



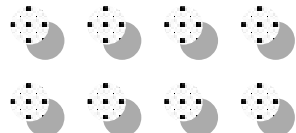
SmartSage – AI that teaches like a wise mentor

By:- Vansh Gupta

03/30/25

Summary

Our project focuses on AI-driven personalized learning for K-12 students. It predicts assessment scores, determines promotion eligibility, and customizes study material based on student levels. The system leverages data analysis, machine learning, and NLP to curate content dynamically for effective learning.





PROBLEM STATEMENTS

- Predict assessment score & promotion decision → Need a model to train on assessment data.
- Decide what to keep/skip at different levels → Need clear rules based on grade levels.
- Predict study material based on student level → Need a model to map materials to student levels.
- Generate adaptive teaching material dynamically → Requires NLP model (Gemini) to simplify/modify content dynamically.

SOLUTIONS

01

- Develop an AI model to analyze past performance, study time, and IQ to predict assessment scores and determine promotion eligibility.

03

- Identify key topics for each grade level, ensuring an optimized learning path by removing redundant or complex topics based on student understanding.

02

- Use AI-driven recommendations to match students with the right learning resources, ensuring personalized and effective learning experiences.

04

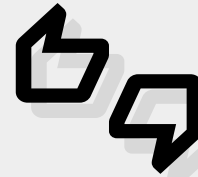
- Implement NLP-based models to simplify, restructure, and generate adaptive learning materials in real time based on student needs.

STEPS TO ACHIEVE

- Step 1: Collect student data (IQ, scores, study time, etc.).



- Step 2: Perform Exploratory Data Analysis (EDA) to find patterns.



- Step 3: Train AI/ML models for prediction and recommendations.
- Step 4: Implement an NLP model to simplify teaching materials.






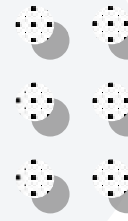
Roles And Responsibility



1. Vansh Gupta (Co-Founder & Project Lead)

- Project Management & Leadership – Overseeing the project's overall vision, execution, and progress.
- Technical Development – Handling coding, software architecture, and integrations.
- Innovation & Problem-Solving – Ensuring new features and optimizations align with project goals.

2. Dipaya Das (Documentation Specialist)

- Report Writing & Documentation – Creating and maintaining structured project reports, user manuals, and technical documentation.
 - Requirement Gathering – Documenting project requirements, use cases, and functional specifications.
 - Presentation & Reports – Preparing PPTs, PDFs, and Word reports for stakeholders.
- 
- 

MODEL USED



- Regression Models – Predict student performance.

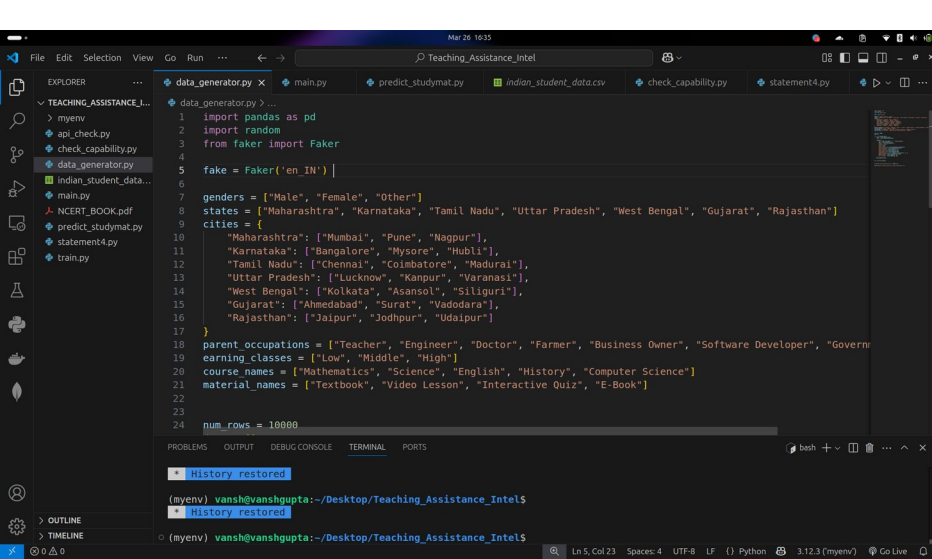


- Classification Models – Identify students at risk.
- NLP Models (Gemini/GPT) – Simplify learning content.

