



Model Development Phase Template

Date	1 November 2024
Team ID	SWTID1726834817
Project Title	Fake News Analysis in social media using NLP
Maximum Marks	4 Marks

Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

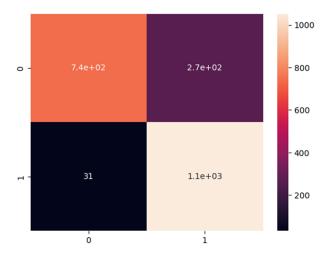
Initial Model Training Code:

```
import pandas as pd
   import numpy as np
   from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
   from sklearn.model_selection import train_test_split
   from sklearn.naive bayes import MultinomialNB
   from sklearn.metrics import accuracy score, confusion matrix
   import seaborn as sns
   import pickle
df = pd.read_csv('news.csv') # Replace with your dataset path
X = df['text'] # Features (text data)
y = df['label'] # Labels (FAKE or REAL)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33,
random state=53)
count_vectorizer = CountVectorizer(stop_words='english')
count_train = count_vectorizer.fit_transform(X_train) # Fit and transform
trainina data
count_test = count_vectorizer.transform(X_test) # Transform test data
tfidf_vectorizer = TfidfVectorizer(stop_words='english', max_df=0.7)
tfidf_train = tfidf_vectorizer.fit_transform(X_train) # Fit and transform
training data
tfidf_test = tfidf_vectorizer.transform(X_test) # Transform test data
nb_classifier = MultinomialNB()
nb_classifier.fit(count_train, y_train) # Fit the model
pred = nb_classifier.predict(count_test) # Predict on test data
nb_classifier.fit(tfidf_train, y_train) # Fit the model
pred = nb_classifier.predict(tfidf_test) # Predict on test data
score = accuracy_score(y_test, pred) # Calculate accuracy
print(score)
0.8565279770444764
cm = confusion_matrix(y_test, pred, labels=['FAKE', 'REAL']) # Confusion
sns.heatmap(cm, annot=True) # Visualizing the confusion matrix
```





<Axes: >



Model Validation and Evaluation Report:

Classification Report	Accuracy	Confusion Matrix
ny the confusion matrix confusion_matrix(y_test, rf_pred, labels=['FAKE', 'REAL']) confusion Matrix:\n", rf_cm) Cate the accuracy scare re = accuracy_score(y_test, rf_pred) Random Forest Model Accuracy:". rf score)	Accuracy: 0.9076996652319465	Confusion Metrix: [[931 77] [116 967]]
	y the confision motrix confusion_matrix(y_test, rf_pred, labels=['FAKE', 'REAL']) confusion Matrix:\n", rf_cm) Late the occuracy scare	y the confusion matrix confusion_matrix(y_test, rf_pred, labels=['FAKE', 'REAL']) confusion_Matrix(\n", rf_cm) Accuracy: 0.9076996652319465 Late the accuracy_score e = accuracy_score(y_test, rf_pred)





Decision Tree	<pre>dt_score = accuracy_score(y_test, dt_pred) print("Decision Tree Model Accuracy:", dt_score) # Display the confusion matrix dt_cm = confusion_matrix(y_test, dt_pred, labels=['FAKE', 'REAL']) print("Confusion Matrix:\n", dt_cm)</pre>	Accuracy: 0.8106169296987088	Confusion Matrix: [[810 198] [198 885]]
KNN	<pre>knn_score = accuracy_score(y_test, knn_pred) print("KNN Model Accuracy:", knn_score) knn_param_grid = { 'n_neighbors': [3, 5, 7, 9, 11], 'weights': ['uniform', 'distance'], 'metric': ['euclidean', 'manhattan'] } # Initialize the KNN classifier knn_classifier = KNeighborsClassifier() # Display the confusion matrix knn_cm = confusion_matrix(y_test, knn_pred, labels=['FAKE', 'REAL']) print("Confusion Matrix:\n", knn_cm)</pre>	Accuracy: 0.5442372070779531	Confusion Metrix: [[1006 2] [951 132]]
Gradient Boosting	<pre># Display the confusion motrix gb_cm = confusion_matrix(y_test, gb_pred, labels=['FAKE', 'REAL']) print("Confusion Matrix:\n", gb_cm) # Colculate the accuracy scare gb_score = accuracy_score(y_test, gb_pred) print("Gradient Boosting Model Accuracy:", gb_score)</pre>	Accuracy: 0.8919177427068389	Confusion Metrix: [[927 81] [145 938]]