

stock-market-predetection

July 29, 2023

```
[1]: import pandas as pd

df = pd.read_csv('/content/stock price predetection.csv')

df
```

```
[1]:
```

	Date	Open	High	Low	Close	Adj Close \
0	2018-02-05	262.000000	267.899994	250.029999	254.259995	254.259995
1	2018-02-06	247.699997	266.700012	245.000000	265.720001	265.720001
2	2018-02-07	266.579987	272.450012	264.329987	264.559998	264.559998
3	2018-02-08	267.079987	267.619995	250.000000	250.100006	250.100006
4	2018-02-09	253.850006	255.800003	236.110001	249.470001	249.470001
...
1004	2022-01-31	401.970001	427.700012	398.200012	427.140015	427.140015
1005	2022-02-01	432.959991	458.480011	425.540009	457.130005	457.130005
1006	2022-02-02	448.250000	451.980011	426.480011	429.480011	429.480011
1007	2022-02-03	421.440002	429.260010	404.279999	405.600006	405.600006
1008	2022-02-04	407.309998	412.769989	396.640015	410.170013	410.170013

	Volume
0	11896100
1	12595800
2	8981500
3	9306700
4	16906900
...	...
1004	20047500
1005	22542300
1006	14346000
1007	9905200
1008	7782400

[1009 rows x 7 columns]

```
[2]: df = df[['Date', 'Close']]

df
```

```
[2]:
```

	Date	Close
0	2018-02-05	254.259995
1	2018-02-06	265.720001
2	2018-02-07	264.559998
3	2018-02-08	250.100006
4	2018-02-09	249.470001
...
1004	2022-01-31	427.140015
1005	2022-02-01	457.130005
1006	2022-02-02	429.480011
1007	2022-02-03	405.600006
1008	2022-02-04	410.170013

[1009 rows x 2 columns]

```
[3]: df['Date']
```

```
[3]:
```

0	2018-02-05
1	2018-02-06
2	2018-02-07
3	2018-02-08
4	2018-02-09
...	...
1004	2022-01-31
1005	2022-02-01
1006	2022-02-02
1007	2022-02-03
1008	2022-02-04

Name: Date, Length: 1009, dtype: object

```
[29]: import datetime

def str_to_datetime(s):
    split = s.split('-')
    year, month, day = int(split[0]), int(split[1]), int(split[2])
    return datetime.datetime(year=year, month=month, day=day)

datetime_object = str_to_datetime('2018-02-09')
datetime_object
```

```
[29]: datetime.datetime(2018, 2, 9, 0, 0)
```

```
[5]: df
```

```
[5]:
```

	Date	Close
0	2018-02-05	254.259995
1	2018-02-06	265.720001

```

2      2018-02-07    264.559998
3      2018-02-08    250.100006
4      2018-02-09    249.470001
...
1004   2022-01-31    427.140015
1005   2022-02-01    457.130005
1006   2022-02-02    429.480011
1007   2022-02-03    405.600006
1008   2022-02-04    410.170013

```

[1009 rows x 2 columns]

```
[6]: df['Date'] = df['Date'].apply(str_to_datetime)
df['Date']
```

<ipython-input-6-f6fc52bb0fa5>:1: SettingWithCopyWarning:
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

```
df['Date'] = df['Date'].apply(str_to_datetime)
```

```
[6]: 0      2018-02-05
1      2018-02-06
2      2018-02-07
3      2018-02-08
4      2018-02-09
...
1004   2022-01-31
1005   2022-02-01
1006   2022-02-02
1007   2022-02-03
1008   2022-02-04
Name: Date, Length: 1009, dtype: datetime64[ns]
```

```
[7]: df.index = df.pop('Date')
df
```

```
[7]:
```

	Close
Date	
2018-02-05	254.259995
2018-02-06	265.720001
2018-02-07	264.559998
2018-02-08	250.100006
2018-02-09	249.470001
...	...