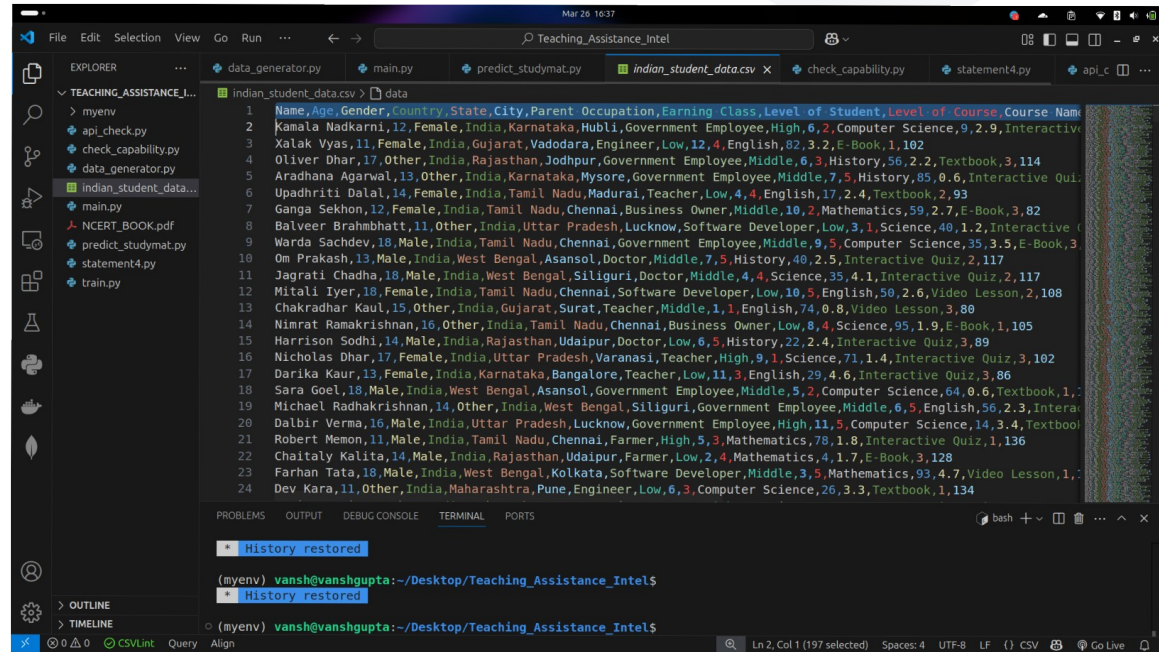


- Recommendation Systems – Suggest study material.

Data

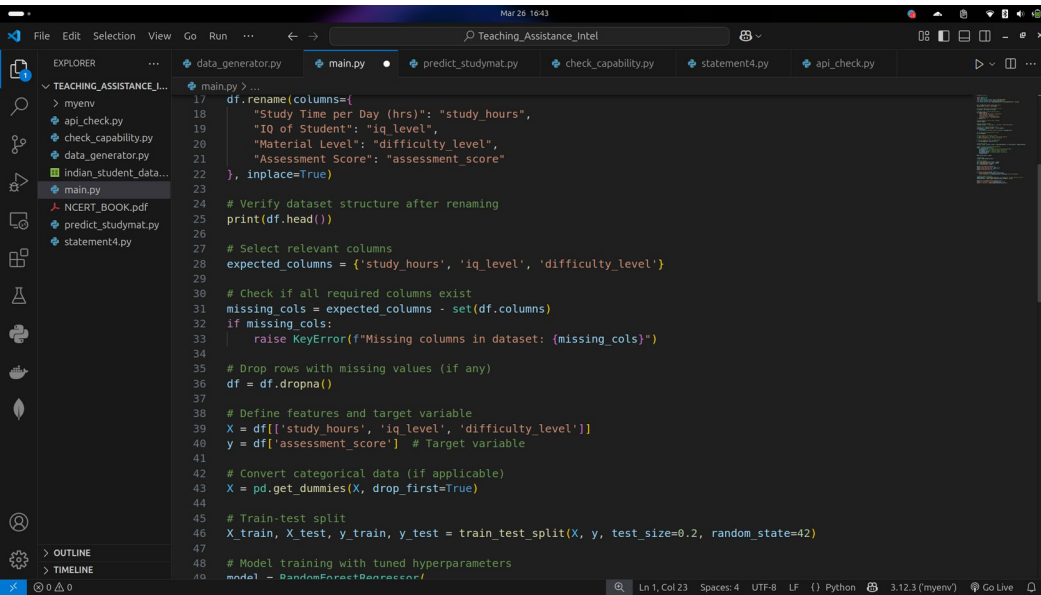


```
1 import pandas as pd
2 import random
3 from faker import Faker
4
5 fake = Faker('en_IN')
6
7 genders = ["Male", "Female", "Other"]
8 states = ["Maharashtra", "Karnataka", "Tamil Nadu", "Uttar Pradesh", "West Bengal", "Gujarat", "Rajasthan"]
9 cities = {
10     "Maharashtra": ["Mumbai", "Pune", "Nagpur"],
11     "Karnataka": ["Bangalore", "Mysore", "Hubli"],
12     "Tamil Nadu": ["Chennai", "Coimbatore", "Madurai"],
13     "Uttar Pradesh": ["Lucknow", "Kanpur", "Varanasi"],
14     "West Bengal": ["Kolkata", "Asansol", "Siliguri"],
15     "Gujarat": ["Ahmedabad", "Surat", "Vadodara"],
16     "Rajasthan": ["Jaipur", "Jodhpur", "Udaipur"]
17 }
18
19 parent_occupations = ["Teacher", "Engineer", "Doctor", "Farmer", "Business Owner", "Software Developer", "Government Employee", "Middle", "High"]
20 earning_classes = ["Low", "Middle", "High"]
21 course_names = ["Mathematics", "Science", "English", "History", "Computer Science"]
22 material_names = ["Textbook", "Video Lesson", "Interactive Quiz", "E-Book"]
23
24 num_rows = 10000
```

