```
import pandas as pd
      import numpy as np
      from sklearn.model_selection import train_test_split
  4
      from tensorflow.keras.models import Sequential
      from tensorflow.keras.layers import Embedding, LSTM, Dense, Dropout
      from tensorflow.keras.preprocessing.text import Tokenizer
      from tensorflow.keras.preprocessing.sequence import pad_sequences
      from sklearn.preprocessing import LabelEncoder
      from sklearn.metrics import classification report, confusion matrix
 10
      import matplotlib.pyplot as plt
 11
      import seaborn as sns
      df = pd.read_csv('/content/Twitter.csv', encoding='latin-1')
  2
      df.head()
₹
              Brand Sentiment
     0 Borderlands
                        Positive
                                  im getting on borderlands and i will murder yo..
                        Positive
                                    I am coming to the borders and I will kill you...
     2 Borderlands
                        Positive
                                     im getting on borderlands and i will kill you ...
                        Positive im coming on borderlands and i will murder you...
       Borderlands
                        Positive
                                   im getting on borderlands 2 and i will murder ...
Next steps:
             Generate code with df
                                     View recommended plots
                                                                    New interactive sheet
 1 # Preprocessing
 2 df = df[['Text', 'Brand', 'Sentiment']] # Select relevant columns
 3 df.dropna(inplace=True) # Remove rows with missing values
→ <ipython-input-80-af6ac83bd871>:3: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame
    See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.">https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.</a>
       df.dropna(inplace=True) # Remove rows with missing values
    4
 1 # Encode categorical features
 2 le brand = LabelEncoder()
 3 df['Brand'] = le_brand.fit_transform(df['Brand'])
 4 le_sentiment = LabelEncoder()
 5 df['Sentiment'] = le_sentiment.fit_transform(df['Sentiment'])
 1 # Tokenize text data
 2 max_words = 10000 # Adjust as needed
 3 tokenizer = Tokenizer(num words=max words)
 4 tokenizer.fit_on_texts(df['Text'])
 5 sequences = tokenizer.texts to sequences(df['Text'])
 1 # Pad sequences to ensure uniform length
 2 max_sequence_length = 100  # Adjust as needed
 3 padded_sequences = pad_sequences(sequences, maxlen=max_sequence_length)
```

```
1 # Split data into training and testing sets
 2 X = np.array(padded_sequences)
 3 y = np.array(df['Sentiment'])
 4 brand = np.array(df['Brand']) # Include Brand as a feature
 5 X_train, X_test, y_train, y_test, brand_train, brand_test = train_test_split(X, y, brand, test_size=0.2
 1 # Model building
 2 model = Sequential()
 3 model.add(Embedding(max_words, 128, input_length=max_sequence_length))
 4 model.add(LSTM(64, dropout=0.2, recurrent_dropout=0.2)) #added dropout for regularization
 5 model.add(Dense(64, activation='relu')) #added a dense layer
 6 # Change activation to 'softmax' for multi-class classification
 7 model.add(Dense(4, activation='softmax')) # Assuming 4 sentiment classes (0, 1, 2, 3) Adjust if needed
/ usr/local/lib/python3.11/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument
      warnings.warn(
 1 # Compile the model with 'categorical_crossentropy' for multi-class
 2 model.compile(loss='sparse categorical crossentropy', optimizer='adam', metrics=['accuracy']) #changed
 1 model.fit(X_train, y_train, epochs=5, batch_size=64, validation_data=(X_test, y_test))
   Epoch 1/5
    925/925 -
                                - 162s 169ms/step - accuracy: 0.5099 - loss: 1.1197 - val_accuracy: 0.7305
    Epoch 2/5
    925/925 -
                                 203s 170ms/step - accuracy: 0.7791 - loss: 0.5889 - val_accuracy: 0.7857
    Epoch 3/5
    925/925 -
                                 202s 171ms/step - accuracy: 0.8435 - loss: 0.4167 - val accuracy: 0.8134
    Epoch 4/5
    925/925 -
                                 200s 169ms/step - accuracy: 0.8787 - loss: 0.3180 - val_accuracy: 0.8259
    Epoch 5/5
    925/925 -
                                - 200s 166ms/step - accuracy: 0.8982 - loss: 0.2661 - val accuracy: 0.8347
    <keras.src.callbacks.history.History at 0x795b2b3a69d0>
 1 loss, accuracy = model.evaluate(X_test, y_test)
 2 print("Test Loss:", loss)
 3 print("Test Accuracy:", accuracy)
<del>→</del> 463/463 -
                                - 11s 24ms/step - accuracy: 0.8312 - loss: 0.5101
    Test Loss: 0.4956231117248535
    Test Accuracy: 0.8347297310829163
 1 # Predict on the test set
 2 y_pred = model.predict(X_test)
 3 y_pred_classes = np.argmax(y_pred, axis=1) # Convert probabilities to class labels
   463/463
                                 12s 22ms/step
```

```
1 # Classification Report
2 print(classification_report(y_test, y_pred_classes))
```

```
₹
                   precision
                                recall f1-score
                                                    support
                0
                        0.19
                                  0.28
                                             0.23
                                                       2696
                        0.33
                                  0.14
                                             0.20
                                                       4380
                        0.22
                                             0.29
                                                       3605
                                  0.40
                        0.25
                                  0.15
                                             0.19
                                                       4119
         accuracy
                                             0.23
                                                      14800
       macro avg
                        0.25
                                  0.24
                                             0.23
                                                      14800
     weighted avg
                        0.26
                                  0.23
                                             0.22
                                                      14800
```

