

Personalized Student Recommendations

Main code

```
import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn  
as sns from sklearn.linear_model import LinearRegression from sklearn.model_selection  
import train_test_split from sklearn.metrics import mean_absolute_error
```

----- 1. Load the Data -----

File Paths (Replace with actual file locations)

```
current_quiz_file = "C:\Users\Ramya\Downloads\current_quiz.csv" historical_quiz_file =  
"C:\Users\Ramya\Downloads\historical_quiz.csv"  
  
try: df_current = pd.read_csv(current_quiz_file) df_historical =  
pd.read_csv(historical_quiz_file) print("Data Loaded Successfully!") except Exception as  
e: print(f"Error loading data: {e}") exit()
```

----- 2. Analyze Student Performance -----

```
def analyze_performance(student_id): """Analyze accuracy trends, weak topics, and  
difficulty levels for a student.""" student_data = df_historical[df_historical['student_id']  
== student_id]  
  
if student_data.empty:  
    print(f"No data found for Student ID: {student_id}")  
    return None, None  
  
# Calculate accuracy by topic  
topic_accuracy = student_data.groupby('topic')['correct'].mean().reset_index()  
topic_accuracy.columns = ['Topic', 'Accuracy']  
  
# Calculate accuracy by difficulty level  
difficulty_analysis = student_data.groupby('difficulty')['correct'].mean().reset_index()  
difficulty_analysis.columns = ['Difficulty Level', 'Accuracy']
```

```
return topic_accuracy, difficulty_analysis
```

----- 3. Generate Personalized Recommendations -----

```
def generate_recommendations(student_id): """Suggest topics and difficulty levels to
focus on.""" student_data = df_historical[df_historical['student_id'] == student_id]
```

```
if student_data.empty:
    return "No data available for recommendations."
```

```
# Identify weak topics (lowest accuracy)
weak_topics = student_data[student_data['correct'] ==
0]['topic'].value_counts().head(3).index.tolist()
```

```
# Identify easy questions student should start with
easy_questions = student_data[student_data['difficulty'] ==
'Easy']['topic'].value_counts().head(3).index.tolist()
```

```
recommendations = {
    "Focus More On": weak_topics if weak_topics else "No weak topics detected",
    "Start With Easy Topics": easy_questions if easy_questions else "No easy topics
detected",
    "Practice More MCQs": "Increase practice in weak areas"
}
```

```
return recommendations
```

----- 4. Define Student Persona -----

```
def define_student_persona(student_id): """Categorize students based on performance
patterns.""" student_data = df_historical[df_historical['student_id'] == student_id]
```

```
if student_data.empty:
    return "No data available for persona classification."
```

```
avg_score = student_data['score'].mean()
```

```

if avg_score > 75:
    persona = "Top Performer - Maintain consistency!"
elif avg_score > 50:
    persona = "Average Performer - Improve weak areas!"
else:
    persona = "Needs Improvement - Focus on basics!"

return persona

```

----- 5. Predict NEET Rank -----

```

def predict_neet_rank(): """Predict student's NEET rank based on quiz performance
using Linear Regression.""" if 'score' not in df_historical.columns: print("Error: 'score'
column missing in dataset.") return None

df_historical['total_score'] = df_historical.groupby('student_id')['score'].transform('sum')

# Simulated NEET Scores (Replace with actual data)
df_historical['neet_rank'] = 1000 - df_historical['total_score'] # Hypothetical formula

X = df_historical[['total_score']]
y = df_historical['neet_rank']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

error = mean_absolute_error(y_test, y_pred)
print(f"NEET Rank Prediction Error: {error:.2f}")

return model

```

----- 6. Main Execution -----

```

if name == "main": student_id = "Student_1" # Change this to any student ID from the
dataset

```

```
# Analyze Performance
topic_accuracy, difficulty_analysis = analyze_performance(student_id)

if topic_accuracy is not None:
    print("Topic-wise Accuracy:\n", topic_accuracy)
    print("Difficulty Level Analysis:\n", difficulty_analysis)

# Generate Recommendations
recommendations = generate_recommendations(student_id)
print("Personalized Recommendations:", recommendations)

# Define Student Persona
persona = define_student_persona(student_id)
print(f"Student Persona for {student_id}: {persona}")

# Predict NEET Rank
model = predict_neet_rank()

# Visualize Insights if data is available
if topic_accuracy is not None:
    plt.figure(figsize=(10, 5))
    sns.barplot(x="Topic", y="Accuracy", data=topic_accuracy, palette="coolwarm")
    plt.title("Topic-wise Accuracy")
    plt.xticks(rotation=45)
    plt.show()
```

READEME :

Personalized-Student-Recommendations

Personalized Student Recommendations for NEET Preparation

Data Overview

This project utilizes two key datasets to analyze and provide personalized recommendations for NEET preparation based on quiz performance:

Current Quiz Data:

Contains details of a student's most recent quiz submission. Includes data such as questions, topics, responses, and the overall score for the latest quiz taken. This dataset serves as the real-time data source for the student's performance.

Historical Quiz Data:

Contains performance data from the last 5 quizzes taken by each student. Includes scores, response maps (which are stored as Key-Value pairs representing the question ID and the selected option ID), and other relevant metrics. This dataset serves to analyze trends and performance patterns over time.

Task Overview

The goal of this project is to develop a Python-based solution that analyzes the quiz performance and generates personalized recommendations for students to improve their NEET preparation.

1. **Analyze the Data:** Explore the dataset schema and identify patterns in student performance based on: **Topics:** Which topics the student excels in and which they struggle with. **Difficulty Levels:** How the student performs across different difficulty levels. **Response Accuracy:** How accurately the student responds to quiz questions.
2. **Generate Insights:** **Weak Areas:** Identify which topics or areas the student needs to focus on based on low accuracy. **Improvement Trends:** Track performance over time and highlight areas where the student has shown improvement. **Performance Gaps:** Identify areas where the student is underperforming and suggest improvements.
3. **Create Recommendations:** **Actionable Steps:** Based on the analysis, generate actionable steps that the student can take to improve their preparation: Focus on weak topics. Practice questions from specific difficulty levels. Review past quiz responses to identify patterns in mistakes. **Bonus Points:** **Student Persona:** Analyze and categorize students based on their quiz performance patterns. Generate insights on strengths and weaknesses using creative labels (e.g., "Top Performer", "Needs Improvement"). **NEET Rank Prediction:** Build a probabilistic model that predicts a student's potential NEET rank based on their quiz performance and historical NEET results.

Project Features:

Personalized Recommendations: Suggest areas to improve based on performance analysis. **Performance Visualization:** Visualize topic-wise accuracy and trends over time. **Probabilistic NEET Rank Prediction:** Predict a student's rank based on their quiz performance and simulate NEET scores. **Student Persona Creation:** Classify students as "Top Performer", "Average Performer", or "Needs Improvement".

Example Output:

Topic-wise Accuracy: A bar chart showing the student's accuracy in different topics. **NEET Rank Prediction:** The model's prediction of the student's NEET rank based on their performance. **Personalized Recommendations:** Suggested areas for improvement and actionable steps.