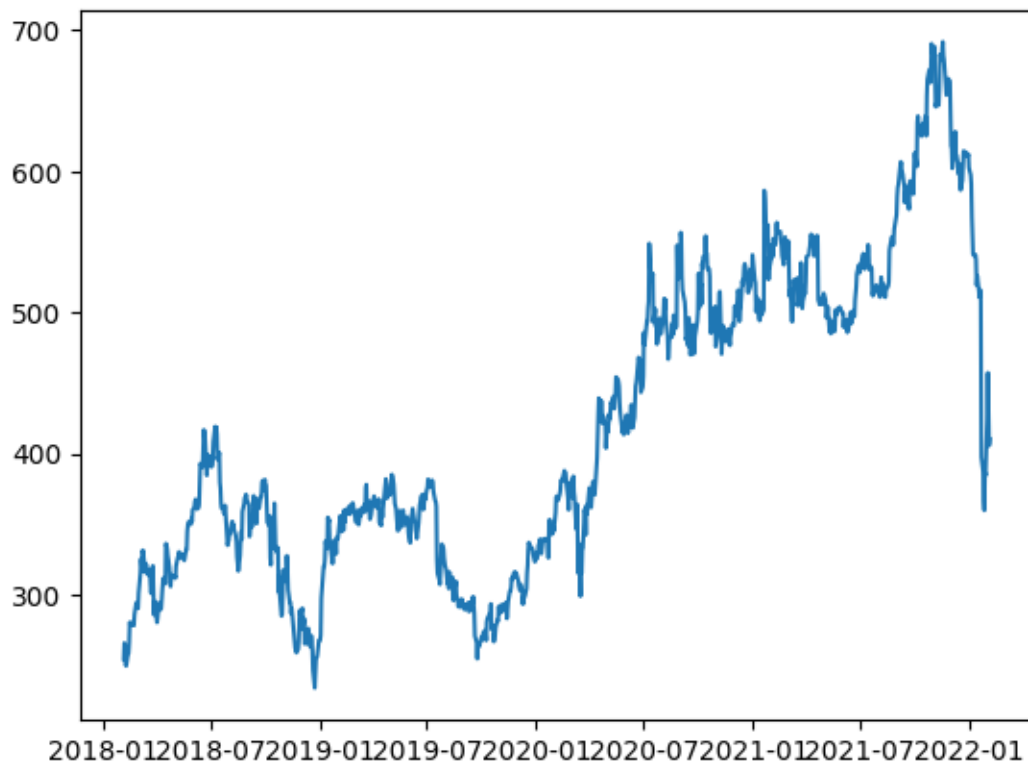
```
2022-01-31  427.140015
2022-02-01  457.130005
2022-02-02  429.480011
2022-02-03  405.600006
2022-02-04  410.170013
```

```
[1009 rows x 1 columns]
```

```
[8]: import matplotlib.pyplot as plt

plt.plot(df.index, df['Close'])
```

```
[8]: [<matplotlib.lines.Line2D at 0x7b73fd8a63b0>]
```



```
[31]: import numpy as np

def df_to_windowed_df(dataframe, first_date_str, last_date_str, n=3):
    first_date = str_to_datetime(first_date_str)
    last_date = str_to_datetime(last_date_str)

    target_date = first_date
```

```

dates = []
X, Y = [], []

last_time = False
while True:
    df_subset = dataframe.loc[:target_date].tail(n+1)

    if len(df_subset) != n+1:
        print(f'Error: Window of size {n} is too large for date {target_date}')
        return

    values = df_subset['Close'].to_numpy()
    x, y = values[:-1], values[-1]

    dates.append(target_date)
    X.append(x)
    Y.append(y)

    next_week = dataframe.loc[target_date:target_date+datetime.
    ↪timedelta(days=7)]
    next_datetime_str = str(next_week.head(2).tail(1).index.values[0])
    next_date_str = next_datetime_str.split('T')[0]
    year_month_day = next_date_str.split('-')
    year, month, day = year_month_day
    next_date = datetime.datetime(day=int(day), month=int(month),
    ↪year=int(year))

    if last_time:
        break

    target_date = next_date

    if target_date == last_date:
        last_time = True

ret_df = pd.DataFrame({})
ret_df['Target Date'] = dates

X = np.array(X)
for i in range(0, n):
    X[:, i]
    ret_df[f'Target-{n-i}'] = X[:, i]

ret_df['Target'] = Y

return ret_df

