```
import pandas as pd
import numpy as np
import re
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Embedding, LSTM, Dense, Dropout
from tensorflow.keras.callbacks import EarlyStopping
from sklearn.metrics import classification_report, confusion_matrix
import matplotlib.pyplot as plt
def generate_fake_news_dataset(n_samples=1000):
    fake_patterns = [
        "A shocking revelation that will change everything",
       "Unbelievable truth hidden from the public",
        "Celebrities revealing underground networks",
        "Breaking news that mainstream media won't tell you"
   ]
   real_patterns = [
        "Scientific research confirms new findings",
        "Expert analysis provides insights",
       "Recent study highlights important trends",
       "Comprehensive report examines critical issues",
        "Researchers uncover significant data"
   ]
   data = {
        'text': [],
        'label': []
   for _ in range(n_samples//2):
       data['text'].append(np.random.choice(fake_patterns) + " " +
                              '.join(np.random.choice(['dramatic', 'incredible', 'unbelievable']) for _ in range(5)))
       data['label'].append(1)
       data['text'].append(np.random.choice(real_patterns) + " " +
                               '.join(np.random.choice(['research', 'study', 'analysis']) for _ in range(5)))
       data['label'].append(0)
   df = pd.DataFrame(data)
   df.to_csv('fake_news_dataset.csv', index=False)
   return df
def preprocess text(text):
   text = text.lower()
   text = re.sub(r'[^a-zA-Z\s]', '', text)
   text = re.sub(r'\s+', ' ', text).strip()
   return text
def main():
   df = generate fake news dataset(n samples=1000)
   print("Dataset Generated:")
   print(df['label'].value_counts())
   df['processed_text'] = df['text'].apply(preprocess_text)
   X_train, X_test, y_train, y_test = train_test_split(
```

```
df['processed_text'],
       df['label'],
       test_size=0.2,
       random_state=42
   tokenizer = Tokenizer(num_words=5000)
   tokenizer.fit_on_texts(X_train)
   X_train_seq = tokenizer.texts_to_sequences(X_train)
   X_test_seq = tokenizer.texts_to_sequences(X_test)
   max\_length = 100
   X_train_pad = pad_sequences(X_train_seq, maxlen=max_length, padding='post')
   X_test_pad = pad_sequences(X_test_seq, maxlen=max_length, padding='post')
   model = Sequential([
       Embedding(5000, 128, input_length=max_length),
       LSTM(128, dropout=0.2),
       Dense(64, activation='relu'),
       Dropout(0.5),
       Dense(1, activation='sigmoid')
   ])
   model.compile(
       optimizer='adam',
       loss='binary_crossentropy',
       metrics=['accuracy']
   )
   early_stopping = EarlyStopping(
       monitor='val_loss',
       patience=5,
       restore_best_weights=True
   history = model.fit(
       X_train_pad, y_train,
       epochs=20,
       batch_size=32,
       validation_split=0.2,
       callbacks=[early_stopping]
   )
   y pred = model.predict(X test pad)
   y_pred_binary = (y_pred > 0.5).astype(int)
   print("\nClassification Report:")
   print(classification_report(y_test, y_pred_binary))
   cm = confusion_matrix(y_test, y_pred_binary)
   plt.figure(figsize=(8, 6))
   plt.imshow(cm, interpolation='nearest', cmap='Blues')
   plt.title('Confusion Matrix')
   plt.colorbar()
   plt.xlabel('Predicted')
   plt.ylabel('Actual')
   plt.show()
if __name__ == "__main__":
   main()
```

```
Dataset Generated: label
```

1 500

0 500

Name: count, dtype: int64

/usr/local/lib/python3.10/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input\_length` is deprecated. Just warnings.warn(

Epoch 1/20 20/20 **- 7s** 182ms/step - accuracy: 0.5008 - loss: 0.6933 - val\_accuracy: 0.5188 - val\_loss: 0.6924 Epoch 2/20 20/20 **– 5s** 276ms/step - accuracy: 0.4822 - loss: 0.6961 - val\_accuracy: 0.5188 - val\_loss: 0.6929 Epoch 3/20 20/20 **– 3s** 169ms/step - accuracy: 0.4885 - loss: 0.6937 - val accuracy: 0.4812 - val loss: 0.6934 Epoch 4/20 20/20 **– 5s** 171ms/step - accuracy: 0.5104 - loss: 0.6939 - val\_accuracy: 0.5188 - val\_loss: 0.6930 Epoch 5/20 20/20 **- 7s** 249ms/step - accuracy: 0.4997 - loss: 0.6938 - val\_accuracy: 0.4812 - val\_loss: 0.6935 Epoch 6/20 20/20 -**- 3s** 172ms/step - accuracy: 0.5062 - loss: 0.6940 - val\_accuracy: 0.4812 - val\_loss: 0.6932 7/7 -**- 1s** 77ms/step

## Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	103
1	0.48	1.00	0.65	97
accuracy			0.48	200
macro avg	0.24	0.50	0.33	200
weighted avg	0.24	0.48	0.32	200

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1531: UndefinedMetricWarning: Precision is ill-defined and be \_warn\_prf(average, modifier, f"{metric.capitalize()} is", len(result))

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