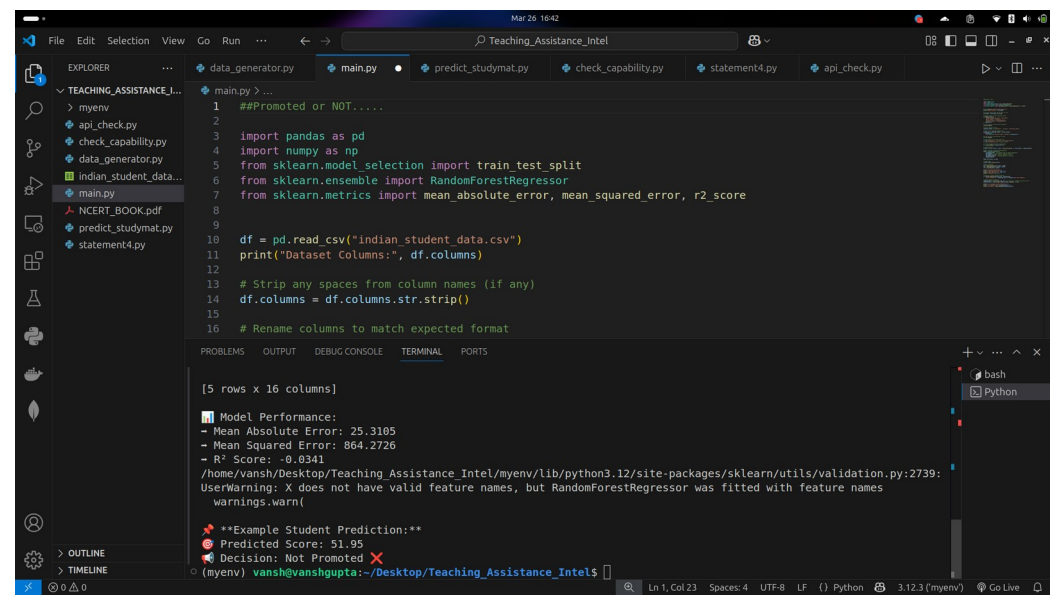


```
1 Name, Age, Gender, Country, State, City, Parent Occupation, Earning Class, Level of Student, Level of Course, Course Name
2 Kamala Nadkarni, 12, Female, India, Karnataka, Hubli, Government Employee, High, 6, 2, Computer Science, 9, 2, 9, Interactive Quiz
3 Xalok Vyas, 11, Female, India, Gujarat, Vadodara, Engineer, Low, 12, 4, English, 82, 3, 2, E-Book, 1, 102
4 Oliver Dhar, 17, Other, India, Rajasthan, Jodhpur, Government Employee, Middle, 6, 3, History, 56, 2, 2, Textbook, 3, 114
5 Aradhana Agarwal, 13, Other, India, Karnataka, Mysore, Government Employee, Middle, 7, 5, History, 85, 0, 6, Interactive Quiz
6 Upadhruti Dalal, 14, Female, India, Tamil Nadu, Madurai, Teacher, Low, 4, 4, English, 17, 2, 4, Textbook, 2, 93
7 Ganga Sekhon, 12, Female, India, Tamil Nadu, Chennai, Business Owner, Middle, 10, 2, Mathematics, 59, 2, 7, E-Book, 3, 82
8 Balveer Brahmabhatt, 11, Other, India, Uttar Pradesh, Lucknow, Software Developer, Low, 3, 1, Science, 40, 1, 2, Interactive Quiz
9 Warda Sachdev, 18, Male, India, Tamil Nadu, Chennai, Government Employee, Middle, 9, 5, Computer Science, 35, 3, 5, E-Book, 3, 82
10 Om Prakash, 13, Male, India, West Bengal, Asansol, Doctor, Middle, 7, 5, History, 40, 2, 5, Interactive Quiz, 2, 117
11 Jagrati Chadha, 18, Male, India, West Bengal, Siliguri, Doctor, Middle, 4, 4, Science, 35, 4, 1, Interactive Quiz, 2, 117
12 Mitali Iyer, 18, Female, India, Tamil Nadu, Chennai, Software Developer, Low, 10, 5, English, 50, 2, 6, Video Lesson, 2, 108
13 Chakradhar Kaul, 15, Other, India, Gujarat, Surat, Teacher, Middle, 1, 1, English, 74, 0, 8, Video Lesson, 3, 80
14 Nimrat Ramakrishnan, 16, Other, India, Tamil Nadu, Chennai, Business Owner, Low, 8, 4, Science, 95, 1, 9, E-Book, 1, 105
15 Harrison Sodhi, 14, Male, India, Rajasthan, Udaipur, Doctor, Low, 6, 5, History, 22, 2, 4, Interactive Quiz, 3, 89
16 Nicholas Dhar, 17, Female, India, Uttar Pradesh, Varanasi, Teacher, High, 9, 1, Science, 71, 1, 4, Interactive Quiz, 3, 102
17 Darika Kaur, 13, Female, India, Karnataka, Bangalore, Teacher, Low, 11, 3, English, 29, 4, 6, Interactive Quiz, 3, 86
18 Sara Goel, 18, Male, India, West Bengal, Asansol, Government Employee, Middle, 5, 2, Computer Science, 64, 0, 6, Textbook, 1, 102
19 Michael Radhakrishnan, 14, Other, India, West Bengal, Siliguri, Government Employee, Middle, 6, 5, English, 56, 2, 3, Interactive Quiz, 3, 86
20 Dalbir Verma, 16, Male, India, Uttar Pradesh, Lucknow, Government Employee, High, 11, 5, Computer Science, 14, 3, 4, Textbook, 1, 102
21 Robert Memon, 11, Male, India, Tamil Nadu, Chennai, Farmer, High, 5, 3, Mathematics, 78, 1, 8, Interactive Quiz, 1, 136
22 Chaitaly Kalita, 14, Male, India, Rajasthan, Udaipur, Farmer, Low, 2, 4, Mathematics, 4, 1, 7, E-Book, 3, 128
23 Farhan Tata, 18, Male, India, West Bengal, Kolkata, Software Developer, Middle, 3, 5, Mathematics, 93, 4, 7, Video Lesson, 1, 102
24 Dev Kara, 11, Other, India, Maharashtra, Pune, Engineer, Low, 6, 3, Computer Science, 26, 3, 3, Textbook, 1, 134
```