```

```

windowed_df = df_to_windowed_df(df,
                                '2018-02-09',
                                '2022-02-04',
                                n=3)

windowed_df

```

```

[31]:
      Target Date  Target-3  Target-2  Target-1  Target
0    2018-02-09  265.720001  264.559998  250.100006  249.470001
1    2018-02-12  264.559998  250.100006  249.470001  257.950012
2    2018-02-13  250.100006  249.470001  257.950012  258.269989
3    2018-02-14  249.470001  257.950012  258.269989  266.000000
4    2018-02-15  257.950012  258.269989  266.000000  280.269989
...
1000 2022-01-31  359.700012  386.700012  384.359985  427.140015
1001 2022-02-01  386.700012  384.359985  427.140015  457.130005
1002 2022-02-02  384.359985  427.140015  457.130005  429.480011
1003 2022-02-03  427.140015  457.130005  429.480011  405.600006
1004 2022-02-04  457.130005  429.480011  405.600006  410.170013

[1005 rows x 5 columns]

```

```

[14]: def windowed_df_to_date_X_y(windowed_dataframe):
        df_as_np = windowed_dataframe.to_numpy()

        dates = df_as_np[:, 0]

        middle_matrix = df_as_np[:, 1:-1]
        X = middle_matrix.reshape((len(dates), middle_matrix.shape[1], 1))

        Y = df_as_np[:, -1]

        return dates, X.astype(np.float32), Y.astype(np.float32)

dates, X, y = windowed_df_to_date_X_y(windowed_df)

dates.shape, X.shape, y.shape

```

```

[14]: ((1005,), (1005, 3, 1), (1005,))

```

```

[15]: q_80 = int(len(dates) * .8)
        q_90 = int(len(dates) * .9)

        dates_train, X_train, y_train = dates[:q_80], X[:q_80], y[:q_80]

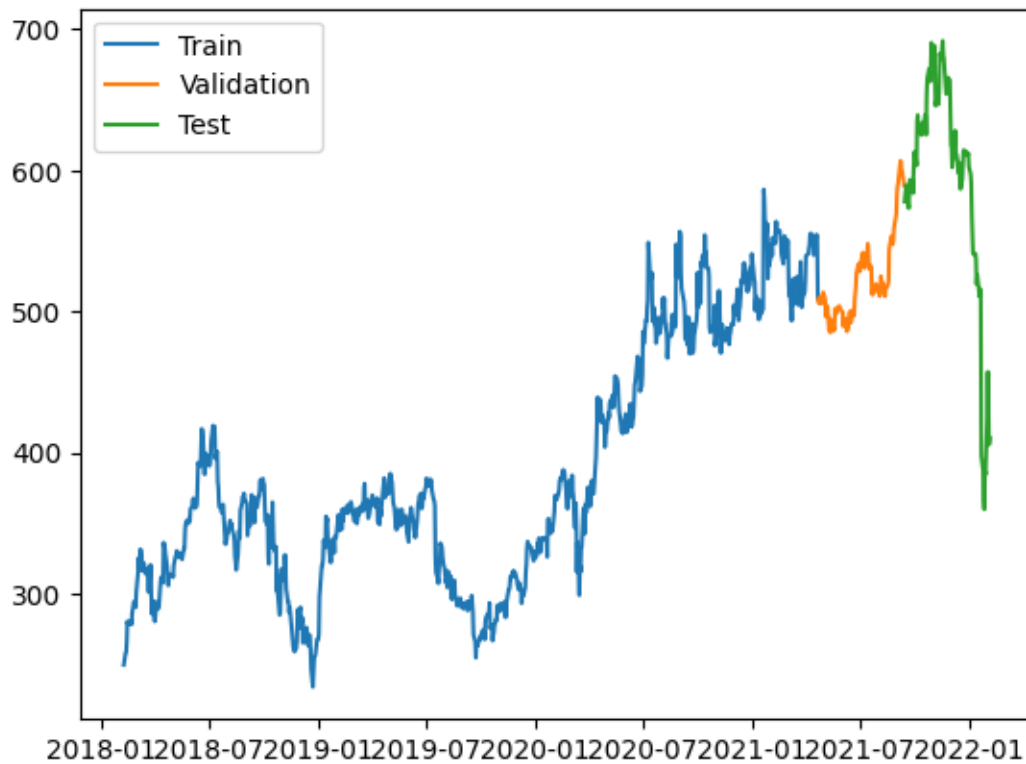
        dates_val, X_val, y_val = dates[q_80:q_90], X[q_80:q_90], y[q_80:q_90]
        dates_test, X_test, y_test = dates[q_90:], X[q_90:], y[q_90:]

