



Assessment Report
on
“Student Performance Prediction”
submitted as partial fulfillment for the award of
BACHELOR OF TECHNOLOGY
DEGREE

SESSION 2024-25

in
CSE(AI&ML)

By

Vanshika Arora(202401100400208)

Under the supervision of

“Mr.Abhishek Shukla”

KIET Group of Institutions, Ghaziabad

Affiliated to

Dr. A.P.J. Abdul Kalam Technical University, Lucknow
(Formerly UPTU)

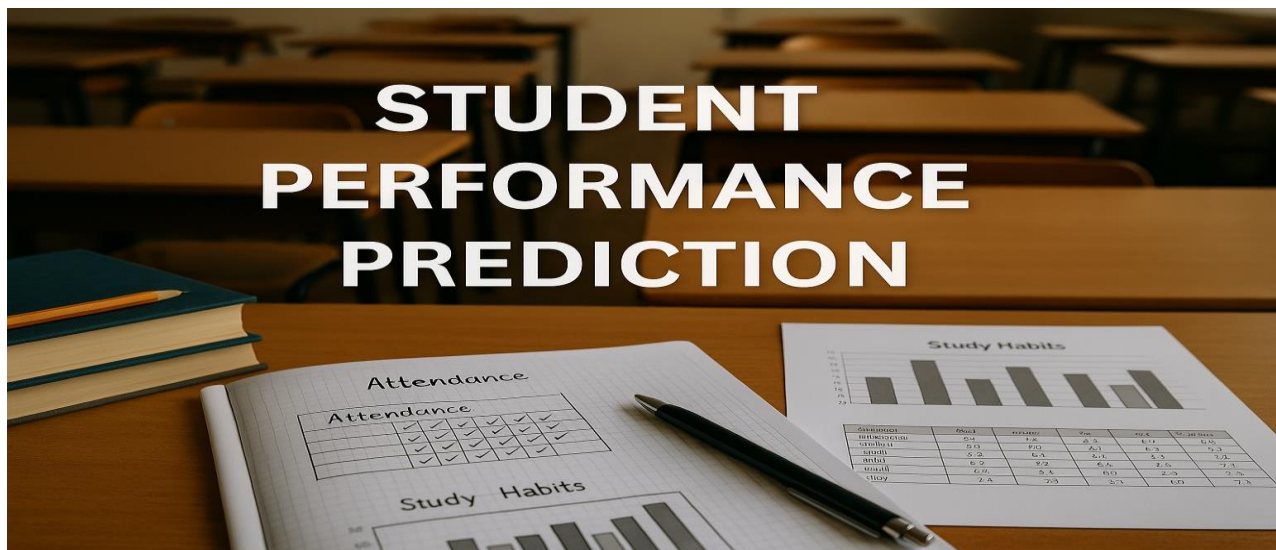
INTRODUCTION

Predicting student performance is a valuable tool for educational institutions to identify at-risk students and implement timely interventions. In this project, we used a classification model to determine whether a student will **pass or fail** based on their academic behaviour, attendance, and support systems.

We used a dataset containing features such as:

- Absences
- Weekly study time
- Tutoring support
- Parental support
- GPA (used to define pass/fail)

The goal was to build a machine learning model that could classify students accurately based on these features. We selected a **Random Forest Classifier** for this task, given its ability to handle both categorical and numerical data effectively and its robustness against overfitting. The model is trained and evaluated using standard performance metrics like **accuracy**, **precision**, and **recall**, with visualization through a **confusion matrix heatmap** for better interpretability.



METHODOLOGY

1. Dataset Preparation:

- Uploaded the dataset Student Performance Prediction.csv to Google Colab.
- A new column Pass was created: students with $GPA \geq 2.0$ were labeled as "Pass" (1), otherwise "Fail" (0).

2. Feature Selection:

- **Selected features:**

Absences, StudyTimeWeekly, Tutoring, and ParentalSupport.

3. Model Building:

- Used a **Random Forest Classifier**, known for handling mixed data types and providing high accuracy.
- The dataset was split into training and testing sets (70/30 split).

4. Evaluation:

- **Confusion Matrix:** Visualized as a heatmap to assess the number of true positives, true negatives, false positives, and false negatives.
- **Accuracy:** Measures the overall correctness of the model.
- **Precision:** Indicates the proportion of correctly predicted positive observations among all predicted positives.
- **Recall:** Measures the proportion of actual positive cases that were correctly identified.

