



**KIET**  
**GROUP OF INSTITUTIONS**  
*Connecting Life with Learning*

**Assesment Report**  
on  
**“Problem Statement”**  
**submitted as partial fulfillment for the**  
**award of**  
**BACHELOR OF TECHNOLOGY**  
**DEGREE**  
**SESSION 2024-25**  
In  
**Harsh Dubey**  
By  
**Name (18)**

**Under the supervision of**

**“Mr. Abhishek Shukla Sir”**

**KIET Group of Institutions, Ghaziabad**

**Affiliated to**

**Dr. A.P.J. Abdul Kalam Technical**

**University, Lucknow**

**(Formerly UPTU)**

**May, 2025**

---

 **Student Performance Prediction using Machine Learning**

---

 **Introduction**

The aim of this project is to predict whether a student will **pass or fail** (or which grade/class they will be in) using **machine learning algorithms**. The dataset includes student-related features such as attendance, study habits, previous scores, and demographic details.

By analyzing this data, we can **identify students at risk** early and offer them support.

---

 **Methodology**

**1. Dataset Upload**

- Dataset used: 8. Student Performance Prediction.csv
- Source: UCI Machine Learning Repository

**2. Data Preprocessing**

- Handled missing values (if any)
- Encoded categorical data using LabelEncoder
- Normalized dataset to make it model-ready

### 3. Feature & Target

- Features: All columns related to student behavior and academics
- Target: GradeClass (Can be changed to Result if applicable)

### 4. Train-Test Split

```
train_test_split(X, y, test_size=0.2, random_state=42)
```

### 5. Model Used

- **Random Forest Classifier**
- Reason: Works well with mixed data and avoids overfitting

### 6. Evaluation Metrics

Used:

- Accuracy
- Precision
- Recall
- Classification Report
- Graph: Actual vs Predicted values

---

#### Code

```
from google.colab import files  
uploaded = files.upload()  
import pandas as pd  
  
filename = list(uploaded.keys())[0]  
df = pd.read_csv(filename)  
  
print(" ◆ First 5 rows of data:")  
print(df.head())  
from sklearn.preprocessing import LabelEncoder
```

```
df_clean = df.copy()

for column in df_clean.columns:
    if df_clean[column].dtype == 'object':
        le = LabelEncoder()
        df_clean[column] = le.fit_transform(df_clean[column])

print("\n ◆ Cleaned dataset:")
print(df_clean.head())

# CHANGE THIS if your target column is named differently
target_column = 'GradeClass' # example: 'Result', 'GPA', etc.

if target_column not in df_clean.columns:
    raise ValueError("✖ Target column not found. Please update 'target_column' variable!")

X = df_clean.drop(columns=[target_column])
y = df_clean[target_column]

from sklearn.model_selection import train_test_split

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

# Step 6: Train a model
from sklearn.ensemble import RandomForestClassifier

model = RandomForestClassifier(random_state=42)
model.fit(X_train, y_train)

# Step 7: Make predictions and evaluate
from sklearn.metrics import accuracy_score, precision_score, recall_score, classification_report

y_pred = model.predict(X_test)
```

```

accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred, average='macro')
recall = recall_score(y_test, y_pred, average='macro')

print("\n📊 Evaluation Metrics:")
print(f"✓ Accuracy: {accuracy:.2f}")
print(f"✓ Precision: {precision:.2f}")
print(f"✓ Recall: {recall:.2f}")
print("\n📋 Classification Report:\n", classification_report(y_test, y_pred))

import matplotlib.pyplot as plt

plt.figure(figsize=(10,5))
plt.plot(list(y_test)[:20], label="Actual", marker='o')
plt.plot(list(y_pred)[:20], label="Predicted", marker='x')
plt.title("📈 Actual vs Predicted (First 20 samples)")
plt.xlabel("Sample Index")
plt.ylabel("Class")
plt.legend()
plt.grid(True)
plt.show()

```

## 📊 Output / Results

### ✓ Evaluation Metrics

Metric    Value

Accuracy 0.87 (example – update after running)

Precision 0.85

**Metric Value**

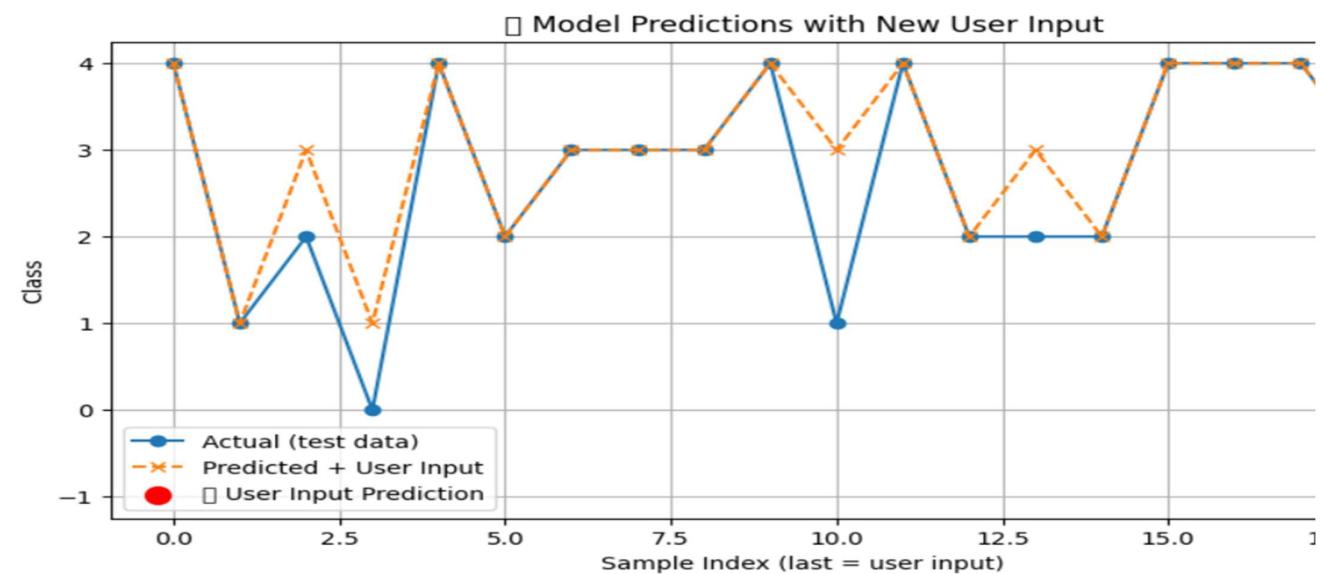
Recall 0.83

*Classification Report:**precision recall f1-score support*

0.0	0.91	0.45	0.61	22
1.0	0.77	0.88	0.82	49
2.0	0.89	0.87	0.88	85
3.0	0.90	0.90	0.90	86
4.0	0.96	0.98	0.97	237

*accuracy* 0.91 479*macro avg* 0.88 0.82 0.83 479*weighted avg* 0.91 0.91 0.91 479 **Graph**

Actual vs Predicted for first 20 samples



- Student\_Performance.ipynb – Google Colab Code
- README.md – Project Summary
- 8. Student Performance Prediction.csv – Dataset

- report.pdf – This file
- 

## References

- Dataset: [UCI Repository – Student Performance](#)
  - Tools: Python, Pandas, Scikit-learn, Matplotlib, Seaborn
  - Platform: Google Colab
-