```

```
plt.plot(dates_train, y_train)
plt.plot(dates_val, y_val)
plt.plot(dates_test, y_test)

plt.legend(['Train', 'Validation', 'Test'])
```

[15]: <matplotlib.legend.Legend at 0x7b73fb83d210>



```
[16]: from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import Adam
from tensorflow.keras import layers

model = Sequential([layers.Input((3, 1)),
                    layers.LSTM(64),
                    layers.Dense(32, activation='relu'),
                    layers.Dense(32, activation='relu'),
                    layers.Dense(1)])

model.compile(loss='mse',
              optimizer=Adam(learning_rate=0.001),
              metrics=['mean_absolute_error'])
```

```
model.fit(X_train, y_train, validation_data=(X_val, y_val), epochs=100)
```

Epoch 1/100

26/26 [=====] - 3s 27ms/step - loss: 154996.9375 -
mean_absolute_error: 384.2372 - val_loss: 270641.6875 - val_mean_absolute_error:
519.4163

Epoch 2/100

26/26 [=====] - 0s 6ms/step - loss: 152540.8594 -
mean_absolute_error: 381.0434 - val_loss: 265031.1875 - val_mean_absolute_error:
513.9865

Epoch 3/100

26/26 [=====] - 0s 5ms/step - loss: 145159.7969 -
mean_absolute_error: 371.1794 - val_loss: 247883.6875 - val_mean_absolute_error:
497.0260

Epoch 4/100

26/26 [=====] - 0s 5ms/step - loss: 128101.0859 -
mean_absolute_error: 347.2981 - val_loss: 217547.3438 - val_mean_absolute_error:
465.5093

Epoch 5/100

26/26 [=====] - 0s 6ms/step - loss: 102066.1719 -
mean_absolute_error: 307.2246 - val_loss: 173090.4375 - val_mean_absolute_error:
415.0206

Epoch 6/100

26/26 [=====] - 0s 6ms/step - loss: 68615.9922 -
mean_absolute_error: 246.6671 - val_loss: 115762.0938 - val_mean_absolute_error:
338.9886

Epoch 7/100

26/26 [=====] - 0s 5ms/step - loss: 33097.1055 -
mean_absolute_error: 158.6124 - val_loss: 60488.9531 - val_mean_absolute_error:
244.2132

Epoch 8/100

26/26 [=====] - 0s 5ms/step - loss: 12174.9805 -
mean_absolute_error: 82.5767 - val_loss: 26947.6699 - val_mean_absolute_error:
161.5529

Epoch 9/100

26/26 [=====] - 0s 6ms/step - loss: 7535.0234 -
mean_absolute_error: 71.3012 - val_loss: 18550.2520 - val_mean_absolute_error:
133.0515

Epoch 10/100

26/26 [=====] - 0s 5ms/step - loss: 7334.0537 -
mean_absolute_error: 72.5959 - val_loss: 20946.7480 - val_mean_absolute_error:
142.0623

Epoch 11/100

26/26 [=====] - 0s 5ms/step - loss: 5710.9946 -
mean_absolute_error: 58.6846 - val_loss: 15614.4160 - val_mean_absolute_error:
121.8209

