





## **Assessment Report**

on

# "Predict Student Dropout"

submitted as partial fulfillment for the award of

# BACHELOR OF TECHNOLOGY DEGREE

**SESSION 2024-25** 

in

CSE(AI)

By

Name: Harsh Srivastava

Roll Number: 202401100300115

Section: B

Under the supervision of

"SHIVANSH SIR"

KIET Group of Institutions, Ghaziabad

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#### 1. Introduction

## **Student Dropout Risk Prediction**

Using Decision Tree Classifier in Google Colab

#### 2. Problem Statement

- \* Predict student dropout risk using machine learning
- \* Based on attendance, grades, and participation
- \* Model: Decision Tree Classifier (for interpretability & performance)
- \* Platform: Google Colab

## 3. Dataset Description

#### Features:

- Attendance (%): 0–100
- Grades: 0–10 scale
- Participation: 0–10 scale

#### Target:

- Dropout Risk: Yes / No
- •

## 4. Methodology

• Data Loading

- CSV file uploaded via Google Colab
- Loaded with pandas.read\_csv()
- python
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- df = pd.read\_csv('student\_data.csv')

#### **Data Preprocessing - Inspection**

- Used .head() to preview first 5 rows
- Used .isnull().sum() to check for missing values
  - ✓ No missing values

#### Slide 6: Data Preprocessing - Features & Target

- Features (X): ['attendance', 'grades', 'participation']
- Target (y): 'dropout\_risk'
- Label Encoded: Yes  $\rightarrow$  1, No  $\rightarrow$  0

python

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from sklearn.preprocessing import LabelEncoder

le = LabelEncoder()

y = le.fit\_transform(df['dropout\_risk'])

#### Slide 7: Train-Test Split

• 80% Training, 20% Testing

• Used train\_test\_split() from sklearn

python

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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)

#### **Slide 8: Model Selection**

**Model Chosen: Decision Tree Classifier** 

#### Why?

- Handles non-linear patterns
- Provides decision rules (interpretability)
- Suitable for small/medium datasets

#### **Slide 9: Model Training**

python

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from sklearn.tree import DecisionTreeClassifier

model = DecisionTreeClassifier()

model.fit(X\_train, y\_train)

#### **Slide 10: Model Evaluation - Predictions**

python

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y\_pred = model.predict(X\_test)

#### **Slide 11: Evaluation Metrics**

• Accuracy Score: Overall correctness

• **Precision**: % of predicted dropouts that were correct

• Recall: % of actual dropouts correctly identified

• **F1-Score**: Balance between precision and recall

python

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from sklearn.metrics import classification\_report, accuracy\_score

#### **Slide 12: Confusion Matrix**

#### **Confusion Matrix Explanation:**

• **TP**: Correctly predicted dropouts

• TN: Correctly predicted non-dropouts

• **FP**: Predicted dropout but not actual

• FN: Missed predicting actual dropout

Visualized using seaborn heatmap:

python

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sns.heatmap(confusion\_matrix, annot=True, cmap='Blues')

#### **Slide 13: Interactive Prediction**

### **Manual Student Input Example:**

python

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new\_data = [[85, 7.5, 6]]

```
prediction = model.predict(new_data)

if prediction[0] == 1:
    print(" *** At Risk")

else:
    print(" *** Not at Risk")
```

#### Slide 14: Conclusion

- ☑ Built an interpretable and accurate student dropout predictor
- ✓ Uses Decision Tree for easy rule extraction
- ✓ Can support early interventions and policy-making in education

# \*CODE

```
# Step 1: Upload the dataset
from google.colab import files
uploaded = files.upload()
import pandas as pd
import io

# Load the uploaded file
filename = list(uploaded.keys())[0]
df = pd.read_csv(io.ByteslO(uploaded[filename]))
```

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# Step 2: Preview the data
print("First 5 rows:")
print(df.head())
# Step 3: Check for missing values
print("\nMissing values:")
print(df.isnull().sum())
# Step 4: Preprocess columns
features = ['attendance', 'grades', 'participation']
target = 'dropout_risk'
# Convert target to numeric
df[target] = df[target].map({'yes': 1, 'no': 0})
# Step 5: Train/test split
from sklearn.model_selection import train_test_split
X = df[features]
y = df[target]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Step 6: Train Decision Tree
from sklearn.tree import DecisionTreeClassifier
model = DecisionTreeClassifier(random_state=42)
```

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model.fit(X_train, y_train)
# Step 7: Predict and evaluate
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
y_pred = model.predict(X_test)
print("\nClassification Report:")
print(classification_report(y_test, y_pred))
# Accuracy
acc = accuracy_score(y_test, y_pred)
print(f"Model Accuracy: {acc:.2f}")
# Confusion Matrix
import matplotlib.pyplot as plt
import seaborn as sns
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
plt.title("Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```

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# Step 8: Manually enter student data to test

print("\n--- Predict Dropout Risk for a New Student ---")

attendance = float(input("Enter attendance (0-100): "))

grades = float(input("Enter grades (0-10): "))

participation = float(input("Enter participation score (0-10): "))

new_student = pd.DataFrame([[attendance, grades, participation]], columns=features)

prediction = model.predict(new_student)[0]

result = " * At Risk of Dropping Out" if prediction == 1 else " * Not at Risk"

print("\nPrediction for this student:", result)
```



