Daniel Garcia - RNN Apple Stock Prediction

November 21, 2022

Apple Stock Market: https://www.kaggle.com/datasets/meetnagadia/apple-stock-price-from-19802021

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

[2]: apple = pd_read_csy("desktop/Apple/Apple/Apple 2005_2019_csy")
```

```
[2]: apple = pd.read_csv("desktop/Apple/Apple_2005_2019.csv")
apple.head()
```

```
[2]:
            Date
                      Open
                                 High
                                                    Close
                                                            Adj Close
                                                                           Volume
                                            Low
        3/1/2005
                                                             0.966323
                  1.156786
                            1.162679
                                       1.117857
                                                 1.130179
                                                                        691992000
     1 4/1/2005
                  1.139107
                            1.169107
                                       1.124464
                                                 1.141786
                                                             0.976247
                                                                       1096810400
     2 5/1/2005
                  1.151071
                            1.165179
                                       1.143750
                                                 1.151786
                                                             0.984798
                                                                        680433600
                                                                        705555200
     3 6/1/2005
                  1.154821
                            1.159107
                                       1.130893
                                                 1.152679
                                                             0.985561
        7/1/2005
                  1.160714
                            1.243393
                                       1.156250
                                                 1.236607
                                                             1.057321
                                                                       2227450400
```

We will create a new feature called 'Average' that calculates the average between Open and Close. We will build our model using this feature.

```
[3]: apple['Average'] = ( apple['Open'] + apple['Close'] ) / 2 apple.head()
```

```
[3]:
            Date
                      Open
                                High
                                           Low
                                                    Close
                                                           Adj Close
                                                                          Volume
        3/1/2005
                  1.156786
                            1.162679
                                      1.117857
                                                 1.130179
                                                            0.966323
                                                                       691992000
                                      1.124464
                                                            0.976247
                                                                      1096810400
     1 4/1/2005
                  1.139107
                            1.169107
                                                 1.141786
     2 5/1/2005
                  1.151071
                            1.165179
                                      1.143750
                                                 1.151786
                                                            0.984798
                                                                       680433600
     3 6/1/2005
                  1.154821
                            1.159107
                                      1.130893
                                                 1.152679
                                                            0.985561
                                                                       705555200
     4 7/1/2005
                  1.160714 1.243393
                                      1.156250
                                                1.236607
                                                            1.057321
                                                                      2227450400
```

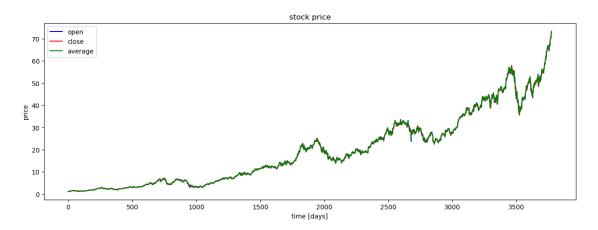
Average

- 0 1.143483
- 1 1.140446
- 2 1.151428
- 3 1.153750
- 4 1.198660

```
[4]: plt.figure(figsize=(15,5)); plt.plot(apple.Open.values, color='blue', label='open')
```

```
plt.plot(apple.Close.values, color='red', label='close')
plt.plot(apple.Average.values, color='green', label='average')
plt.title('stock price')
plt.xlabel('time [days]')
plt.ylabel('price')
plt.legend(loc='best')
```

[4]: <matplotlib.legend.Legend at 0x14571fbb0>



0.1 Split Train and Test

If we randomly choose samples for train and test, we might find several train samples similar to the test ones and get high accuracy, but this would be fake.

Therefore we will split the train and test set between the first 80% samples and the last 20%.

