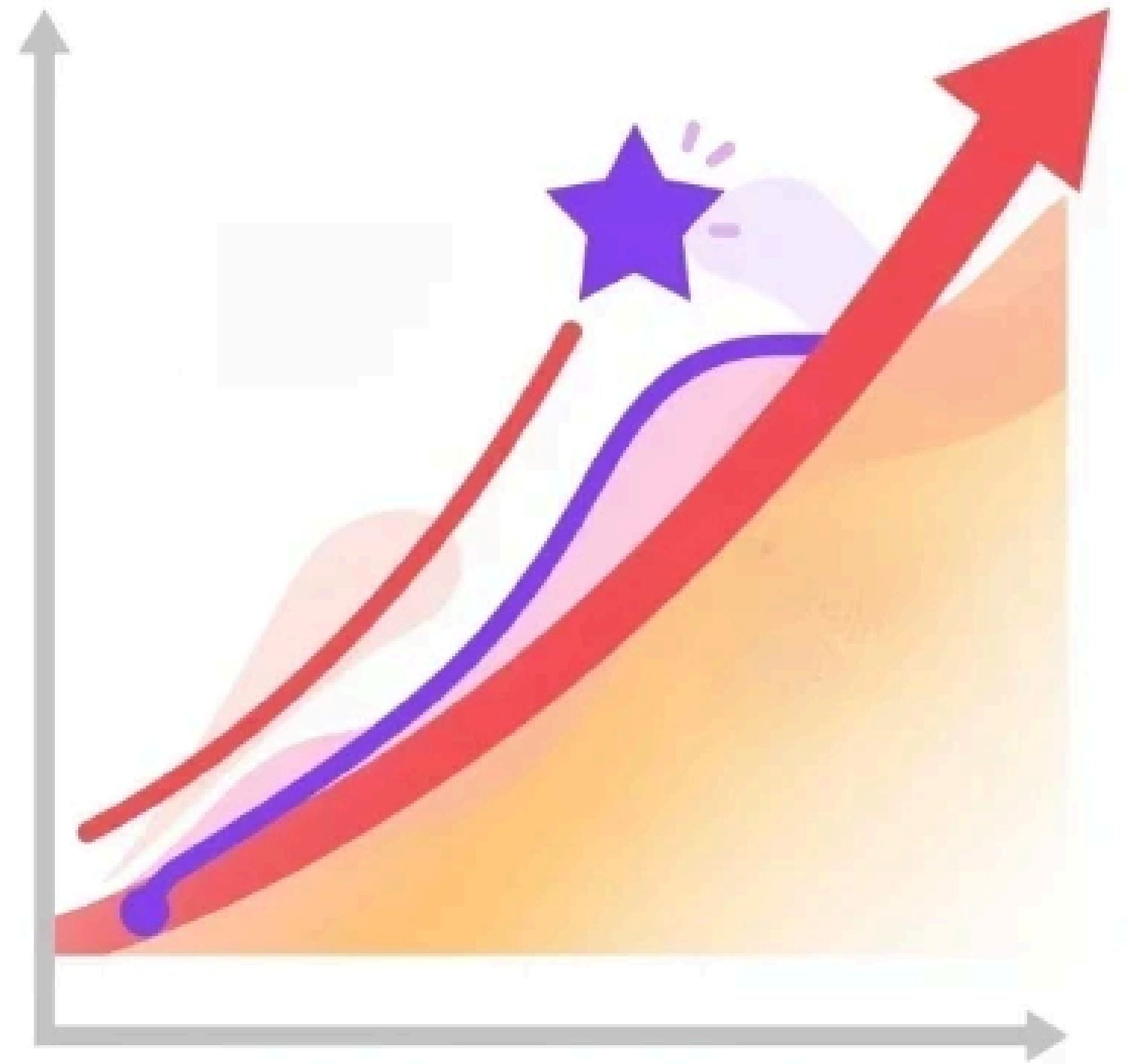


0.96

R-squared Value
the model explains 96.7% of the
relationship between hours and
scores. Closer than 1 is the better

Prediction

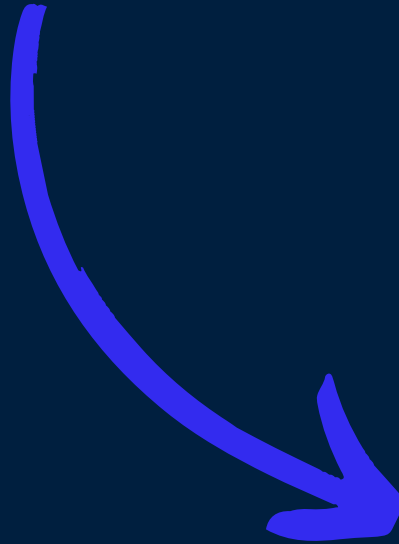
- Input = 6 study hours
“manually”
- Predicted score = $9.78 \times 6 + 2.48 \approx 60.92$



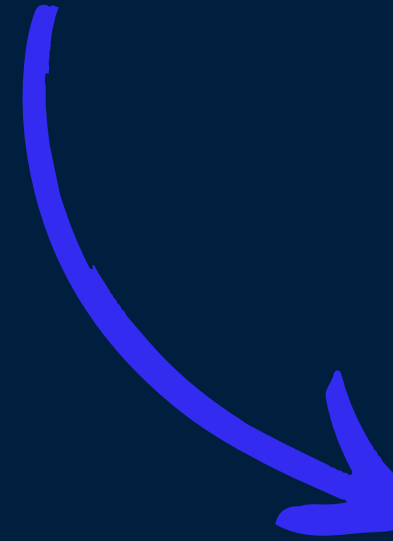
Challenges

- Small dataset (only 25 records).
- Simple model (only one feature).
- Real-life data may include other factors like attendance, sleep, or stress.

Conclusion



This project demonstrates the power of
Machine Learning in education



It provides valuable insights into how
study time affects student
performance.
More study hours → Higher score

Thank You for Your Attention

Email

ranaawad20199@gmail.com

Github