Part 1: Importing Libraries and Loading Data

```
In [1]: # Importing necessary libraries
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import yfinance as yf
        from sklearn.preprocessing import MinMaxScaler
        # Downloading Apple's stock data
        apple_stock = yf.download('AAPL', start='2015-01-01', end='2024-07-01')
        # Display the first few rows of the dataset
        print("Initial dataset:")
        print(apple_stock.head())
        # Please check for missing values
        print("\nChecking for missing values:")
        print(apple_stock.isnull().sum())
      [******** 100%********* 1 of 1 completed
      Initial dataset:
                                 High
                                             Low
                                                     Close Adj Close
                                                                         Volume
                       0pen
      Date
      2015-01-02 27.847500 27.860001 26.837500 27.332500 24.402178 212818400
      2015-01-05 27.072500 27.162500 26.352501 26.562500 23.714725 257142000
      2015-01-06 26.635000 26.857500 26.157499 26.565001 23.716951 263188400
      2015-01-07 26.799999 27.049999 26.674999 26.937500 24.049515 160423600
      2015-01-08 27.307501 28.037500 27.174999 27.972500 24.973558 237458000
      Checking for missing values:
      0pen
                   0
      High
                   0
      Low
      Close
      Adj Close
      Volume
```

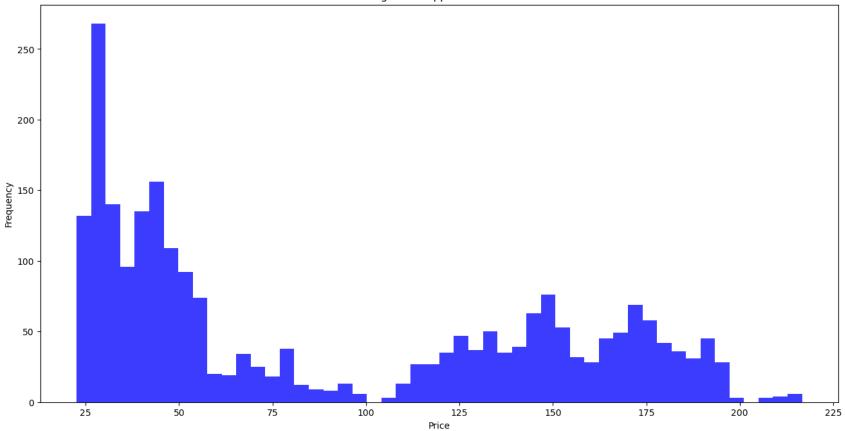
Part 2: Data Preprocessing

dtype: int64

```
In [2]: # Data preprocessing
        data = apple_stock['Close'].values
        data = data.reshape(-1, 1)
        scaler = MinMaxScaler(feature_range=(0, 1))
        scaled_data = scaler.fit_transform(data)
        # Splitting the data into training and testing sets
        train_data_len = int(np.ceil(len(scaled_data) * 0.8))
        train_data = scaled_data[0:train_data_len, :]
        test_data = scaled_data[train_data_len - 60:, :]
        # Creating training data set
        x_train, y_train = [], []
        for i in range(60, len(train_data)):
            x_train.append(train_data[i-60:i, 0])
            y_train.append(train_data[i, 0])
        x_train, y_train = np.array(x_train), np.array(y_train)
        # Reshaping the data for LSTM
        x_train = np.reshape(x_train, (x_train.shape[0], x_train.shape[1], 1))
        Part 3: Creating and Training the LSTM Model
In [3]: from keras.models import Sequential
        from keras.layers import Dense, LSTM
        # Building the LSTM model
        model = Sequential()
        model.add(LSTM(units=50, return_sequences=True, input_shape=(x_train.shape[1], 1)))
        model.add(LSTM(units=50, return_sequences=False))
        model.add(Dense(units=25))
        model.add(Dense(units=1))
        # Compiling the model
        model.compile(optimizer='adam', loss='mean_squared_error')
        # Training the model
        model.fit(x_train, y_train, batch_size=1, epochs=1)
```

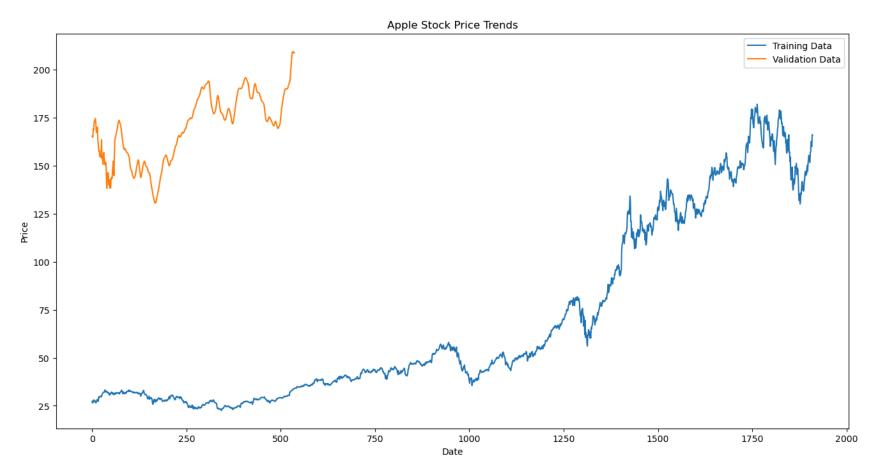
```
# Creating testing data set
        x_test, y_test = [], data[train_data_len:, :]
        for i in range(60, len(test data)):
            x_test.append(test_data[i-60:i, 0])
        x \text{ test} = np.array(x \text{ test})
        x_test = np.reshape(x_test, (x_test.shape[0], x_test.shape[1], 1))
        # Making predictions
        predictions = model.predict(x test)
        predictions = scaler.inverse_transform(predictions)
        # Preparing data for plotting
        train = data[:train data len]
        valid = data[train_data_len:]
        valid = np.append(valid[:60], predictions)
       C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\keras\src\layers\rnn\rnn.py:204: UserWarning: Do not pas
       s an `input shape`/`input dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` obje
       ct as the first layer in the model instead.
         super().__init__(**kwargs)
       1851/1851 ---
                                     - 22s 10ms/step - loss: 0.0039
       15/15 -
                                - 1s 27ms/step
In [ ]: Part 4: Histogram of Stock Prices
In [5]: # Plot 1: Histogram of stock prices
        plt.figure(figsize=(16, 8))
        plt.hist(data, bins=50, alpha=0.75, color='blue')
        plt.title('Histogram of Apple Stock Prices')
        plt.xlabel('Price')
        plt.ylabel('Frequency')
        plt.show()
```





In []: Part 5: Line Graph of Stock Trends

```
In [6]: # Plot 2: Line graph of stock trends
plt.figure(figsize=(16, 8))
plt.plot(train, label='Training Data')
plt.plot(valid, label='Validation Data')
plt.title('Apple Stock Price Trends')
plt.xlabel('Date')
plt.ylabel('Price')
plt.legend()
plt.show()
```



In []: Part 6: Line Graph of Stock Price Prediction

In [7]: # Plot 3: Line graph of stock price prediction
 plt.figure(figsize=(16, 8))
 plt.plot(valid[60:], label='Predictions', color='red')
 plt.plot(data[train_data_len:], label='Actual Price', color='blue')
 plt.title('Apple Stock Price Prediction')
 plt.xlabel('Date')
 plt.ylabel('Price')
 plt.legend()
 plt.show()





In []: