AI-PROJECT

Project Report:

Plagiarism Checker

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Submitted to:

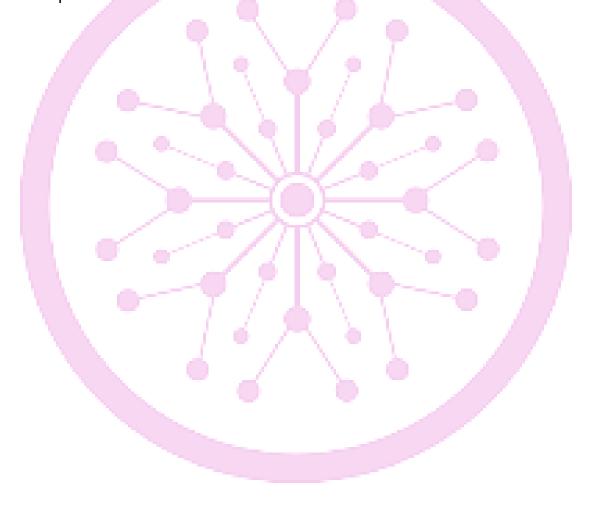
Prof.RASIKH ALI

Program:

BS DATA SCIENCE

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PROJECT OVERVIEW

The Plagiarism Checker is a Python-based web application designed to detect plagiarism and AI-generated text in documents or user-provided text. This project uses Flask for the web interface and machine learning models for analysis. It leverages a TF-IDF vectorizer to process text and a pre-trained classification model to determine if the given text exhibits characteristics of plagiarism.

OBJECTIVES

1. Detect Plagiarism:

Identify instances of text duplication or AI-generated text within the provided input.

2. User-Friendly Interface:

Provide an intuitive web-based interface for users to input and test text for potential plagiarism.

3. Scalability:

Ensure that it is extendable for future enhancements, including integration with larger datasets.

TECHNOLOGIES AND TOOLS USED

Programming Language: Python

Web Framework: Flask

■ Machine Learning:

o Pre-trained classification model (model.pkl)

o TF-IDF Vectorizer (tfidf vectorizer.pkl)

Frontend: HTML templates rendered via Flask

Backend: Flask routes and APIs

Dataset: Provided dataset for reference and training the model.

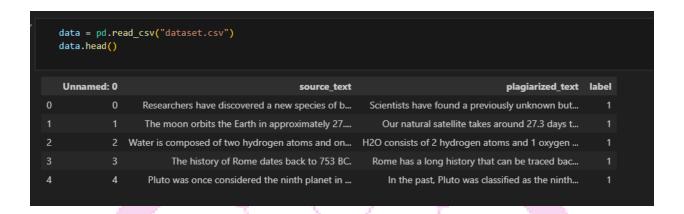
BASIC FUNCTIONS USED

Libraries used:

□ nltk - NLP toolkit
 □ cosine_similarity - Text similarity
 □ pandas - Data manipulation
 □ string - String operations
 □ stopwords - Common word filter
 □ joblib - Model serialization
 □ LogisticRegression - Binary classifier
 □ train_test_split - Data splitting
 □ accuracy_score - Performance metric
 □ TfidfVectorizer - Text vectorization

```
import nltk
nltk.download("popular")
from sklearn.metrics.pairwise import cosine_similarity
import pandas as pd
import string
from nltk.corpus import stopwords
import joblib
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, classification_report,confusion_matrix
from sklearn.feature_extraction.text import TfidfVectorizer
```

Load Dataset:



Clean Text:

The **preprocess_text** function cleans text by removing punctuation, converting it to lowercase, and removing stopwords. It is applied to the **source_text** and **plagiarized_text** columns to prepare data for analysis.

