```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   %matplotlib inline
   import pandas_datareader as web
   import math
   from sklearn.preprocessing import MinMaxScaler
   from keras.models import Sequential
   from keras.layers import Dense, LSTM
   plt.style.use('fivethirtyeight')
```

In [2]: df = web.DataReader('AAPL', data_source = 'yahoo', start='2012-01-01', end='2019df

Out[2]:

	High	Low	Open	Close	Volume	Adj Close
Date						
2012-01-03	14.732143	14.607143	14.621429	14.686786	302220800.0	12.575914
2012-01-04	14.810000	14.617143	14.642857	14.765714	260022000.0	12.643497
2012-01-05	14.948214	14.738214	14.819643	14.929643	271269600.0	12.783868
2012-01-06	15.098214	14.972143	14.991786	15.085714	318292800.0	12.917509
2012-01-09	15.276786	15.048214	15.196429	15.061786	394024400.0	12.897017
2019-12-11	67.775002	67.125000	67.202499	67.692497	78756800.0	66.616837
2019-12-12	68.139999	66.830002	66.945000	67.864998	137310400.0	66.786591
2019-12-13	68.824997	67.732498	67.864998	68.787498	133587600.0	67.694443
2019-12-16	70.197502	69.245003	69.250000	69.964996	128186000.0	68.853218
2019-12-17	70.442497	69.699997	69.892502	70.102501	114158400.0	68.988541

2003 rows × 6 columns

```
In [3]: df.shape
```

Out[3]: (2003, 6)

```
In [4]: plt.figure(figsize=(16,8))
   plt.title('Close Price History')
   plt.plot(df['Close'])
   plt.xlabel('Date', fontsize=18)
   plt.ylabel('Close Price USD($)', fontsize=18)
```

Out[4]: Text(0, 0.5, 'Close Price USD(\$)')



```
In [5]: data = df.filter(['Close'])
    dataset = data.values
    training_data_len = math.ceil(len(dataset)* 0.8)
    training_data_len
```

Out[5]: 1603

```
In [8]: train data = scaled data[0:training data len,:]
        # Split the data into x train and y train data sets
        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])
            if i<= 61:
                print(x_train)
                print(y_train)
                print()
        [array([0.01316509, 0.01457063, 0.01748985, 0.02026915, 0.01984303,
               0.02080338, 0.02036454, 0.01962679, 0.01862191, 0.02173194,
               0.02453668, 0.02367172, 0.01893355, 0.02345548, 0.01900352,
               0.03569838, 0.03440732, 0.0360927, 0.03973694, 0.04194384,
               0.0417594 , 0.0410789 , 0.04397903, 0.04670744, 0.04979839,
               0.05479095, 0.0652785, 0.06543749, 0.07127594, 0.07563885,
               0.06814049, 0.07102789, 0.07097066, 0.07906688, 0.07791571,
               0.08004628, 0.08387497, 0.08600558, 0.09214292, 0.09661394,
               0.09790501, 0.09835659, 0.09071194, 0.08886753, 0.08914103,
               0.09632778, 0.09835024, 0.10269409, 0.11293358, 0.12659476,
               0.12403805, 0.1240444, 0.13392141, 0.13701237, 0.13481179,
               0.13280207, 0.13070964, 0.13766105, 0.14243103, 0.14442805])]
        [0.13949272033425864]
        [array([0.01316509, 0.01457063, 0.01748985, 0.02026915, 0.01984303,
               0.02080338, 0.02036454, 0.01962679, 0.01862191, 0.02173194,
               0.02453668, 0.02367172, 0.01893355, 0.02345548, 0.01900352,
               0.03569838, 0.03440732, 0.0360927, 0.03973694, 0.04194384,
               0.0417594 , 0.0410789 , 0.04397903, 0.04670744, 0.04979839,
               0.05479095, 0.0652785, 0.06543749, 0.07127594, 0.07563885,
               0.06814049, 0.07102789, 0.07097066, 0.07906688, 0.07791571,
               0.08004628, 0.08387497, 0.08600558, 0.09214292, 0.09661394,
               0.09790501, 0.09835659, 0.09071194, 0.08886753, 0.08914103,
               0.09632778, 0.09835024, 0.10269409, 0.11293358, 0.12659476,
               0.12403805, 0.1240444 , 0.13392141, 0.13701237, 0.13481179,
               0.13280207, 0.13070964, 0.13766105, 0.14243103, 0.14442805]), array([0.0
        1457063, 0.01748985, 0.02026915, 0.01984303, 0.02080338,
               0.02036454, 0.01962679, 0.01862191, 0.02173194, 0.02453668,
               0.02367172, 0.01893355, 0.02345548, 0.01900352, 0.03569838,
               0.03440732, 0.0360927, 0.03973694, 0.04194384, 0.0417594,
               0.0410789 , 0.04397903, 0.04670744, 0.04979839, 0.05479095,
               0.0652785 , 0.06543749 , 0.07127594 , 0.07563885 , 0.06814049 ,
               0.07102789, 0.07097066, 0.07906688, 0.07791571, 0.08004628,
               0.08387497, 0.08600558, 0.09214292, 0.09661394, 0.09790501,
               0.09835659, 0.09071194, 0.08886753, 0.08914103, 0.09632778,
               0.09835024, 0.10269409, 0.11293358, 0.12659476, 0.12403805,
               0.1240444 , 0.13392141, 0.13701237, 0.13481179, 0.13280207,
               0.13070964, 0.13766105, 0.14243103, 0.14442805, 0.13949272])]
        [0.13949272033425864, 0.13293562570222134]
```

```
In [9]: x train, y train = np.array(x train), np.array(y train)
In [13]: # Reshape the data
        x_train = np.reshape(x_train,(x_train.shape[0],x_train.shape[1],1))
        x train.shape
Out[13]: (1543, 60, 1)
In [15]: # build the LSTM model
        model = Sequential()
        model.add(LSTM(50, return_sequences=True, input_shape=(x_train.shape[1],1)))
        model.add(LSTM(50, return sequences=False))
        model.add(Dense(25))
        model.add(Dense(1))
In [16]: # compile the model
        model.compile(optimizer = 'adam', loss = 'mean_squared_error')
In [17]: |#traiin the model
        model.fit(x_train,y_train, batch_size=1, epochs= 1)
        Out[17]: <keras.callbacks.History at 0x29758c60d60>
In [26]: # create the testing data set
        # create a new array containing scaled values from index 1543 to 2003
        test data = scaled data[training data len-60: ,:]
        # create the data sets x_test and y_test
        x test = []
        y_test = dataset[training_data_len:,:]
        for i in range(60,len(test data)):
            x test.append(test data[i-60:i,0])
In [27]: # convert the data to a numpy araray
        x_test = np.array(x_test)
In [28]: x_test.shape
Out[28]: (400, 60)
In [29]: x test = np.reshape(x test, (x test.shape[0], x test.shape[1], 1))
```

```
In [30]: x_test.shape
Out[30]: (400, 60, 1)

In [31]: # get the models predicted price values
    predictions = model.predict(x_test)
    predictions = scaler.inverse_transform(predictions)

In [32]: #evaluate our model
    # get the root mean squared error (RMSE)
    rmse = np.sqrt(np.mean(predictions - y_test) **2)
    rmse

Out[32]: 0.005707483291625976
```

```
In [36]: # plot the data
    train = data[:training_data_len]
    valid = data[training_data_len:]
    valid['Predictions'] = predictions
    # visualize the model
    plt.figure(figsize=(16,8))
    plt.title('Model')
    plt.xlabel('Date', fontsize = 18)
    plt.ylabel('Close Price USD ($)', fontsize= 18)
    plt.plot(train['Close'])
    plt.plot(valid[['Close', 'Predictions']])
    plt.legend(['Train', 'Val', 'Predict'], loc = 'lower right')
```