PROMOTED OR NOT



```
17 df.rename(columns={
18     "Study Time per Day (hrs)": "study_hours",
19     "IQ of Student": "iq_level",
20     "Material Level": "difficulty_level",
21     "Assessment Score": "assessment_score"
22 }, inplace=True)
23
24 # Verify dataset structure after renaming
25 print(df.head())
26
27 # Select relevant columns
28 expected_columns = {'study_hours', 'iq_level', 'difficulty_level'}
29
30 # Check if all required columns exist
31 missing_cols = expected_columns - set(df.columns)
32 if missing_cols:
33     raise KeyError(f"Missing columns in dataset: {missing_cols}")
34
35 # Drop rows with missing values (if any)
36 df = df.dropna()
37
38 # Define features and target variable
39 X = df[['study_hours', 'iq_level', 'difficulty_level']]
40 y = df['assessment_score'] # Target variable
41
42 # Convert categorical data (if applicable)
43 X = pd.get_dummies(X, drop_first=True)
44
45 # Train-test split
46 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
47
48 # Model training with tuned hyperparameters
49 model = RandomForestRegressor(
```



```
1 ##Promoted or NOT....
2
3 import pandas as pd
4 import numpy as np
5 from sklearn.model_selection import train_test_split
6 from sklearn.ensemble import RandomForestRegressor
7 from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
8
9
10 df = pd.read_csv("indian_student_data.csv")
11 print("Dataset Columns:", df.columns)
12
13 # Strip any spaces from column names (if any)
14 df.columns = df.columns.str.strip()
15
16 # Rename columns to match expected format
17
18
19 [5 rows x 16 columns]
20
21 Model Performance:
22 - Mean Absolute Error: 25.3105
23 - Mean Squared Error: 864.2726
24 - R^2 Score: -0.0341
25 /home/vansh/Desktop/Teaching_Assistance_Intel/myenv/lib/python3.12/site-packages/sklearn/utils/validation.py:2739:
26 UserWarning: X does not have valid feature names, but RandomForestRegressor was fitted with feature names
27 warnings.warn(
28
29 **Example Student Prediction:**
30 Predicted Score: 51.95
31 Decision: Not Promoted
32 (myenv) vansh@vanshgupta:~/Desktop/Teaching_Assistance_Intel$
```

STUDY MATERIAL PREDICTION

The image displays two side-by-side screenshots of a Jupyter Notebook interface, showing code for a study material prediction model.