```
[5]: test_size = 0.2
k = np.round(len(apple)*(1-test_size)).astype(int)
apple_train, apple_test = apple[:k], apple[k:]
print("{} train samples, {} test samples".format(apple_train.shape[0],__
apple_test.shape[0]))
```

3020 train samples, 755 test samples

```
[6]: apple_train.head()
```

```
[6]:
            Date
                       Open
                                            Low
                                                     Close
                                                            Adj Close
                                                                            Volume
                                 High
        3/1/2005
                  1.156786
                                                             0.966323
     0
                             1.162679
                                       1.117857
                                                  1.130179
                                                                         691992000
     1 4/1/2005
                  1.139107
                             1.169107
                                       1.124464
                                                  1.141786
                                                             0.976247
                                                                        1096810400
                  1.151071
     2 5/1/2005
                                                             0.984798
                                                                         680433600
                             1.165179
                                       1.143750
                                                  1.151786
     3 6/1/2005
                  1.154821
                             1.159107
                                       1.130893
                                                  1.152679
                                                             0.985561
                                                                         705555200
     4 7/1/2005
                 1.160714
                             1.243393
                                       1.156250
                                                  1.236607
                                                             1.057321
                                                                        2227450400
```

```
Average
      0 1.143483
      1 1.140446
      2 1.151428
      3 1.153750
      4 1.198660
 [7]: average_train = apple_train.iloc[:,7:].values
      average_train
 [7]: array([[ 1.1434825],
             [ 1.1404465],
             [ 1.1514285],
             [29.2225
                        ],
             [29.285
                        ],
             [29.1475
                        ]])
 [8]: from sklearn.preprocessing import MinMaxScaler
      sc = MinMaxScaler(feature_range = (0,1))
      apple_scaled = sc.fit_transform(average_train)
      apple_scaled
 [8]: array([[9.48800286e-05],
             [0.0000000e+00],
             [3.43205690e-04],
             [8.77610685e-01],
             [8.79563914e-01],
             [8.75266811e-01]])
 [9]: x_train = []
      y_train = []
      lent = apple_scaled.shape[0]
      for i in range(k,lent):
          x_train.append(apple_scaled[i-k:i, 0])
          y_train.append(apple_scaled[i,0])
      x_train,y_train = np.array(x_train),np.array(y_train)
      print(x_train.shape, y_train.shape)
     (3019, 1) (3019,)
[10]: | x_train = np.reshape(x_train, (x_train.shape[0],x_train.shape[1],1))
```

0.2 Baseline: RNN

```
[11]: from keras.models import Sequential
     from keras.layers import Dense
     from keras.layers import SimpleRNN
     from keras.layers import Dropout
     apple_rnn = Sequential()
     apple_rnn.add(SimpleRNN(units = 50, return_sequences = True, input_shape = __
       (x_{train.shape}[1],1))
     Metal device set to: Apple M1 Max
     2022-11-21 20:08:33.079127: I
     tensorflow/core/common runtime/pluggable_device/pluggable_device factory.cc:306]
     Could not identify NUMA node of platform GPU ID 0, defaulting to 0. Your kernel
     may not have been built with NUMA support.
     2022-11-21 20:08:33.079255: I
     tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:272]
     Created TensorFlow device (/job:localhost/replica:0/task:0/device:GPU:0 with 0
     MB memory) -> physical PluggableDevice (device: 0, name: METAL, pci bus id:
     <undefined>)
[12]: apple_rnn.add(Dropout(0.2))
     apple_rnn.add(SimpleRNN(units = 50,return_sequences = False))
     apple rnn.add(Dropout(0.2))
     apple_rnn.add(Dense(units = 1))
[13]: import tensorflow as tf
     apple_rnn.compile(optimizer = 'adam',loss = 'mean_squared_error')
     with tf.device('/cpu:0'):
         apple_rnn.fit(x_train,y_train,epochs = 100, batch_size = 32)
     Epoch 1/100
     2022-11-21 20:08:33.231620: W
     tensorflow/core/platform/profile_utils/cpu_utils.cc:128] Failed to get CPU
     frequency: 0 Hz
      1/95 [...] - ETA: 1:07 - loss: 0.1917
     2022-11-21 20:08:33.804013: I
     tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
     Plugin optimizer for device_type GPU is enabled.
     95/95 [========== ] - 1s 5ms/step - loss: 0.0225
     Epoch 2/100
     Epoch 3/100
```

