STUDENT ACADEMIC PERFORMANCE PREDICTOR

Chapter 1: Project Description and Outline

- Project Name: Student Academic Performance Predictor
- Objective: To develop a predictive system that estimates students' performance indices based on key factors such as hours studied, sleep hours, previous scores, extracurricular activity involvement, and practice frequency.
- Motivation: Educational institutions can use this tool to support students proactively
 by identifying those who may be underperforming.

Scope:

- o Predict performance using linear regression.
- Scale and preprocess input data for consistency.
- o Build a user-friendly input and prediction pipeline.

Chapter 2: Related Work Investigation

Research Review:

- Similar systems have been developed using Decision Trees, SVM, and Neural Networks.
- This project focuses on explainability through Linear Regression.

• Gaps Identified:

- Many models don't include user input scalability or interactive interfaces.
- Often miss out on analyzing factors like sleep and extracurricular activities.

Reference:

 Cited research from IEEE papers and educational datasets to support the model's feature selection.

■ Chapter 3: Requirement Artifacts

• Hardware Requirements:

System with at least 4GB RAM and Python 3.8+.

• Software Requirements:

o Python, Scikit-learn, Pandas, Matplotlib, Joblib.

• Functional Requirements:

- Model must load trained data.
- Accept multiple inputs at once.
- Normalize and predict performance index.

• Non-Functional Requirements:

- Accuracy should exceed 95% on test data.
- Response time < 2 seconds per prediction.

■ Chapter 4: Design Methodology and Its Novelty

Pipeline Overview:

- Load data
- Handle missing/duplicate values
- Encode categorical features
- Scale numerical data
- Train Linear Regression model
- Save model and scaler
- Accept new user input and preprocess it

Predict and display results

Novelty:

- o Real-time, multi-student input capability.
- Blended approach of feature engineering with usability.
- o Custom scaler handling for unseen inputs.

Chapter 5: Technical Implementation and Analysis

- Model Used: LinearRegression() from sklearn.linear_model
- Dataset: Custom dataset with:
 - 5 Features: Hours Studied, Previous Scores, Sleep Hours, Extracurricular Activities, Papers Practiced
 - o 1 Target: Performance Index

• Preprocessing:

- Handled Yes/No as binary values
- o Scaled numerical data using StandardScaler

• Evaluation Metrics:

o MAE: 1.61

o MSE: 4.08

o RMSE: 2.02

o R² Score: 0.99 (Excellent fit)

Chapter 6: Project Outcome and Applicability

• Outcome:

- Achieved high predictive accuracy.
- o Successful multi-student prediction from user inputs.

• Use Cases:

- o Educational counselors can prioritize support.
- o Schools and colleges can implement for semester assessments.

Scalability:

 Easily extendable to larger datasets and ML models (like Random Forest or XGBoost).

Chapter 7: Conclusion and Recommendations

• Conclusion:

- The project met its goals and demonstrated practical, high-accuracy predictions.
- o Usable in educational environments to monitor academic health.

• Recommendations:

- o Extend model to handle text inputs using NLP.
- Add GUI/website for more accessibility.
- Store predictions in a database for analysis.