

# Practical 02

```
1 import tensorflow as tf
2 from tensorflow.keras.preprocessing.text import Tokenizer
3 from tensorflow.keras.preprocessing.sequence import pad_sequences
4 from tensorflow.keras.models import Sequential
5 from tensorflow.keras.layers import Embedding, LSTM, Bidirectional, Dense, Dropout
6 from tensorflow.keras.datasets import imdb
7 from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
8 import matplotlib.pyplot as plt
9 import numpy as np
10
11 vocab_size = 10000
12 max_length = 250
13 (X_train, y_train), (X_test, y_test) = imdb.load_data(num_words=vocab_size)
14 X_train = pad_sequences(X_train, maxlen=max_length, padding='post')
15 X_test = pad_sequences(X_test, maxlen=max_length, padding='post')
16 model = Sequential([
17     Embedding(vocab_size, 256, input_length=max_length),
18     Bidirectional(LSTM(128, return_sequences=True)),
19     LSTM(64),
20     Dense(64, activation='relu'),
21     Dropout(0.5),
22     Dense(1, activation='sigmoid')
23 ])
24 model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=0.0005),
25               loss='binary_crossentropy',
26               metrics=['accuracy'])
27 history = model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10,
28                    batch_size=64, verbose=1)
29 y_pred = model.predict(X_test)
30 y_pred = (y_pred > 0.5).astype(int)
31 accuracy = accuracy_score(y_test, y_pred)
32 precision = precision_score(y_test, y_pred)
33 recall = recall_score(y_test, y_pred)
34 f1 = f1_score(y_test, y_pred)
35
36 print(f"\n◆ Accuracy: {accuracy:.4f}")
37 print(f"◆ Precision: {precision:.4f}")
38 print(f"◆ Recall: {recall:.4f}")
39 print(f"◆ F1 Score: {f1:.4f}")
40
41 plt.figure(figsize=(10,5))
42 plt.subplot(1,2,1)
43 plt.plot(history.history['accuracy'], label='Train Accuracy')
44 plt.plot(history.history['val_accuracy'], label='Test Accuracy')
45 plt.xlabel("Epochs")
46 plt.ylabel("Accuracy")
47 plt.legend()
48 plt.title("Training vs. Validation Accuracy")
49 plt.subplot(1,2,2)
50 plt.plot(history.history['loss'], label='Train Loss')
51 plt.plot(history.history['val_loss'], label='Test Loss')
52 plt.xlabel("Epochs")
53 plt.ylabel("Loss")
54 plt.legend()
55 plt.title("Training vs. Validation Loss")
56 plt.show()
```

# Output

```
391/391 120s 306ms/step - accuracy: 0.7171 - loss: 0.5302 - val_accuracy: 0.8347 - val_loss: 0.3833
Epoch 6/10
391/391 117s 298ms/step - accuracy: 0.8858 - loss: 0.2928 - val_accuracy: 0.8702 - val_loss: 0.3092
Epoch 7/10
391/391 116s 298ms/step - accuracy: 0.9326 - loss: 0.1936 - val_accuracy: 0.8756 - val_loss: 0.3117
Epoch 8/10
391/391 117s 300ms/step - accuracy: 0.9570 - loss: 0.1315 - val_accuracy: 0.8721 - val_loss: 0.3499
Epoch 9/10
391/391 118s 303ms/step - accuracy: 0.9732 - loss: 0.0958 - val_accuracy: 0.8694 - val_loss: 0.4027
Epoch 10/10
391/391 118s 303ms/step - accuracy: 0.9781 - loss: 0.0752 - val_accuracy: 0.8680 - val_loss: 0.4391
782/782 57s 72ms/step

◆ Accuracy: 0.8680
◆ Precision: 0.8529
◆ Recall: 0.8894
◆ F1 Score: 0.8707
2025-04-22 11:08:33.968 Python[12480:8538170] +[IMKClient subclass]: chose IMKClient_Modern
2025-04-22 11:08:33.968 Python[12480:8538170] +[IMKInputSession subclass]: chose IMKInputSession_Modern
```