```
95/95 [========= ] - 0s 4ms/step - loss: 0.0043
Epoch 4/100
Epoch 5/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0034
Epoch 6/100
Epoch 7/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0026
Epoch 8/100
95/95 [=========] - Os 4ms/step - loss: 0.0025
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
95/95 [========== ] - Os 4ms/step - loss: 0.0018
Epoch 13/100
95/95 [============== ] - 0s 4ms/step - loss: 0.0017
Epoch 14/100
95/95 [============== ] - 0s 4ms/step - loss: 0.0017
Epoch 15/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0016
Epoch 16/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0016
Epoch 17/100
Epoch 18/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0016
Epoch 19/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0016
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
95/95 [=========] - Os 4ms/step - loss: 0.0013
Epoch 25/100
Epoch 26/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0013
Epoch 27/100
```

```
95/95 [========= ] - 0s 4ms/step - loss: 0.0013
Epoch 28/100
Epoch 29/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0013
Epoch 30/100
Epoch 31/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0012
Epoch 32/100
95/95 [=========] - Os 4ms/step - loss: 0.0012
Epoch 33/100
Epoch 34/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0012
Epoch 35/100
Epoch 36/100
95/95 [========== ] - Os 4ms/step - loss: 0.0012
Epoch 37/100
Epoch 38/100
Epoch 39/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0012
Epoch 40/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0011
Epoch 41/100
Epoch 42/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0011
Epoch 43/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0010
Epoch 44/100
Epoch 45/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0011
Epoch 46/100
Epoch 47/100
Epoch 48/100
95/95 [=========] - Os 4ms/step - loss: 0.0011
Epoch 49/100
Epoch 50/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0011
Epoch 51/100
```

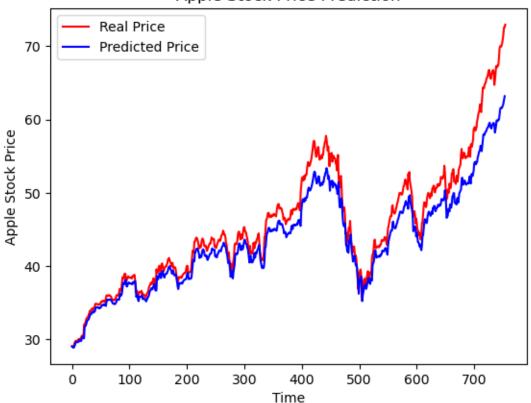
```
95/95 [========= ] - 0s 4ms/step - loss: 0.0010
Epoch 52/100
Epoch 53/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0012
Epoch 54/100
Epoch 55/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0011
Epoch 56/100
95/95 [=========] - Os 4ms/step - loss: 0.0011
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
95/95 [========== ] - Os 4ms/step - loss: 0.0012
Epoch 61/100
Epoch 62/100
Epoch 63/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0011
Epoch 64/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0012
Epoch 65/100
95/95 [========== ] - 0s 4ms/step - loss: 0.0011
Epoch 66/100
Epoch 67/100
Epoch 68/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0010
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
95/95 [========== ] - 0s 4ms/step - loss: 0.0011
Epoch 75/100
```

```
95/95 [========= ] - 0s 4ms/step - loss: 0.0010
Epoch 76/100
Epoch 77/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0010
Epoch 78/100
Epoch 79/100
95/95 [========== ] - Os 4ms/step - loss: 9.9989e-04
Epoch 80/100
95/95 [=========] - Os 4ms/step - loss: 0.0010
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
95/95 [========== ] - Os 4ms/step - loss: 0.0011
Epoch 85/100
Epoch 86/100
Epoch 87/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0010
Epoch 88/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0010
Epoch 89/100
95/95 [=========== - - os 4ms/step - loss: 0.0012
Epoch 90/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0010
Epoch 91/100
Epoch 92/100
Epoch 93/100
95/95 [========== ] - Os 4ms/step - loss: 0.0011
Epoch 94/100
Epoch 95/100
Epoch 96/100
95/95 [=========] - Os 4ms/step - loss: 0.0010
Epoch 97/100
Epoch 98/100
95/95 [========== ] - 0s 4ms/step - loss: 0.0011
Epoch 99/100
```