Unnamed: 0 researchers discovered new species butterfly a 1 1 moon orbits earth approximately 273 days scientists found previously unknown butterfly 2 2 water composed two hydrogen atoms one oxygen atom 3 3 history rome dates back 753 bc 4 4 4 pluto considered ninth planet solar system					
1 1 moon orbits earth approximately 273 days natural satellite takes around 273 days comple 1 2 vater composed two hydrogen atoms one oxygen atom h2o consists 2 hydrogen atoms 1 oxygen atom 1 3 history rome dates back 753 bc rome long history traced back 753 bc past pluto classified ninth planet suns planet 1	label	plagiarized_text	source_text	Unnamed: 0	
2 2 water composed two hydrogen atoms one oxygen atom 3 3 history rome dates back 753 bc rome long history traced back 753 bc 4 4 pluto considered ninth planet solar system 5 397 playing musical instruments enhances creativity 6 398 studying history helps understanding present understanding present aided studying history 7 367 399 listening classical music improve focus focus improved listening classical music 7 368 400 practicing yoga enhances physical flexibility physical flexibility enhanced practicing yoga 7 369 401 volunteering fosters community spirit community spirit fostered volunteering	1	scientists found previously unknown butterfly	researchers discovered new species butterfly a	0	0
3 history rome dates back 753 bc rome long history traced back 753 bc 1 4 4 pluto considered ninth planet solar system past pluto classified ninth planet suns planet 1	1	natural satellite takes around 273 days comple	moon orbits earth approximately 273 days	1	1
4 4 pluto considered ninth planet solar system past pluto classified ninth planet suns planet 1	1	h2o consists 2 hydrogen atoms 1 oxygen atom	water composed two hydrogen atoms one oxygen atom	2	2
	1	rome long history traced back 753 bc	history rome dates back 753 bc	3	3
playing musical instruments enhances creativity creativity enhanced playing musical instruments 0 studying history helps understanding present understanding present aided studying history 0 listening classical music improve focus focus improved listening classical music 0 practicing yoga enhances physical flexibility physical flexibility enhanced practicing yoga 0 volunteering fosters community spirit community spirit fostered volunteering 0	1	past pluto classified ninth planet suns planet	pluto considered ninth planet solar system	4	4
366 398 studying history helps understanding present understanding present aided studying history 0 367 399 listening classical music improve focus focus improved listening classical music 0 368 400 practicing yoga enhances physical flexibility physical flexibility enhanced practicing yoga 0 369 401 volunteering fosters community spirit community spirit fostered volunteering 0					
367 399 listening classical music improve focus focus improved listening classical music 0 368 400 practicing yoga enhances physical flexibility physical flexibility enhanced practicing yoga 0 369 401 volunteering fosters community spirit community spirit fostered volunteering 0	0	creativity enhanced playing musical instruments	playing musical instruments enhances creativity	397	365
368 400 practicing yoga enhances physical flexibility physical flexibility enhanced practicing yoga 0 369 401 volunteering fosters community spirit community spirit fostered volunteering 0	0	understanding present aided studying history	studying history helps understanding present	398	366
369 401 volunteering fosters community spirit community spirit fostered volunteering 0	0	focus improved listening classical music	listening classical music improve focus	399	367
	0	physical flexibility enhanced practicing yoga	practicing yoga enhances physical flexibility	400	368
370 rows × 4 columns	0	community spirit fostered volunteering	volunteering fosters community spirit	401	369
			s	ws × 4 column	370 rc

Train Test Split:

This code trains a Logistic Regression model using X_{train} and y_{train} , then predicts labels for X_{test} . It evaluates the model's performance with accuracy, a classification report (precision, recall, F1-score), and a confusion matrix.

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

[30]

Applying logisticRegression
```

```
model = LogisticRegression()
   model.fit(X_train,y_train)
   y_pred = model.predict(X_test)
   print("Accuracy:", accuracy score(y test,y pred))
   print("Classification ",classification_report(y_test,y_pred))
   print("Confusion ", confusion_matrix(y_test,y_pred))
Accuracy: 0.8243243243243243
Classification
                             precision
                                          recall f1-score
                                                             support
           0
                  0.79
                            0.86
                                      0.82
                                                  35
           1
                  0.86
                            0.79
                                      0.83
                                                  39
   accuracy
                                      0.82
                                                  74
  macro avg
                  0.83
                            0.83
                                      0.82
                                                  74
weighted avg
                  0.83
                            0.82
                                      0.82
                                                  74
Confusion [[30 5]
[ 8 31]]
```

Random Forest Model:

This code uses a Random Forest Classifier to train on X_train and y_train, predict labels for X_test, and evaluate the model. It calculates accuracy, generates a classification report (precision, recall, F1-score), and displays a confusion matrix to analyze performance.