Left Screenshot (predict_studymat.py):

```
1 ##Study Material Prediction According to Student [...]
2
3 import pandas as pd
4 import numpy as np
5 from sklearn.model_selection import train_test_split
6 from sklearn.ensemble import RandomForestClassifier
7 from sklearn.preprocessing import StandardScaler, LabelEncoder
8 from sklearn.metrics import accuracy_score, classification_report
9 from sklearn.pipeline import Pipeline
10 from sklearn.compose import ColumnTransformer
11 from sklearn.preprocessing import OneHotEncoder
12 file_path = "indian_student_data.csv"
13 # Load dataset
14 def load_and_preprocess_data(file_path):
15     try:
16         # Load dataset
```

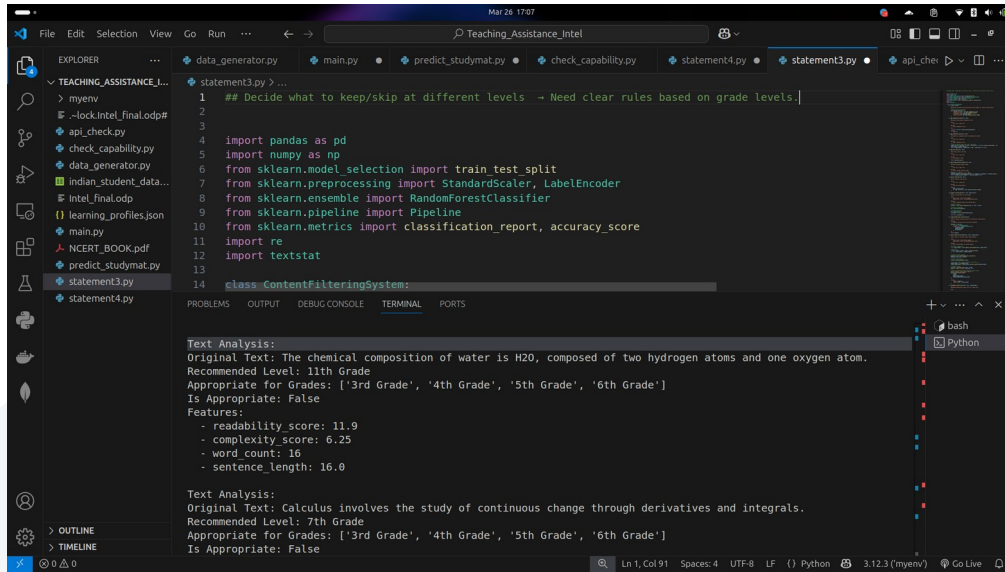
The bottom of the left screenshot shows a **Classification Report** table:

	precision	recall	f1-score	support
E-Book	0.28	0.28	0.28	515
Interactive Quiz	0.26	0.23	0.24	501
Textbook	0.22	0.23	0.23	478
Video Lesson	0.25	0.26	0.26	506
accuracy 0.25 2000				
macro avg 0.25 0.25 0.25 2000				
weighted avg 0.25 0.25 0.25 2000				

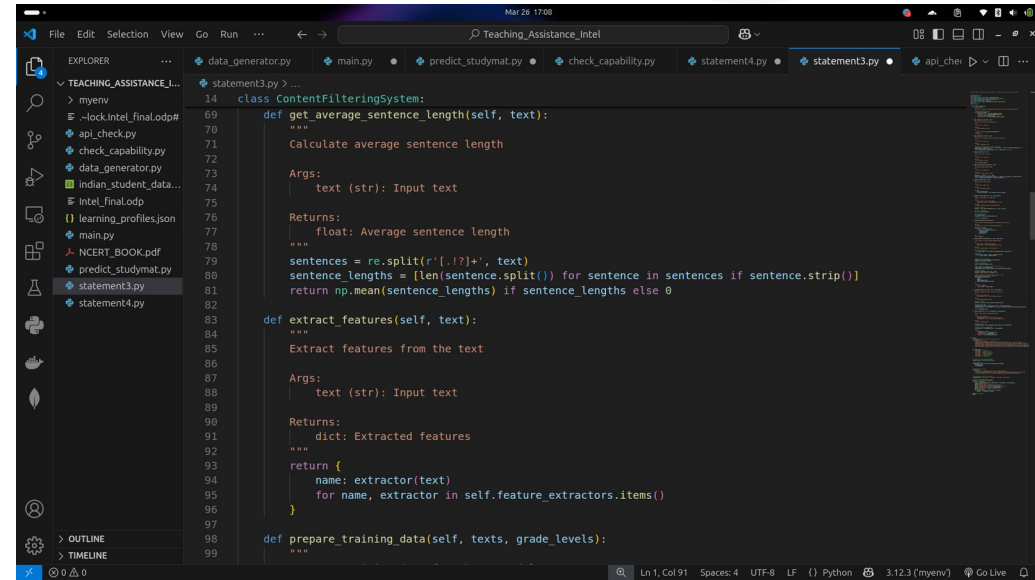
Right Screenshot (main.py):

```
53 def create_ml_pipeline():
54     # Preprocessing for numerical features
55     preprocessor = ColumnTransformer(
56         transformers=[
57             ('num', StandardScaler(), numeric_features)
58         ])
59
60     # Create pipeline
61     pipeline = Pipeline([
62         ('preprocessor', preprocessor),
63         ('classifier', RandomForestClassifier(
64             n_estimators=300,
65             max_depth=20,
66             min_samples_split=3,
67             random_state=42
68         ))
69     ])
70
71     return pipeline
72
73
74
75
76 # Main training and evaluation function
77 def train_and_evaluate_model(file_path):
78     # Load and preprocess data
79     df = load_and_preprocess_data(file_path)
80     if df is None:
81         return None
82
83     # Prepare features and target
84     X, y, label_encoder = prepare_data(df)
85
86     # Split the data
87     X_train, X_test, y_train, y_test = train_test_split(
88         X, y, test_size=0.2, random_state=42
```