```
95/95 [=========] - Os 4ms/step - loss: 0.0011
     Epoch 100/100
     95/95 [========== ] - Os 4ms/step - loss: 0.0010
[14]: real_stock_apple = apple_test.iloc[:,7:].values
[15]: dataset_total = apple['Average']
     inputs = dataset_total[len(dataset_total) - len(apple_test)-k:].values
[16]: inputs = inputs.reshape(-1,1)
     inputs = sc.transform(inputs)
[17]: x_test = []
     lent2 = apple_test.shape[0]
     for i in range(k,lent2):
         x_test.append(inputs[i-k:i,0])
     x_test = np.array(x_test)
     x_test.shape
[17]: (754, 1)
[18]: x_test = np.reshape(x_test, (x_test.shape[0],x_test.shape[1],1))
     x_test.shape
[18]: (754, 1, 1)
     0.3 RNN Prediction
[19]: | predicted_price = apple_rnn.predict(x_test)
     24/24 [========] - Os 6ms/step
     2022-11-21 20:09:15.970435: I
     tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
     Plugin optimizer for device_type GPU is enabled.
[20]: predicted_price.shape
[20]: (754, 1)
[21]: | predicted_price = sc.inverse_transform(predicted_price)
[22]: plt.plot(real_stock_apple,color = 'red', label = 'Real Price')
     plt.plot(predicted price, color = 'blue', label = 'Predicted Price')
     plt.title('Apple Stock Price Prediction')
```

```
plt.xlabel('Time')
plt.ylabel('Apple Stock Price')
plt.legend()
plt.show()
```

Apple Stock Price Prediction



```
[23]: real_stock_apple.shape

[23]: (755, 1)

[24]: from sklearn.metrics import r2_score
    real_stock_apple = real_stock_apple[:(len(real_stock_apple)-k)]
    for i in range(real_stock_apple.shape[1]):
        print(r2_score(real_stock_apple[:, i], predicted_price[:, i]))
```

0.8943320267298761

0.4 Experiment 1 LSTM

```
[25]: from keras.layers import LSTM
    apple_lstm = Sequential()
    apple_lstm.add(LSTM(units = 50,return_sequences = True,input_shape = (x_train.
     \hookrightarrowshape[1],1)))
[26]: apple_lstm.add(Dropout(0.2))
    apple_lstm.add(LSTM(units = 50,return_sequences = False))
    apple_lstm.add(Dropout(0.2))
    apple_lstm.add(Dense(units = 1))
[27]: apple_lstm.compile(optimizer = 'adam',loss = 'mean_squared_error')
    with tf.device('/cpu:0'):
       apple_lstm.fit(x_train,y_train,epochs = 100, batch_size = 32)
   Epoch 1/100
   2022-11-21 20:09:17.714567: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
   Epoch 2/100
   95/95 [======== ] - 0s 4ms/step - loss: 0.0068
   Epoch 3/100
   95/95 [======== ] - 0s 4ms/step - loss: 0.0032
   Epoch 4/100
   95/95 [=========== - - os 4ms/step - loss: 0.0029
   Epoch 5/100
   95/95 [========= ] - 0s 4ms/step - loss: 0.0026
   Epoch 6/100
   Epoch 7/100
   Epoch 8/100
   Epoch 9/100
   Epoch 10/100
   95/95 [========= ] - Os 4ms/step - loss: 0.0018
   Epoch 11/100
   95/95 [======== ] - 0s 4ms/step - loss: 0.0018
   Epoch 12/100
   95/95 [========== ] - 0s 4ms/step - loss: 0.0017
   Epoch 13/100
   95/95 [======== ] - 0s 4ms/step - loss: 0.0016
```