```
from sklearn.ensemble import RandomForestClassifier
# Instantiate the model
model = RandomForestClassifier(n_estimators=100, random_state=42)
# Fit the model
model.fit(X_train, y_train)
# Make predictions
y_pred = model.predict(X_test)
# Calculate accuracy
accuracy = accuracy_score(y_test, y_pred)
# Generate classification report
classification_rep = classification_report(y_test, y_pred)
# Generate confusion matrix
cm = confusion_matrix(y_test, y_pred)
# Print results
print("Accuracy:", accuracy)
print("Classification_rep)
print(classification_rep)
print("Confusion Matrix:")
print(cm)
```

```
Accuracy: 0.7972972972973
Classification Report:
             precision recall f1-score
                                            support
          0
                           0.97
                                                 35
                  0.71
                                     0.82
                                                 39
                  0.96
                           0.64
                                     0.77
                                                 74
   accuracy
                                     0.80
  macro avg
                  0.83
                           0.81
                                     0.79
                                                 74
weighted avg
                 0.84
                           0.80
                                     0.79
                                                 74
Confusion Matrix:
[[34 1]
[14 25]]
```

```
from sklearn.ensemble import RandomForestClassifier
   model = RandomForestClassifier(n_estimators=100, random_state=42)
   # Fit the model
   model.fit(X_train, y_train)
   y_pred = model.predict(X_test)
   print("Accuracy:", accuracy_score(y_test,y_pred))
   print("Classification ",classification_report(y_test,y_pred))
   print("Confusion ", confusion_matrix(y_test,y_pred))
Accuracy: 0.7972972972972973
Classification
                              precision
                                          recall f1-score
                                                              support
           0
                  0.71
                             0.97
                                       0.82
                  0.96
                             0.64
                                       0.77
                                                   39
                                                   74
   accuracy
                                       0.80
   macro avg
                  0.83
                             0.81
                                       0.79
                                                   74
                  0.84
                             0.80
                                       0.79
                                                   74
weighted avg
Confusion [[34 1]
 [14 25]]
```

Naiv Bays Model:

This code implements the Multinomial Naive Bayes algorithm, training it on X_train and y_train and predicting X_test. It evaluates the model's performance using accuracy, a classification report, and a confusion matrix.

```
from sklearn.naive_bayes import MultinomialNB
   model = MultinomialNB()
   model.fit(X_train, y_train)
   # Make predictions
   y_pred = model.predict(X_test)
   print("Accuracy:", accuracy_score(y_test,y_pred))
   print("Classification ",classification_report(y_test,y_pred))
   print("Confusion ", confusion_matrix(y_test,y_pred))
Accuracy: 0.8648648648648649
Classification
                             precision
                                          recall f1-score
                                                             support
          a
                  0.86
                            0.86
                                      0.86
                                                  35
                  0.87
                            0.87
                                      0.87
                                                  39
                                      0.86
                                                  74
   accuracy
  macro avg
                  0.86
                            0.86
                                      0.86
                                                  74
weighted avg
                  0.86
                            0.86
                                      0.86
                                                  74
Confusion [[30 5]
[ 5 34]]
```

SVM:

This code employs a Support Vector Machine (SVM) with a linear kernel to classify data. It trains the model on X_train and y_train, predicts labels for X_test, and evaluates performance using accuracy, a classification report, and a confusion matrix.

```
# Instantiate the model
model = SVC(kernel='linear', random_state=42)

model.fit(X_train, y_train)
# Make predictions
y_pred = model.predict(X_test)

print("Accuracy:", accuracy_score(y_test,y_pred))
print("Classification ",classification_report(y_test,y_pred))
print("Confusion ", confusion_matrix(y_test,y_pred))
```

Superior university

```
Accuracy: 0.8783783783783784
Classification
                              precision
                                           recall f1-score
                                                              support
           0
                   0.86
                             0.89
                                       0.87
                                                   35
           1
                   0.89
                             0.87
                                       0.88
                                                   39
                                                   74
    accuracy
                                       0.88
   macro avg
                             0.88
                                       0.88
                                                   74
                   0.88
weighted avg
                   0.88
                             0.88
                                       0.88
                                                   74
Confusion [[31 4]
 [ 5 34]]
```

Save SVM and Vectorizer