KEEP/SKIP DIFFERNT LEVEL

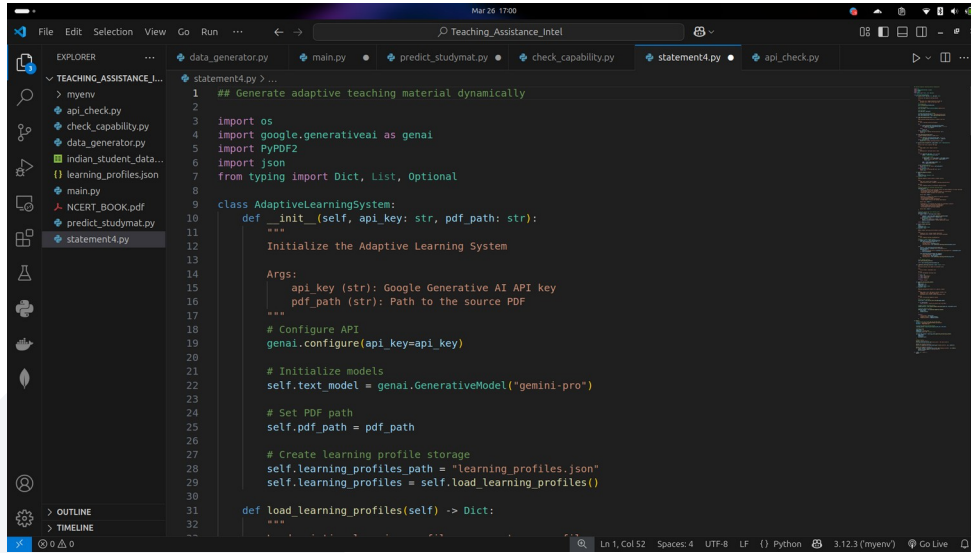


```
1  ## Decide what to keep/skip at different levels -> Need clear rules based on grade levels.
2
3
4  import pandas as pd
5  import numpy as np
6  from sklearn.model_selection import train_test_split
7  from sklearn.preprocessing import StandardScaler, LabelEncoder
8  from sklearn.ensemble import RandomForestClassifier
9  from sklearn.pipeline import Pipeline
10 from sklearn.metrics import classification_report, accuracy_score
11 import re
12 import textstat
13
14 class ContentFilteringSystem:
15
16     def __init__(self):
17         self.feature_extractors = {}
18
19     def get_average_sentence_length(self, text):
20         """
21         Calculate average sentence length
22
23         Args:
24             text (str): Input text
25
26         Returns:
27             float: Average sentence length
28         """
29         sentences = re.split(r'[.!?]+', text)
30         sentence_lengths = [len(sentence.split()) for sentence in sentences if sentence.strip()]
31         return np.mean(sentence_lengths) if sentence_lengths else 0
32
33     def extract_features(self, text):
34         """
35         Extract features from the text
36
37         Args:
38             text (str): Input text
39
40         Returns:
41             dict: Extracted features
42         """
43         return {
44             name: extractor(text)
45             for name, extractor in self.feature_extractors.items()
46         }
47
48     def prepare_training_data(self, texts, grade_levels):
49         """
50         Prepare training data for the model
51
52         Args:
53             texts (list): List of texts
54             grade_levels (list): List of recommended grade levels
55
56         Returns:
57             tuple: (X, y) where X is a matrix of features and y is a list of recommended grade levels
58         """
59         X = []
60         y = []
61         for text, grade_level in zip(texts, grade_levels):
62             features = self.extract_features(text)
63             X.append(features)
64             y.append(grade_level)
65
66         return np.array(X), y
67
68     def train(self):
69         """
70         Train the model
71
72         Returns:
73             RandomForestClassifier: Trained model
74         """
75         X, y = self.prepare_training_data(self.texts, self.grade_levels)
76         model = RandomForestClassifier()
77         model.fit(X, y)
78         return model
79
80     def predict(self, text):
81         """
82         Predict the recommended grade level for a given text
83
84         Args:
85             text (str): Input text
86
87         Returns:
88             str: Recommended grade level
89         """
90         features = self.extract_features(text)
91         model = self.train()
92         prediction = model.predict([features])
93         return prediction[0]
94
95     def analyze_text(self, text):
96         """
97         Analyze a text and provide recommendations
98
99         Args:
100            text (str): Input text
101
102         Returns:
103            dict: Analysis results including original text, recommended level, appropriate grades, and features
104         """
105         original_text = text
106         recommended_level = self.predict(text)
107         appropriate_grades = self.get_appropriate_grades(recommended_level)
108         is_appropriate = recommended_level in appropriate_grades
109         features = self.extract_features(text)
110
111         return {
112             'original_text': original_text,
113             'recommended_level': recommended_level,
114             'appropriate_grades': appropriate_grades,
115             'is_appropriate': is_appropriate,
116             'features': features
117         }
118
119     def get_appropriate_grades(self, recommended_level):
120         """
121         Get appropriate grades for a recommended level
122
123         Args:
124             recommended_level (str): Recommended grade level
125
126         Returns:
127             list: List of appropriate grades
128         """
129         appropriate_grades = []
130         for grade in self.grade_levels:
131             if grade <= recommended_level:
132                 appropriate_grades.append(grade)
133         return appropriate_grades
134
135 # Example usage
136 if __name__ == '__main__':
137     system = ContentFilteringSystem()
138
139     # Text 1: Chemistry
140     text1 = "The chemical composition of water is H2O, composed of two hydrogen atoms and one oxygen atom."
141     analysis1 = system.analyze_text(text1)
142     print("Text Analysis:")
143     print(f"Original Text: {text1}")
144     print(f"Recommended Level: {analysis1['recommended_level']}")
145     print(f"Appropriate for Grades: {analysis1['appropriate_grades']}")
146     print(f"Is Appropriate: {analysis1['is_appropriate']}")
147     print("Features:")
148     for feature, value in analysis1['features'].items():
149         print(f"- {feature}: {value}")
150
151     # Text 2: Calculus
152     text2 = "Calculus involves the study of continuous change through derivatives and integrals."
153     analysis2 = system.analyze_text(text2)
154     print("Text Analysis:")
155     print(f"Original Text: {text2}")
156     print(f"Recommended Level: {analysis2['recommended_level']}")
157     print(f"Appropriate for Grades: {analysis2['appropriate_grades']}")
158     print(f"Is Appropriate: {analysis2['is_appropriate']}")
```