CODE :

Step 1: Upload the dataset

```
from google.colab import files
```

```
uploaded = files.upload()
```

Step 2: Import libraries and load the data

```
import pandas as pd
```

```
import seaborn as sns
```

```
import matplotlib.pyplot as plt
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.ensemble import RandomForestClassifier
```

```
from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score
```

Load the dataset

```
df = pd.read_csv("8. Student Performance Prediction.csv")
```

Step 3: Define Pass/Fail label based on GPA

```
df['Pass'] = (df['GPA'] >= 2.0).astype(int)
```

Step 4: Select features and labels

```
features = ['Absences', 'StudyTimeWeekly', 'Tutoring', 'ParentalSupport']
```

```
X = df[features]
```

```
y = df['Pass']
```

Step 5: Train-test split

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
```

Step 6: Train the model

```
model = RandomForestClassifier(random_state=42)
```

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

Step 7: Make predictions

```
y_pred = model.predict(X_test)
```

Step 8: Confusion Matrix

```
cm = confusion_matrix(y_test, y_pred)
```

```
labels = ['Fail', 'Pass']
```

Step 9: Plot Confusion Matrix Heatmap

```
plt.figure(figsize=(6, 4))
```

```
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=labels, yticklabels=labels)
```

```
plt.xlabel('Predicted')
```

```
plt.ylabel('Actual')
```

```
plt.title('Confusion Matrix Heatmap')
```

```
plt.tight_layout()
```

```
plt.show()
```

Step 10: Evaluation Metrics

```
accuracy = accuracy_score(y_test, y_pred)
```

```
precision = precision_score(y_test, y_pred)
```

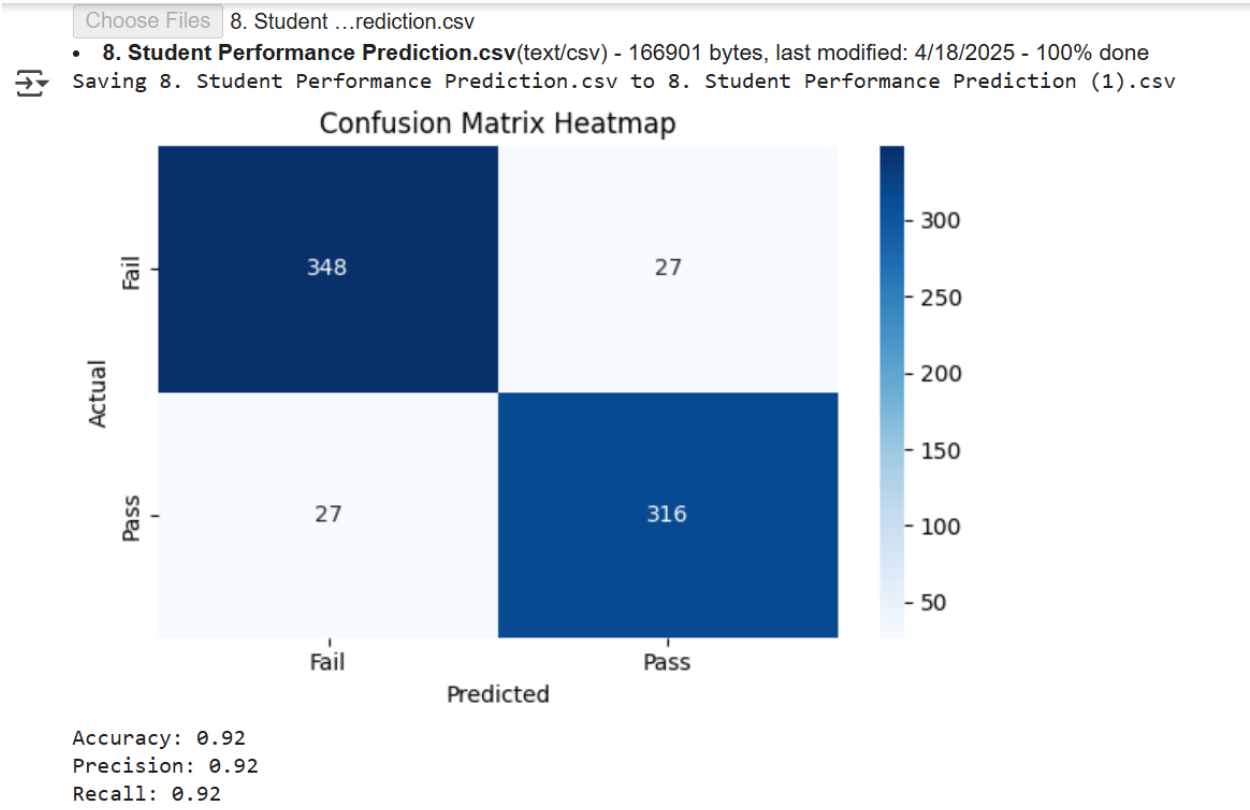
```
recall = recall_score(y_test, y_pred)
```

```
print(f'Accuracy: {accuracy:.2f}')
```

```
print(f'Precision: {precision:.2f}')
```

```
print(f'Recall: {recall:.2f}')
```

OUTPUT



REFERENCES/CREDITS

- Dataset: *Student Performance Prediction.csv* (provided)
- Libraries: pandas, scikit-learn, matplotlib, seaborn
- Platform: Google Colab (<https://colab.research.google.com/>)
- Random Forest Classifier: Scikit-learn Documentation