```
Epoch 14/100
Epoch 15/100
Epoch 16/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0015
Epoch 17/100
95/95 [============= ] - 0s 4ms/step - loss: 0.0016
Epoch 18/100
95/95 [============= ] - 0s 4ms/step - loss: 0.0015
Epoch 19/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0014
Epoch 20/100
Epoch 21/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0013
Epoch 22/100
95/95 [=========] - Os 4ms/step - loss: 0.0014
Epoch 23/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0014
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0012
Epoch 29/100
Epoch 30/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0012
Epoch 31/100
Epoch 32/100
95/95 [============ ] - 0s 4ms/step - loss: 0.0012
Epoch 33/100
95/95 [============= ] - 0s 4ms/step - loss: 0.0012
Epoch 34/100
95/95 [=========] - Os 4ms/step - loss: 0.0012
Epoch 35/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0012
Epoch 36/100
95/95 [=========] - Os 4ms/step - loss: 0.0011
Epoch 37/100
95/95 [========== ] - Os 4ms/step - loss: 0.0011
```

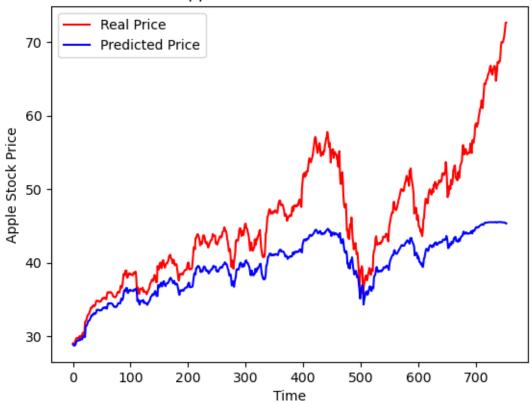
```
Epoch 38/100
Epoch 39/100
Epoch 40/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0011
Epoch 41/100
95/95 [============= ] - 0s 4ms/step - loss: 0.0011
Epoch 42/100
95/95 [============= ] - 0s 4ms/step - loss: 0.0011
Epoch 43/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0012
Epoch 44/100
Epoch 45/100
95/95 [========== ] - 0s 4ms/step - loss: 0.0011
Epoch 46/100
95/95 [=========] - Os 4ms/step - loss: 0.0011
Epoch 47/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0011
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
95/95 [========= ] - 0s 4ms/step - loss: 0.0010
Epoch 54/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0010
Epoch 55/100
Epoch 56/100
95/95 [============ ] - 0s 4ms/step - loss: 0.0010
Epoch 57/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0010
Epoch 58/100
95/95 [=========] - Os 4ms/step - loss: 0.0010
Epoch 59/100
Epoch 60/100
95/95 [=========] - Os 4ms/step - loss: 0.0010
Epoch 61/100
95/95 [=========== ] - Os 4ms/step - loss: 9.5674e-04
```

```
Epoch 62/100
Epoch 63/100
Epoch 64/100
95/95 [======== ] - 0s 4ms/step - loss: 0.0010
Epoch 65/100
Epoch 66/100
95/95 [============== ] - Os 4ms/step - loss: 9.5662e-04
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
95/95 [=========] - Os 4ms/step - loss: 0.0010
Epoch 71/100
Epoch 72/100
Epoch 73/100
95/95 [============= ] - Os 4ms/step - loss: 9.4019e-04
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
95/95 [============ ] - Os 4ms/step - loss: 9.3433e-04
Epoch 79/100
Epoch 80/100
95/95 [============== ] - Os 4ms/step - loss: 9.3545e-04
Epoch 81/100
95/95 [============= ] - Os 4ms/step - loss: 9.6886e-04
Epoch 82/100
95/95 [=========== ] - 0s 4ms/step - loss: 9.0796e-04
Epoch 83/100
Epoch 84/100
95/95 [=========] - Os 4ms/step - loss: 0.0010
Epoch 85/100
95/95 [============ ] - Os 4ms/step - loss: 9.8892e-04
```

```
Epoch 86/100
   Epoch 87/100
   95/95 [============ ] - Os 4ms/step - loss: 9.1647e-04
   Epoch 88/100
   Epoch 89/100
   95/95 [============= ] - 0s 4ms/step - loss: 0.0011
   Epoch 90/100
   95/95 [======== ] - 0s 4ms/step - loss: 0.0010
   Epoch 91/100
   Epoch 92/100
   95/95 [======== ] - 0s 4ms/step - loss: 0.0010
   Epoch 93/100
   Epoch 94/100
   Epoch 95/100
   Epoch 96/100
   Epoch 97/100
   95/95 [=========== ] - Os 4ms/step - loss: 9.4539e-04
   Epoch 98/100
   Epoch 99/100
   Epoch 100/100
   [28]: predicted_price_lstm = apple_lstm.predict(x_test)
   24/24 [========] - 1s 3ms/step
   2022-11-21 20:09:57.617688: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
   2022-11-21 20:09:57.665369: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
   2022-11-21 20:09:57.691751: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
[29]: predicted_price_lstm = sc.inverse_transform(predicted_price_lstm)
```

```
[30]: plt.plot(real_stock_apple,color = 'red', label = 'Real Price')
   plt.plot(predicted_price_lstm, color = 'blue', label = 'Predicted Price')
   plt.title('Apple Stock Price Prediction')
   plt.xlabel('Time')
   plt.ylabel('Apple Stock Price')
   plt.legend()
   plt.show()
```

Apple Stock Price Prediction



```
[31]: for i in range(real_stock_apple.shape[1]): print(r2_score(real_stock_apple[:, i], predicted_price_lstm[:, i]))
```

0.1460723358988939

0.5 Experiment 2 LSTM - No Dropouts

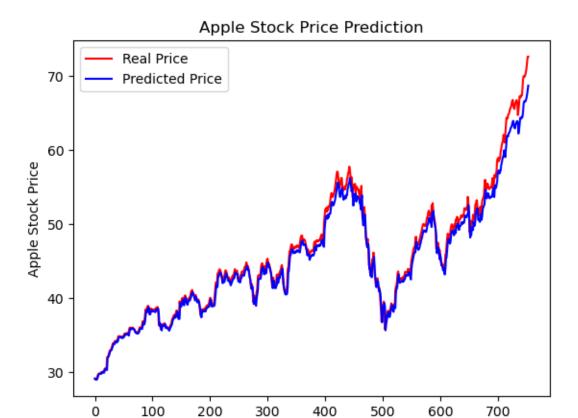
```
[53]: apple_lstm_2.add(LSTM(units = 50, return_sequences = False))
   apple_lstm_2.add(Dense(units = 1))
[54]: apple_lstm_2.compile(optimizer = 'adam',loss = 'mean_squared_error')
   with tf.device('/cpu:0'):
     apple_lstm_2.fit(x_train,y_train,epochs = 100, batch_size = 32)
   Epoch 1/100
   2022-11-21 20:30:58.948517: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
   Epoch 2/100
   Epoch 3/100
   Epoch 4/100
   95/95 [============ ] - Os 4ms/step - loss: 4.5155e-04
   Epoch 5/100
   Epoch 6/100
   Epoch 7/100
   Epoch 8/100
   95/95 [============ ] - Os 4ms/step - loss: 1.1437e-04
   Epoch 9/100
   Epoch 10/100
   95/95 [=========== ] - Os 4ms/step - loss: 6.3541e-05
   Epoch 11/100
   95/95 [=========== ] - Os 4ms/step - loss: 5.6568e-05
   Epoch 12/100
   Epoch 13/100
   Epoch 14/100
   95/95 [============ ] - Os 4ms/step - loss: 5.3874e-05
   Epoch 15/100
   Epoch 16/100
   95/95 [============ ] - Os 4ms/step - loss: 5.3837e-05
   Epoch 17/100
   95/95 [============ ] - Os 4ms/step - loss: 5.2499e-05
   Epoch 18/100
   95/95 [=========== ] - Os 4ms/step - loss: 5.3277e-05
```

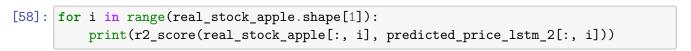
```
Epoch 19/100
95/95 [============== ] - Os 4ms/step - loss: 5.2430e-05
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
95/95 [============== ] - Os 4ms/step - loss: 5.4897e-05
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
95/95 [=========== ] - Os 4ms/step - loss: 5.7466e-05
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
95/95 [============= ] - Os 4ms/step - loss: 5.7133e-05
Epoch 35/100
95/95 [============= ] - Os 4ms/step - loss: 5.6276e-05
Epoch 36/100
Epoch 37/100
95/95 [============== ] - Os 4ms/step - loss: 5.8660e-05
Epoch 38/100
Epoch 39/100
95/95 [=========== ] - Os 4ms/step - loss: 6.6454e-05
Epoch 40/100
Epoch 41/100
95/95 [============ ] - Os 4ms/step - loss: 5.5288e-05
Epoch 42/100
95/95 [============ ] - Os 4ms/step - loss: 6.1391e-05
```