Epoch 12/100

26/26 [=====] - 0s 6ms/step - loss: 4388.4004 -
mean_absolute_error: 52.4315 - val_loss: 14112.3740 - val_mean_absolute_error:
116.4280
Epoch 13/100
26/26 [=====] - 0s 6ms/step - loss: 3056.8621 -
mean_absolute_error: 42.8179 - val_loss: 7808.3057 - val_mean_absolute_error:
84.5798
Epoch 14/100
26/26 [=====] - 0s 6ms/step - loss: 1602.6653 -
mean_absolute_error: 29.4872 - val_loss: 3988.1030 - val_mean_absolute_error:
57.9745
Epoch 15/100
26/26 [=====] - 0s 5ms/step - loss: 702.4193 -
mean_absolute_error: 17.7832 - val_loss: 1873.0781 - val_mean_absolute_error:
35.7536
Epoch 16/100
26/26 [=====] - 0s 5ms/step - loss: 343.6540 -
mean_absolute_error: 12.5222 - val_loss: 1048.8779 - val_mean_absolute_error:
23.6797
Epoch 17/100
26/26 [=====] - 0s 6ms/step - loss: 233.2283 -
mean_absolute_error: 10.6512 - val_loss: 734.0581 - val_mean_absolute_error:
18.5219
Epoch 18/100
26/26 [=====] - 0s 6ms/step - loss: 183.5535 -
mean_absolute_error: 9.7191 - val_loss: 505.4833 - val_mean_absolute_error:
14.3068
Epoch 19/100
26/26 [=====] - 0s 5ms/step - loss: 173.3000 -
mean_absolute_error: 9.7049 - val_loss: 498.0577 - val_mean_absolute_error:
14.6273
Epoch 20/100
26/26 [=====] - 0s 5ms/step - loss: 152.0413 -
mean_absolute_error: 8.8392 - val_loss: 386.7752 - val_mean_absolute_error:
12.6382
Epoch 21/100
26/26 [=====] - 0s 5ms/step - loss: 176.3841 -
mean_absolute_error: 9.9926 - val_loss: 358.3921 - val_mean_absolute_error:
12.2324
Epoch 22/100
26/26 [=====] - 0s 5ms/step - loss: 143.2870 -
mean_absolute_error: 8.7372 - val_loss: 373.6602 - val_mean_absolute_error:
14.5344
Epoch 23/100
26/26 [=====] - 0s 5ms/step - loss: 161.2704 -
mean_absolute_error: 9.5224 - val_loss: 316.6615 - val_mean_absolute_error:
11.4183
Epoch 24/100

26/26 [=====] - 0s 5ms/step - loss: 134.2278 -
mean_absolute_error: 8.5973 - val_loss: 438.6989 - val_mean_absolute_error:
15.7713
Epoch 25/100
26/26 [=====] - 0s 5ms/step - loss: 133.1034 -
mean_absolute_error: 8.5429 - val_loss: 292.7980 - val_mean_absolute_error:
10.9124
Epoch 26/100
26/26 [=====] - 0s 6ms/step - loss: 134.4724 -
mean_absolute_error: 8.5050 - val_loss: 326.9892 - val_mean_absolute_error:
12.1983
Epoch 27/100
26/26 [=====] - 0s 5ms/step - loss: 132.6534 -
mean_absolute_error: 8.3531 - val_loss: 262.6467 - val_mean_absolute_error:
10.8086
Epoch 28/100
26/26 [=====] - 0s 6ms/step - loss: 124.8328 -
mean_absolute_error: 8.1036 - val_loss: 271.5528 - val_mean_absolute_error:
10.3862
Epoch 29/100
26/26 [=====] - 0s 6ms/step - loss: 145.1362 -
mean_absolute_error: 9.0476 - val_loss: 327.0142 - val_mean_absolute_error:
12.7272
Epoch 30/100
26/26 [=====] - 0s 6ms/step - loss: 153.6421 -
mean_absolute_error: 9.4326 - val_loss: 241.7687 - val_mean_absolute_error:
10.1551
Epoch 31/100
26/26 [=====] - 0s 5ms/step - loss: 144.6448 -
mean_absolute_error: 9.1171 - val_loss: 334.4837 - val_mean_absolute_error:
13.7688
Epoch 32/100
26/26 [=====] - 0s 5ms/step - loss: 145.9703 -
mean_absolute_error: 8.9010 - val_loss: 240.0051 - val_mean_absolute_error:
10.6825
Epoch 33/100
26/26 [=====] - 0s 6ms/step - loss: 141.2333 -
mean_absolute_error: 8.7895 - val_loss: 231.1074 - val_mean_absolute_error:
10.5566
Epoch 34/100
26/26 [=====] - 0s 6ms/step - loss: 150.0935 -
mean_absolute_error: 9.1135 - val_loss: 252.3089 - val_mean_absolute_error:
11.9935
Epoch 35/100
26/26 [=====] - 0s 5ms/step - loss: 125.5260 -
mean_absolute_error: 8.1293 - val_loss: 259.5168 - val_mean_absolute_error:
10.8996
Epoch 36/100