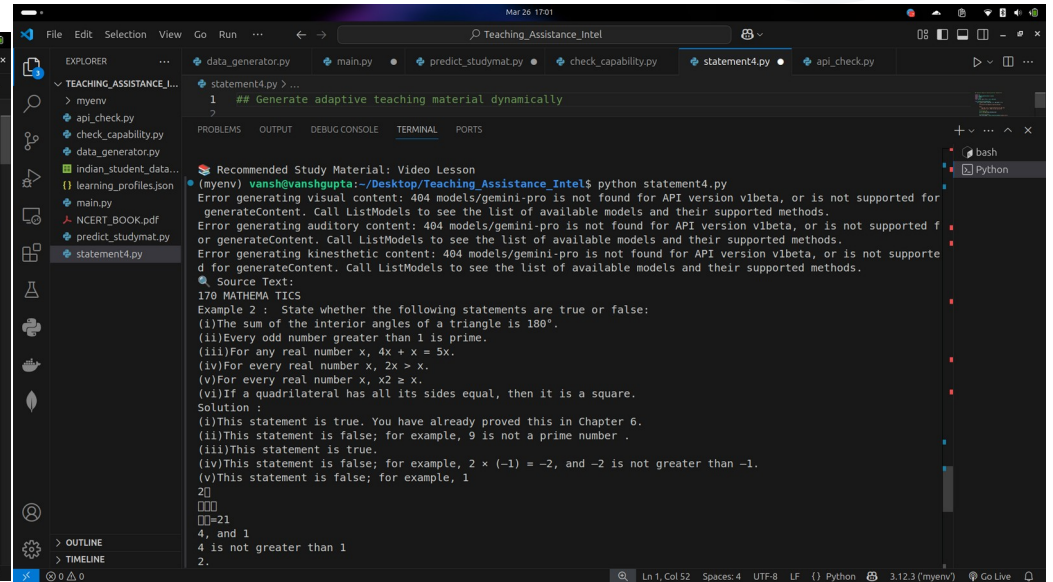


```
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26         Returns:
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29         sentences = re.split(r'[.!?]+', text)
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43         return {
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48     def prepare_training_data(self, texts, grade_levels):
49         """
50         Prepare training data for the model
51
52         Args:
53             texts (list): List of texts
54             grade_levels (list): List of recommended grade levels
55
56         Returns:
57             tuple: (X, y) where X is a matrix of features and y is a list of recommended grade levels
58         """
59         X = []
60         y = []
61         for text, grade_level in zip(texts, grade_levels):
62             features = self.extract_features(text)
63             X.append(features)
64             y.append(grade_level)
65
66         return np.array(X), y
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73             RandomForestClassifier: Trained model
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78         return model
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80     def predict(self, text):
81         """
82         Predict the recommended grade level for a given text
83
84         Args:
85             text (str): Input text
86
87         Returns:
88             str: Recommended grade level
89         """
90         features = self.extract_features(text)
91         model = self.train()
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93         return prediction[0]
94
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104         """
105         original_text = text
106         recommended_level = self.predict(text)
107         appropriate_grades = self.get_appropriate_grades(recommended_level)
108         is_appropriate = recommended_level in appropriate_grades
109         features = self.extract_features(text)
110
111         return {
112             'original_text': original_text,
113             'recommended_level': recommended_level,
114             'appropriate_grades': appropriate_grades,
115             'is_appropriate': is_appropriate,
116             'features': features
117         }
118
119     def get_appropriate_grades(self, recommended_level):
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121         Get appropriate grades for a recommended level
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123         Args:
124             recommended_level (str): Recommended grade level
125
126         Returns:
127             list: List of appropriate grades
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129         appropriate_grades = []
130         for grade in self.grade_levels:
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132                 appropriate_grades.append(grade)
133         return appropriate_grades
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135 # Example usage
136 if __name__ == '__main__':
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139     # Text 1: Chemistry
140     text1 = "The chemical composition of water is H2O, composed of two hydrogen atoms and one oxygen atom."
141     analysis1 = system.analyze_text(text1)
142     print("Text Analysis:")
143     print(f"Original Text: {text1}")
144     print(f"Recommended Level: {analysis1['recommended_level']}")
145     print(f"Appropriate for Grades: {analysis1['appropriate_grades']}")
146     print(f"Is Appropriate: {analysis1['is_appropriate']}")
147     print("Features:")
148     for feature, value in analysis1['features'].items():
149         print(f"- {feature}: {value}")
150
151     # Text 2: Calculus
152     text2 = "Calculus involves the study of continuous change through derivatives and integrals."
153     analysis2 = system.analyze_text(text2)
154     print("Text Analysis:")
155     print(f"Original Text: {text2}")
156     print(f"Recommended Level: {analysis2['recommended_level']}")
157     print(f"Appropriate for Grades: {analysis2['appropriate_grades']}")
158     print(f"Is Appropriate: {analysis2['is_appropriate']}")
```