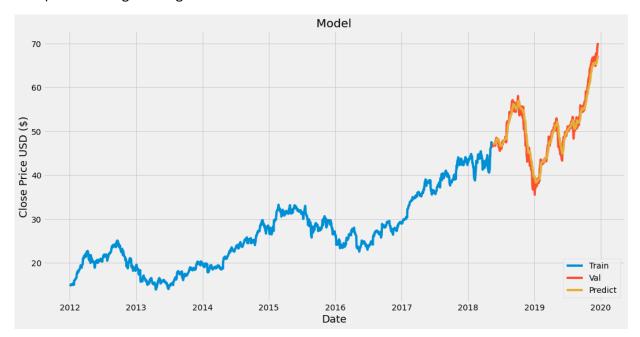
C:\Users\ehsan\AppData\Local\Temp/ipykernel_16560/2019468251.py:4: SettingWithC
opyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

valid['Predictions'] = predictions

Out[36]: <matplotlib.legend.Legend at 0x2975ea16670>



```
In [37]: # Show the valid predicted prices
valid
```

Out[37]:

Date			
2018-05-17	46.747501	47.137703	
2018-05-18	46.577499	47.154289	
2018-05-21	46.907501	47.115097	
2018-05-22	46.790001	47.103394	
2018-05-23	47.090000	47.089199	
2019-12-11	67.692497	65.918266	
2019-12-12	67.864998	66.155891	
2019-12-13	68.787498	66.401611	
2019-12-16	69.964996	66.743469	
2019-12-17	70.102501	67.234062	

400 rows × 2 columns

Close Predictions

```
In [38]: # get the quote
         apple quote = web.DataReader('AAPL', data source='yahoo',start='2012-01-01',end=
         # create a new dataframe
         new df = apple quote.filter(['Close'])
         # Get the las 60 day closing price values and convert the dataframe to an array
         last 60 days = new df[-60:].values
         # Scale the data to be values between 0 and 1
         last 60 days scaled = scaler.transform(last 60 days)
         # create an empty list
         X_{test} = []
         # append the past 60 days
         X_test.append(last_60_days_scaled)
         # Convert the X test data set to numpy array
         X test = np.array(X test)
         # Reshape the data to be 3D
         X_test = np.reshape(X_test,(X_test.shape[0],X_test.shape[1],1))
         # get the predicted scaled price
         pred price = model.predict(X test)
         # undo the scaling
         pred price = scaler.inverse transform(pred price)
         print(pred_price)
```

[[67.725784]]