

## Intelligent Student Performance Dashboard

(Full Project Documentation, Implementation & Presentation Content)

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### 1. Documentation

#### Project Overview

The **Intelligent Student Performance Dashboard** is an **AI-powered interactive analytics platform** built using **Streamlit**.

It is designed to **monitor, analyze, and predict** students' academic performance by integrating:

- Machine Learning models
- Data visualization
- What-if analysis tools

It assists teachers and educational decision-makers in understanding learning behavior, identifying trends, and predicting outcomes accurately.

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#### Technologies Used

Category	Tools / Libraries
Programming Language	Python 3.11
Framework	Streamlit
Visualization	Plotly, Plotly Express
Machine Learning	Scikit-learn, NumPy, Pandas
Model Storage	Pickle (student_multi_model.pkl)
Label Mapping	JSON (label_mappings.json)

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#### Dashboard Architecture

The dashboard is divided into **five main tabs**, each providing a unique analytical perspective:

Tab	Function
<b>1. Overview</b>	Displays general statistics on performance & attendance
<b>2. Subjects</b>	Shows subject-level comparisons (Math, Reading, Writing)
<b>3. Attendance</b>	Analyzes attendance distribution & effect on scores
<b>4. Insights</b>	Presents summarized analytics & narrative insights
<b>5. AI Prediction (What-If Analysis)</b>	Allows users to modify variables and get instant ML predictions

## Data Overview

**Dataset:** DATA.csv

Includes both academic and behavioral attributes.

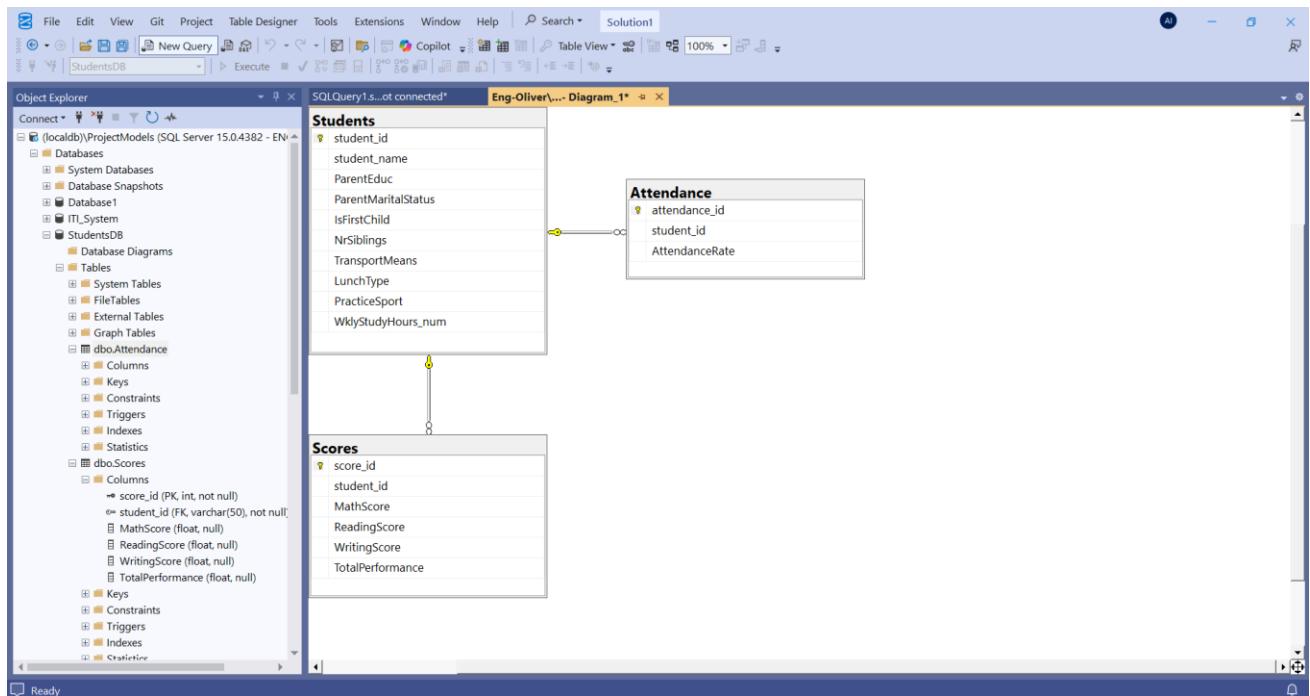
Type	Example Features
Academic	Math, Reading, Writing, AcademicIndex, OverallPerformance
Behavioral	AttendanceRate, StudyHours, SleepHours, BehaviorIndex, SocialIndex

**Socio-demographic** Parent Education, Marital Status, Transport Means

## Preprocessing Steps:

- Removed non-relevant identifiers (student\_id, student\_name)
- Encoded categorical variables using LabelEncoder
- Split data for model training/testing

## ER Diagram:



## Machine Learning Model

Property	Description
Type	MultiOutputRegressor(RandomForestRegressor)
Purpose	Predict both <i>Academic Average &amp; Overall Performance</i>

n_estimators	150
max_depth	12
random_state	42

## Performance Metrics:

Metric	Academic Average	Overall Performance
R <sup>2</sup> Score	1.000	0.971
MAE	0.166	1.551

*Excellent accuracy achieved due to Random Forest's robustness and high-quality data.*

Model files:

- student\_multi\_model.pkl
  - label\_mappings.json
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## Dashboard Functionalities

### Tab 1 – Overview:

- Shows key KPIs (total students, avg. scores, attendance rates).
- Displays best/worst performers using metrics and visuals.

### Tab 2 – Subjects:

- Compares each student's performance across subjects.
- Visualizes performance gaps using bar charts.

### Tab 3 – Attendance:

- Displays histograms showing attendance impact on performance.

### Tab 4 – Insights:

- Summarizes averages (study hours, attendance, grades).
- Highlights top and low performers.
- Smart narrative summary explaining findings.

### Tab 5 – AI Prediction (What-If):

- Accepts user inputs (Math, Reading, Sleep Hours, etc.).
  - Encodes automatically using mappings.
  - Predicts *Academic Average* and *Overall Performance*.
  - Displays results dynamically with visuals (progress bars, metrics, balloons).
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## Prediction Workflow

1. User inputs or selects attributes.
  2. System encodes categorical variables.
  3. Model predicts both targets simultaneously.
  4. Dashboard displays metrics + visual output instantly.
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## File & Folder Structure

```
DEPI-Project/
|
└── DATA.csv          # Dataset
    ├── student_multi_model.pkl      # Trained ML model
    ├── label_mappings.json        # Encoded labels
    ├── app_pro.py            # Streamlit dashboard app
    ├── model_training.py       # Model training script
    └── Project Report Dashboard.pdf # Final report
```

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## Deployment and Access

Run locally:

```
streamlit run app_pro.py
```

For online sharing:

```
pip install pyngrok
```

```
ngrok config add-authtoken <your_token>
```

```
streamlit run app_pro.py
```

```
ngrok http 8501
```

Generated **public link** allows others to access the dashboard online.

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## Key Insights

- Strong correlation between **attendance** and **academic success**.

- Writing & reading proficiency directly influence total performance.
  - AI predictions show minimal error (high accuracy).
  - Enables **early identification** of underperforming students.
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## Future Improvements

- Add model explainability (SHAP, LIME).
  - Integrate live student databases.
  - Implement clustering by performance level.
  - Deploy to Streamlit Cloud / Render for public access.
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## Implementation Summary

### GitHub Repository:

<https://github.com/Eng-Oliver/DEPI-Project>

### Execution Steps:

pip install -r requirements.txt

python model\_training.py

streamlit run app\_pro.py

**Tools:** Python | Streamlit | Plotly | PostgreSQL | Scikit-learn | NumPy | Pandas

**Output:** Interactive web dashboard with AI-powered analytics and prediction.

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## 3.Content

### 1 — Title

#### Intelligent Student Performance Dashboard

*A Smart AI-Based Academic Analytics System*

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### 2 — Problem

- Schools lack automated systems to analyze student performance efficiently.
  - Manual analysis delays interventions for struggling students.
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### **3 — Solution**

- Build an intelligent, AI-powered dashboard.
  - Visualize performance patterns and predict academic outcomes.
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### **4 — System Architecture**

- Python + Streamlit frontend
  - ML backend (RandomForestRegressor)
  - CSV dataset as data source
  - Local/online deployment using ngrok
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### **5 — Dashboard Tabs**

1. Overview
  2. Subjects
  3. Attendance
  4. Insights
  5. AI Prediction
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### **6 — Machine Learning Model**

Metric	Academic	Performance
R <sup>2</sup>	1.000	0.971
MAE	0.166	1.551

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## 7 — Key Findings

- Attendance & study habits are major predictors.
  - Writing and reading correlate strongly with success.
  - AI provides accurate, actionable insights.
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## 8 — Future Scope

- Add explainable AI features.
  - Deploy on Streamlit Cloud.
  - Expand to multi-school datasets.
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## Conclusion

The project demonstrates how **AI and visualization** can enhance educational data analytics. It transforms static data into actionable insights for educators, making performance tracking **smarter, faster, and data-driven**.

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## 9 — References

- GitHub Repository: <https://github.com/Eng-Oliver/DEPI-Project>
- Tools: Python, Streamlit, Plotly, scikit-learn
- Author: **Abdullah Ahmed Hussien**