```
import pickle

pickle.dump(model,open("model.pkl",'wb'))

pickle.dump(tfidf_vectorizer, open('tfidf_vectorizer.pkl','wb'))
```

Load Model And Vectorizer:

```
model = pickle.load(open('model.pkl','rb'))
tfidf_vectorizer = pickle.load(open('tfidf_vectorizer.pkl','rb'))
```

Detection System:

```
def detect(input_text):
    # vectorize text
    vectorized_text = tfidf_vectorizer.transform([input_text])
    # thun will do prediction by model
    result = model.predict(vectorized_text)
    return "Plagiarim Detected" if result[0] == 1 else "No Plagiarism"

# example ( it is a plagarized text)
    input_text = "Researchers have discovered a new species of butterfly in the Amazon rainforest."

# example ( it has no plagiarism)
    input_text = "Playing musical instruments enhances creativity."

detect(input_text)

# example ( it has no plagarism)
    input_text = "Practicing yoga enhances physical flexibility."
    detect(input_text)
```

```
# example ( it has no plagarism)
input_text = 'Practicing yoga enhances physical flexibility.'
detect(input_text)

'No Plagiarism'

# sklearn version
import sklearn
sklearn.__version__
'1.3.2'
```

IMPLEMENTATION DETAILS

1. Backend (app.py)

The core logic for the application resides in **app.py**. Key features include:

• Text Input Handling:

Users provide input via a form on the webpage.

Plagiarism Detection:

- The input text is transformed using the TF-IDF vectorizer.
- The vectorized data is passed to the pre-trained model to classify the text as "Plagiarism Detected" or "No Plagiarism."

Error Handling:

The application manages errors gracefully, such as missing form data, ensuring a smooth user experience.

```
vdef detect(input_text):
    vectorized_text = tfidf_vectorizer.transform([input_text])
    result = model.predict(vectorized_text)
    return "Plagiarism Detected" if result[0] == 1 else "No Plagiarism"
```

2. Web Interface

- Home Page: Displays a form where users can input text for analysis.
- **Results Page:** Shows the analysis results, indicating whether plagiarism was detected.

```
@app.route('/')
def home():
    return render_template('index.html')
```

3. Machine Learning Components

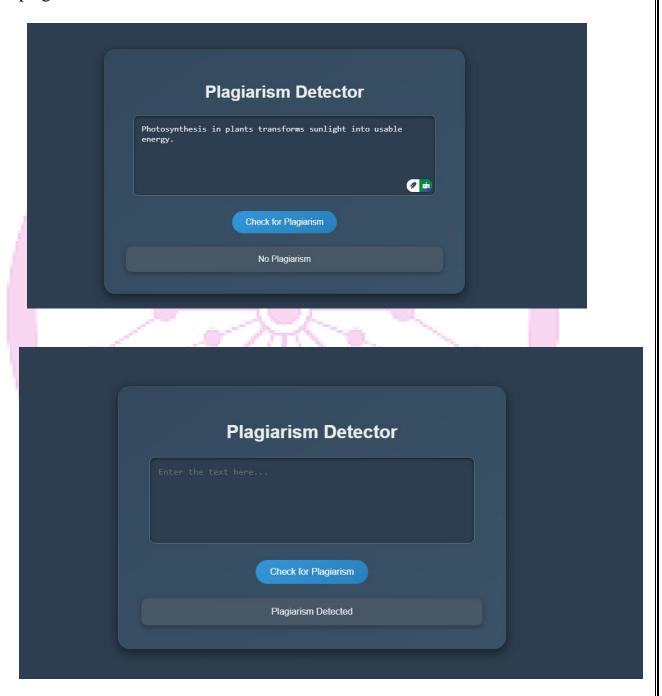
- Model (model.pkl): A trained classification model capable of identifying patterns indicative of plagiarism.
- TF-IDF Vectorizer (tfidf_vectorizer.pkl): Converts textual input into numerical form, enabling machine learning predictions.

Conclusion:

The Plagiarism Checker is a functional web-based application that efficiently detects plagiarism using machine learning. With its scalable architecture and intuitive interface, it serves as a foundational tool for text analysis and is well-suited for integration into educational and professional environments.

HOME PAGE

In this text "Photosynthesis in plants transforms sunlight into usable energy" plagiarism has been detected.



No Plagiarism



