```
In [4]: import numpy as np
       import pandas as pd
       import matplotlib.pyplot as plt
       import tensorflow as tf
       from tensorflow.keras.models import Sequential
       from tensorflow.keras.layers import LSTM, Dense, Dropout
       from sklearn.preprocessing import MinMaxScaler
In [5]: df=pd.read_csv("./airplane_passenger_prediction.csv")
In [6]: df
Out[6]:
           Month Passengers
        0 1949-01
                    112
        1 1949-02
                    118
        2 1949-03
                    132
        3 1949-04
                    129
        4 1949-05
                    121
       139 1960-08
                    606
       140 1960-09
                    508
       141 1960-10
                    461
       142 1960-11
                    390
       143 1960-12
                    432
      144 rows × 2 columns
In [7]: | df["Month"]=pd.to_datetime(df["Month"])
       df.set_index("Month",inplace=True)
       df
              Passengers
         Month
       1949-01-01
                   112
       1949-02-01
                   118
       1949-03-01
                   132
       1949-04-01
                   129
       1949-05-01
                   121
       1960-08-01
                   606
       1960-09-01
                   508
       1960-10-01
                   461
       1960-11-01
                   390
       1960-12-01
                   432
      144 rows × 1 columns
In [8]: scaler=MinMaxScaler(feature_range=(0,1))
       data_scaled=scaler.fit_transform(df)
       def create_sequence(data, seq_length):
        x, y=[],[]
        for i in range(len(data)-seq_length):
          x.append(data[i:(i+seq_length)])
          y.append(data[i+seq_length])
        return np.array(x),np.array(y)
In [11]: seq_length=5
       X,y=create_sequence(data_scaled,seq_length)
In [12]: #split data into train-test split
       size=int(0.8*len(X))
       X_train, X_test=X[:size], X[size:]
       y_train,y_test=y[:size],y[size:]
In [13]: #Define the LSTM model
       model=Sequential([LSTM(50, return_sequences=True, input_shape=(seq_length, 1)), LSTM(50), Dense(1)])
In [14]: model.compile(optimizer="adam", loss="mse")
       model.fit(X_train,y_train,epochs=50,batch_size=16,verbose=1)
       Epoch 1/50
       Epoch 2/50
       7/7 [=========== ] - 0s 16ms/step - loss: 0.0240
       7/7 [=========== ] - 0s 19ms/step - loss: 0.0179
       Epoch 4/50
       7/7 [========= ] - 0s 11ms/step - loss: 0.0144
       Epoch 5/50
       7/7 [========= ] - 0s 15ms/step - loss: 0.0118
       Epoch 6/50
       7/7 [=========== ] - 0s 12ms/step - loss: 0.0114
       Epoch 7/50
       7/7 [=========== ] - 0s 13ms/step - loss: 0.0100
       Epoch 8/50
       7/7 [=======] - 0s 18ms/step - loss: 0.0092
       Epoch 9/50
       7/7 [=======] - 0s 11ms/step - loss: 0.0089
       Epoch 10/50
       7/7 [========== ] - 0s 14ms/step - loss: 0.0087
       Epoch 13/50
       7/7 [========= ] - 0s 12ms/step - loss: 0.0086
       Epoch 14/50
       Epoch 15/50
       Epoch 16/50
       7/7 [============ ] - 0s 12ms/step - loss: 0.0084
       Epoch 17/50
       Epoch 18/50
       7/7 [======== ] - 0s 11ms/step - loss: 0.0083
       Epoch 19/50
       7/7 [======= ] - 0s 12ms/step - loss: 0.0083
       Epoch 20/50
       7/7 [========= ] - 0s 11ms/step - loss: 0.0083
       Epoch 21/50
       7/7 [======== ] - 0s 23ms/step - loss: 0.0084
       Epoch 22/50
       7/7 [============= ] - 0s 26ms/step - loss: 0.0084
       Epoch 23/50
       Epoch 24/50
       Epoch 25/50
       Epoch 26/50
       7/7 [=========] - 0s 10ms/step - loss: 0.0080
       Epoch 28/50
       7/7 [======= ] - 0s 10ms/step - loss: 0.0079
       Epoch 29/50
       7/7 [========= ] - 0s 10ms/step - loss: 0.0082
       Epoch 30/50
       7/7 [========== ] - 1s 111ms/step - loss: 0.0076
       Epoch 31/50
       7/7 [======== ] - 0s 13ms/step - loss: 0.0077
       Epoch 32/50
       7/7 [=======] - 0s 16ms/step - loss: 0.0076
       Epoch 33/50
       7/7 [======== ] - 0s 21ms/step - loss: 0.0075
       Epoch 34/50
       7/7 [======== ] - 0s 12ms/step - loss: 0.0076
       7/7 [========= ] - 0s 11ms/step - loss: 0.0076
       Epoch 36/50
       7/7 [============ ] - 0s 12ms/step - loss: 0.0073
       Epoch 37/50
       7/7 [============ ] - Os 11ms/step - loss: 0.0077
       Epoch 38/50
       7/7 [============= ] - 0s 11ms/step - loss: 0.0080
       Epoch 39/50
       7/7 [========= ] - 0s 11ms/step - loss: 0.0074
       Epoch 40/50
       7/7 [=======] - 0s 11ms/step - loss: 0.0072
       Epoch 41/50
       7/7 [======= ] - 0s 11ms/step - loss: 0.0070
       Epoch 42/50
       7/7 [========== ] - 0s 12ms/step - loss: 0.0077
       7/7 [========= ] - Os 14ms/step - loss: 0.0069
       7/7 [========= ] - 0s 12ms/step - loss: 0.0070
       Epoch 45/50
       7/7 [========= ] - 0s 11ms/step - loss: 0.0067
       Epoch 46/50
       7/7 [======== ] - 0s 11ms/step - loss: 0.0067
       Epoch 47/50
       Epoch 48/50
       7/7 [=========== ] - 0s 12ms/step - loss: 0.0064
       Epoch 49/50
       7/7 [========= ] - 0s 11ms/step - loss: 0.0065
       Epoch 50/50
       7/7 [========= ] - 0s 11ms/step - loss: 0.0063
       <keras.callbacks.History at 0x26de47674c0>
Out[14]:
In [15]: predictions=model.predict(X_test)
       1/1 [======] - 3s 3s/step
In [16]: predictions=scaler.inverse_transform(predictions)
In [17]: predictions
Out[17]: array([[434.537],
            [432.76926],
            [412.17795],
            [378.1969],
            [354.6024],
            [343.441 ],
            [339.3838],
            [353.66962],
            [368.94315],
            [384.16858],
            [407.48694],
            [446.86337],
            [478.76862],
            [476.536],
            [453.2577],
            [419.64987],
            [401.6198],
            [393.82507],
            [388.5291],
            [393.27304],
            [410.07288],
            [425.4471],
            [452.74762],
            [498.32596],
            [527.7141],
            [517.89636],
            [492.57117],
            [452.51154]], dtype=float32)
In [18]: y_test=scaler.inverse_transform(y_test)
In [20]: plt.figure(figsize=(10,5))
       plt.plot(df.index[-len(y_test):],y_test,label="Actual data")
       plt.plot(df.index[-len(y_test):],predictions,label="predicted data")
       plt.show()
                Actual data
                predicted data
       600
       550
       500
       450
       400
       350
       300
                                            1959-10
                                     1959-07
                                                   1960-01 1960-04
                                                                           1960-10
                                                                                   1961-01
              1958-10
                     1959-01 1959-04
                                                                   1960-07
```