```
from google.colab import drive
drive.mount('/content/drive')
```

→ Mounted at /content/drive

Import Essential Tools

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import re
import string
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, confusion_matrix
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
from nltk.stem import PorterStemmer
from nltk.tokenize import word_tokenize
import nltk
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('wordnet')
    [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk_data]
                  Package stopwords is already up-to-date!
    [nltk_data] Downloading package punkt to /root/nltk_data...
    [nltk_data]
                  Package punkt is already up-to-date!
    [nltk_data] Downloading package wordnet to /root/nltk_data...
                 Package wordnet is already up-to-date!
    [nltk_data]
```

Text Classification using Machine Learning Models

Instructions: Trump Tweet Sentiment Classification

1. Load the Dataset

Load the dataset named "trump_tweet_sentiment_analysis.csv" using pandas. Ensure the dataset contains at least two columns: "text" and "label".

2. Text Cleaning and Tokenization

Apply a text preprocessing pipeline to the "text" column. This should include:

- · Lowercasing the text
- o Removing URLs, mentions, punctuation, and special characters
- Removing stopwords
- Tokenization (optional: stemming or lemmatization)
- o "Complete the above function"

3. Train-Test Split

Split the cleaned and tokenized dataset into training and testing sets using train_test_split from sklearn.model_selection.

4. TF-IDF Vectorization

Import and use the TfidfVectorizer from sklearn.feature_extraction.text to transform the training and testing texts into numerical feature vectors.

5. Model Training and Evaluation

Import **Logistic Regression** (or any machine learning model of your choice) from sklearn.linear_model. Train it on the TF-IDF-embedded training data, then evaluate it using the test set.

• Print the **classification report** using classification_report from sklearn.metrics.

Text Classification Exercise

Load Dataset

Train Test Split

```
df = pd.read_csv("/content/drive/MyDrive/AI and ML/Week8/trum_tweet_sentiment_analysis.csv")
df.columns
→ Index(['text', 'Sentiment'], dtype='object')
assert 'text' in df.columns and 'Sentiment' in df.columns, "Dataset must contain 'text' and 'sentiment' columns."

    Cleaning and Tokenization

→ Helper Functions

def lower_case(text):
  return text.lower()
def remove_url(text):
  return re.sub(r"http\S+|www\S+|https\S+", '', text, flags=re.MULTILINE)
def remove_mentions(text):
  return re.sub(r'@\w+', '', text)
def remove_punctuations(text):
  return text.translate(str.maketrans('', '', string.punctuation))
def remove_stopwords(tokens):
  stop_words = set(stopwords.words('english'))
  tokens = [word for word in tokens if word not in stop_words and word.isalpha()]
  return tokens
def lemmatize_words(tokens):
  lemmatizer = WordNetLemmatizer()
  tokens = [lemmatizer.lemmatize(token) for token in tokens]
  return tokens
def stemm_words(text):
  porter = PorterStemmer()
  stemm_tokens = []
  for word in text:
    stemm_tokens.append(porter.stem(word))
  return stemm_tokens

→ Build a Text Cleaning Pipeline

def text_cleaning_pipeline(text, rule = "lemmatize"):
  text = lower_case(text)
  text = remove_url(text)
  text = remove_mentions(text)
  text = remove_punctuations(text)
  tokens = word_tokenize(text)
  tokens = remove_stopwords(tokens)
  tokens = lemmatize_words(tokens)
  return " ".join(tokens)
df['clean_text'] = df['text'].apply(text_cleaning_pipeline)
```

X_train, X_test, y_train, y_test = train_test_split(df['clean_text'], df['Sentiment'], test_size=0.2, random_state=42, strat

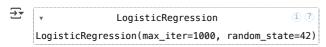
→ TF-IDF Vectorization

```
vectorizer = TfidfVectorizer(max_features=5000)
X_train_tfidf = vectorizer.fit_transform(X_train)
X_test_tfidf = vectorizer.transform(X_test)
```

Model Training and Evaluation

Model Training

```
\label{eq:model} \begin{tabular}{ll} model = LogisticRegression(max\_iter=1000, random\_state=42) \\ model.fit(X\_train\_tfidf, y\_train) \\ \end{tabular}
```



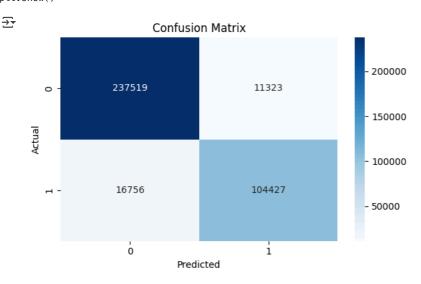
→ Evaluation

```
y_pred = model.predict(X_test_tfidf)
print("Classification Report:\n")
print(classification_report(y_test, y_pred))
```

→ Classification Report:

	precision	recall	f1-score	support
0 1	0.93 0.90	0.95 0.86	0.94 0.88	248842 121183
accuracy macro avg weighted avg	0.92 0.92	0.91 0.92	0.92 0.91 0.92	370025 370025 370025

```
cm = confusion_matrix(y_test, y_pred, labels=model.classes_)
plt.figure(figsize=(6, 4))
sns.heatmap(cm, annot=True, fmt='d', cmap="Blues", xticklabels=model.classes_, yticklabels=model.classes_)
plt.title("Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.tight_layout()
plt.show()
```



```
pred_df = pd.DataFrame({
    'Cleaned Text': X_test,
    'Actual Sentiment': y_test,
    'Predicted Sentiment': y_pred
})
pred_df.head()
```



	Cleaned Text	Actual Sentiment	Predicted Sentiment
1432084	rt maralago member pay trump hundred thousand	0	0
133054	rt seriously arkansas even trump know samesex	1	1
345307	rt bercow prefers north korea president trump	0	0
717727	rt breaking trump right look found raided mosq	1	1
741002	rt edited robocop trump speech actually make s	0	0