```
Epoch 43/100
95/95 [============= ] - Os 4ms/step - loss: 6.3169e-05
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
95/95 [============== ] - Os 4ms/step - loss: 5.7198e-05
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
95/95 [============ ] - Os 4ms/step - loss: 5.7652e-05
Epoch 59/100
95/95 [============ ] - Os 4ms/step - loss: 6.3310e-05
Epoch 60/100
Epoch 61/100
95/95 [============== ] - Os 4ms/step - loss: 5.8899e-05
Epoch 62/100
Epoch 63/100
95/95 [============ ] - Os 4ms/step - loss: 6.5397e-05
Epoch 64/100
Epoch 65/100
95/95 [============ ] - 0s 4ms/step - loss: 5.8389e-05
Epoch 66/100
95/95 [============ ] - Os 4ms/step - loss: 6.0170e-05
```

```
Epoch 67/100
95/95 [============= ] - Os 4ms/step - loss: 5.6733e-05
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
95/95 [============== ] - Os 4ms/step - loss: 5.6917e-05
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
95/95 [============== ] - Os 4ms/step - loss: 6.8410e-05
Epoch 83/100
95/95 [============= ] - Os 4ms/step - loss: 6.5968e-05
Epoch 84/100
Epoch 85/100
95/95 [============== ] - Os 4ms/step - loss: 5.8935e-05
Epoch 86/100
95/95 [============= ] - Os 4ms/step - loss: 6.2860e-05
Epoch 87/100
95/95 [============ ] - Os 4ms/step - loss: 5.8402e-05
Epoch 88/100
Epoch 89/100
95/95 [============ ] - Os 4ms/step - loss: 5.5947e-05
Epoch 90/100
95/95 [============ ] - Os 4ms/step - loss: 5.8004e-05
```

```
Epoch 91/100
   Epoch 92/100
   Epoch 93/100
   Epoch 94/100
   95/95 [============== ] - Os 4ms/step - loss: 5.6792e-05
   Epoch 95/100
   95/95 [============= ] - Os 4ms/step - loss: 5.6470e-05
   Epoch 96/100
   95/95 [============ ] - Os 4ms/step - loss: 5.9217e-05
   Epoch 97/100
   Epoch 98/100
   Epoch 99/100
   Epoch 100/100
   95/95 [=========== ] - Os 4ms/step - loss: 5.9632e-05
[55]: predicted_price_lstm_2 = apple_lstm_2.predict(x_test)
   24/24 [======== ] - 1s 3ms/step
   2022-11-21 20:31:42.063794: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
   2022-11-21 20:31:42.111951: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
   2022-11-21 20:31:42.134861: I
   tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
   Plugin optimizer for device_type GPU is enabled.
[56]: predicted_price_lstm_2 = sc.inverse_transform(predicted_price_lstm_2)
[57]: plt.plot(real_stock_apple,color = 'red', label = 'Real Price')
    plt.plot(predicted_price_lstm_2, color = 'blue', label = 'Predicted Price')
    plt.title('Apple Stock Price Prediction')
    plt.xlabel('Time')
    plt.ylabel('Apple Stock Price')
    plt.legend()
    plt.show()
```





Time

0.9811742529412535

0