```
from keras.datasets import imdb
from keras.preprocessing.sequence import pad_sequences
import numpy as np
import matplotlib.pyplot as plt
# Load dataset
(vocab_size, maxlen) = (10000, 200)
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=vocab_size)
# Explore dataset
print("Number of training samples:", len(x_train))
print("Number of test samples:", len(x_test))
print("Average review length:", np.mean([len(i) for i in x_train]))
print("Label distribution:", np.unique(y_train, return_counts=True))
x_train = pad_sequences(x_train, maxlen=maxlen)
x_test = pad_sequences(x_test, maxlen=maxlen)
# Visualize padded vs. original
print("Original:", x_train[0][-20:])
print("Padded:", x_train[0])
Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/imdb.npz</a>
     17464789/17464789
                                            0s Ous/step
     Number of training samples: 25000
     Number of test samples: 25000
     Average review length: 238.71364
     Label distribution: (array([0, 1]), array([12500, 12500]))
                                                               16 4472 113 103
                                                                                   32
     Original: [ 65 16 38 1334 88 12
                                               16 283
                                                          5
        15 16 5345
                     19 178
                                32]
                5 25 100 43 838 112
                                              50 670
                                                         2
                                                              9
     Padded: [
                                                                  35 480 284
             4 172 112 167
       150
                               2 336 385
                                                     4 172 4536 1111
                                                39
       546 38 13 447
                             4 192
                                      50
                                           16
                                                 6 147 2025
                                                               19
                                                                    14
                                                                         22
        4 1920 4613
                      469
                                      71
                                                                         76
                             4
                                 22
                                           87
                                                12
                                                     16
                                                         43
                                                              530
                                                                    38
        15 13 1247
                           22
                                17
                                    515
                                           17
                                                12
                                                     16
                                                         626
                                                               18
                       8 316
                                 8 106
                                                 4 2223 5244
                                                                   480
                                                                         66
        62 386
                 12
                                            5
                                                               16
      3785
            33
                  4
                      130
                           12
                                 16
                                      38
                                          619
                                                     25
                                                        124
                                                               51
                                                                    36
                                                                        135
             25 1415
                       33
                            6
                                 22
                                      12
                                          215
                                                     77
                                                                        407
                             4 107
        16
            82
                                     117 5952
                                                15
                                                           4
                                                                     7
                                                                       3766
                   2
                        8
                                                    256
         5 723
                  36
                       71
                            43 530 476
                                          26
                                               400
                                                    317
                                                          46
      1029
            13 104
                       88
                            4 381
                                      15
                                          297
                                                98
                                                     32 2071
                                                               56
                                                                    26
                                                                        141
        6 194 7486
                       18
                             4 226
                                      22
                                           21
                                               134
                                                   476
                                                          26
                                                              480
                                                                     5
                                                                        144
        30 5535 18
                       51
                            36
                                28
                                     224
                                           92
                                                25
                                                   104
                                                          4
                                                              226
                                                                    65
                                                                         16
        38 1334
                  88
                       12
                            16 283
                                           16 4472 113 103
      5345
            19
                178
                       32]
from keras.models import Sequential
from keras.layers import Embedding, SimpleRNN, Dense
model = Sequential([
    Embedding(input_dim=10000, output_dim=32, input_length=200),
    SimpleRNN(32),
    Dense(1, activation='sigmoid')
])
model.compile(optimizer='rmsprop', loss='binary_crossentropy', metrics=['accuracy'])
model.summary()
    /usr/local/lib/python3.11/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input_length` is deprecated. Just
       warnings.warn(
```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	?	0 (unbuilt)
simple_rnn (SimpleRNN)	?	0 (unbuilt)
dense (Dense)	?	0 (unbuilt)

Total params: 0 (0.00 B) Trainable params: 0 (0.00 B) Non-trainable params: 0 (0.00 B)

```
\label{eq:history} \mbox{ = model.fit} (\mbox{x\_train, y\_train, epochs=5, batch\_size=128, validation\_split=0.2)}
# Plot accuracy and loss
plt.plot(history.history['accuracy'], label='train acc')
plt.plot(history.history['val_accuracy'], label='val acc')
plt.title('Accuracy')
plt.legend()
plt.show()
plt.plot(history.history['loss'], label='train loss')
plt.plot(history.history['val_loss'], label='val loss')
plt.title('Loss')
plt.legend()
plt.show()
# Evaluate
test_loss, test_acc = model.evaluate(x_test, y_test)
print("Test Accuracy:", test_acc)

→ Epoch 1/5

     157/157 -
                                   13s 71ms/step - accuracy: 0.5949 - loss: 0.6456 - val_accuracy: 0.8266 - val_loss: 0.4102
     Epoch 2/5
     157/157
                                   19s 59ms/step - accuracy: 0.8201 - loss: 0.4093 - val_accuracy: 0.8160 - val_loss: 0.4006
     Epoch 3/5
     157/157 -
                                   10s 55ms/step - accuracy: 0.8706 - loss: 0.3207 - val_accuracy: 0.8628 - val_loss: 0.3399
     Epoch 4/5
     157/157 -
                                   11s 61ms/step - accuracy: 0.8930 - loss: 0.2710 - val_accuracy: 0.7684 - val_loss: 0.4727
     Epoch 5/5
     157/157 ·
                                    11s 64ms/step - accuracy: 0.9095 - loss: 0.2331 - val_accuracy: 0.8308 - val_loss: 0.3813
                                         Accuracy
                   train acc
      0.90
                   val acc
      0.85
      0.80
      0.75
      0.70
              0.0
                     0.5
                             1.0
                                     1.5
                                            2.0
                                                    2.5
                                                            3.0
                                                                   3.5
                                                                           4.0
                                           Loss
                                                                     train loss
      0.55
                                                                     val loss
      0.50
      0.45
       0.40
       0.35
       0.30
       0.25
                     0.5
                                     1.5
                                                                   3.5
              0.0
                             1.0
                                            2.0
                                                    2.5
                                                            3.0
                                                                           4.0
                                    9s 11ms/step - accuracy: 0.8371 - loss: 0.3845
     Test Accuracy: 0.8365600109100342
```

```
SimpleRNN(64)

Try <SimpleRNN name=simple_rnn_1, built=False>

from keras.layers import LSTM

model = Sequential([
    Embedding(input_dim=10000, output_dim=32, input_length=200),
    LSTM(32),
    Dense(1, activation='sigmoid')
])

# Change vocab_size to 5000 or 2000 in imdb.load_data
from keras.layers import Dropout

model = Sequential([
    Embedding(input_dim=10000, output_dim=32, input_length=200),
    SimpleRNN(32, dropout=0.2),
    Dense(1, activation='sigmoid')
])
```