GENERATE TEACHING MATERIAL



```
1  ## Generate adaptive teaching material dynamically
2
3  import os
4  import google.generativeai as genai
5  import PyPDF2
6  import json
7  from typing import Dict, List, Optional
8
9  class AdaptiveLearningSystem:
10     def __init__(self, api_key: str, pdf_path: str):
11         """
12         Initialize the Adaptive Learning System
13
14         Args:
15             api_key (str): Google Generative AI API key
16             pdf_path (str): Path to the source PDF
17         """
18         # Configure API
19         genai.configure(api_key=api_key)
20
21         # Initialize models
22         self.text_model = genai.GenerativeModel("gemini-pro")
23
24         # Set PDF path
25         self.pdf_path = pdf_path
26
27         # Create learning profile storage
28         self.learning_profiles_path = "learning_profiles.json"
29         self.learning_profiles = self.load_learning_profiles()
30
31     def load_learning_profiles(self) -> Dict:
32         """
```

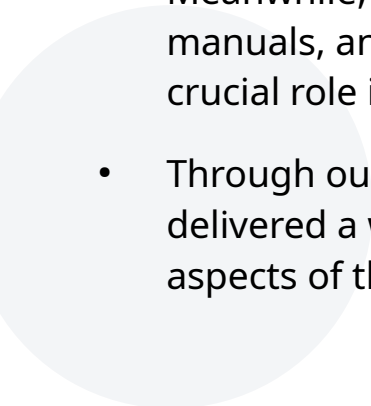
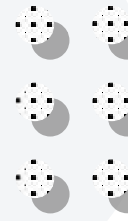


```
1  ## Generate adaptive teaching material dynamically
2
3  import os
4  import google.generativeai as genai
5  import PyPDF2
6  import json
7  from typing import Dict, List, Optional
8
9  class AdaptiveLearningSystem:
10     def __init__(self, api_key: str, pdf_path: str):
11         """
12         Initialize the Adaptive Learning System
13
14         Args:
15             api_key (str): Google Generative AI API key
16             pdf_path (str): Path to the source PDF
17         """
18         # Configure API
19         genai.configure(api_key=api_key)
20
21         # Initialize models
22         self.text_model = genai.GenerativeModel("gemini-pro")
23
24         # Set PDF path
25         self.pdf_path = pdf_path
26
27         # Create learning profile storage
28         self.learning_profiles_path = "learning_profiles.json"
29         self.learning_profiles = self.load_learning_profiles()
30
31     def load_learning_profiles(self) -> Dict:
32         """
```



CONCLUSION



- In this project, we successfully executed our objectives by efficiently dividing tasks and responsibilities. As the project lead, I, Vansh Gupta, managed the overall development, technical implementation, and project execution. My role involved ensuring smooth coordination, decision-making, and driving the project forward with innovative solutions.
 - Meanwhile, Dipaya Das focused on documentation, ensuring that all project reports, technical manuals, and presentations were well-structured and comprehensive. The documentation played a crucial role in maintaining clarity, tracking progress, and presenting our work effectively.
 - Through our collaborative efforts, we achieved the desired outcomes, streamlined the workflow, and delivered a well-documented and technically sound project. This structured approach ensured that all aspects of the project were handled efficiently, leading to a successful completion.
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THANK YOU

