Importing libaries

```
import pandas as pd
    import matplotlib.pyplot as plt
    import seaborn as sns
    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
    import re
    import string
    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_tab')
    nltk.download('wordnet')

→ [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk_data] Unzipping corpora/stopwords.zip.
    [nltk_data] Downloading package punkt_tab to /root/nltk_data...
    [nltk_data] Unzipping tokenizers/punkt_tab.zip.
    [nltk_data] Downloading package wordnet to /root/nltk_data...
    True
```

Exercise

Load Dataset

```
[] df = pd.read_csv("/content/drive/MyDrive/Artificial intelligence and Machine learning/Week-8/Workshop/trum_tweet_sentiment_analysis.csv")

[] df.columns

Tindex(['text', 'Sentiment'], dtype='object')

[] assert 'text' in df.columns and 'Sentiment' in df.columns, "Dataset must contain 'text' and 'sentiment' columns."
```

Cleaning and Tokenization

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)

    Train Test Split

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

    Train Test Split

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

    TF-IDF Vectorization

[ ] vectorizer = TfidfVectorizer(max_features=5000)
    X_train_tfidf = vectorizer.fit_transform(X_train)
X_test_tfidf = vectorizer.transform(X_test)
Evaluation and Model Training

    Model Training

                                                                                                      + Code + Text
[ ] model = LogisticRegression(max_iter=1000, random_state=42)
     model.fit(X_train_tfidf, y_train)
                 LogisticRegression
    LogisticRegression(max_iter=1000, random_state=42)
```

Model Training

```
[ ] model = LogisticRegression(max_iter=1000, random_state=42)
model.fit(X_train_tfidf, y_train)
```

```
LogisticRegression LogisticRegression(max_iter=1000, random_state=42)
```

Evaluation

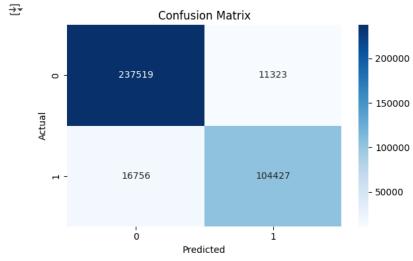
```
y_pred = model.predict(X_test_tfidf)

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()
```



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