26/26 [=====] - 0s 6ms/step - loss: 123.0203 -
mean_absolute_error: 8.0662 - val_loss: 215.0604 - val_mean_absolute_error:
9.2880
Epoch 37/100
26/26 [=====] - 0s 5ms/step - loss: 119.3579 -
mean_absolute_error: 7.9570 - val_loss: 228.2355 - val_mean_absolute_error:
9.7337
Epoch 38/100
26/26 [=====] - 0s 6ms/step - loss: 143.9167 -
mean_absolute_error: 8.9914 - val_loss: 212.9570 - val_mean_absolute_error:
10.6698
Epoch 39/100
26/26 [=====] - 0s 8ms/step - loss: 124.8427 -
mean_absolute_error: 8.1303 - val_loss: 201.5243 - val_mean_absolute_error:
9.1840
Epoch 40/100
26/26 [=====] - 0s 8ms/step - loss: 122.2687 -
mean_absolute_error: 8.0143 - val_loss: 285.9185 - val_mean_absolute_error:
14.0102
Epoch 41/100
26/26 [=====] - 0s 8ms/step - loss: 128.5870 -
mean_absolute_error: 8.4487 - val_loss: 241.0711 - val_mean_absolute_error:
10.7622
Epoch 42/100
26/26 [=====] - 0s 9ms/step - loss: 144.6851 -
mean_absolute_error: 9.0708 - val_loss: 269.5070 - val_mean_absolute_error:
12.1778
Epoch 43/100
26/26 [=====] - 0s 9ms/step - loss: 144.7680 -
mean_absolute_error: 8.9673 - val_loss: 281.2650 - val_mean_absolute_error:
12.3390
Epoch 44/100
26/26 [=====] - 0s 9ms/step - loss: 166.0449 -
mean_absolute_error: 9.7934 - val_loss: 251.4678 - val_mean_absolute_error:
12.1553
Epoch 45/100
26/26 [=====] - 0s 8ms/step - loss: 122.9152 -
mean_absolute_error: 8.1216 - val_loss: 315.1829 - val_mean_absolute_error:
13.8814
Epoch 46/100
26/26 [=====] - 0s 8ms/step - loss: 134.7073 -
mean_absolute_error: 8.6140 - val_loss: 232.0075 - val_mean_absolute_error:
9.6392
Epoch 47/100
26/26 [=====] - 0s 12ms/step - loss: 123.3405 -
mean_absolute_error: 8.1037 - val_loss: 234.4566 - val_mean_absolute_error:
12.0321
Epoch 48/100

26/26 [=====] - 1s 27ms/step - loss: 139.1393 -
mean_absolute_error: 8.7122 - val_loss: 226.2381 - val_mean_absolute_error:
9.7101
Epoch 49/100
26/26 [=====] - 0s 8ms/step - loss: 129.9930 -
mean_absolute_error: 8.4888 - val_loss: 400.7219 - val_mean_absolute_error:
17.0841
Epoch 50/100
26/26 [=====] - 0s 6ms/step - loss: 134.7233 -
mean_absolute_error: 8.6200 - val_loss: 196.5728 - val_mean_absolute_error:
10.4564
Epoch 51/100
26/26 [=====] - 0s 5ms/step - loss: 115.5324 -
mean_absolute_error: 7.6912 - val_loss: 179.8604 - val_mean_absolute_error:
8.5853
Epoch 52/100
26/26 [=====] - 0s 6ms/step - loss: 126.7636 -
mean_absolute_error: 8.2791 - val_loss: 285.3101 - val_mean_absolute_error:
13.0291
Epoch 53/100
26/26 [=====] - 0s 6ms/step - loss: 174.1700 -
mean_absolute_error: 9.8724 - val_loss: 198.6720 - val_mean_absolute_error:
9.3454
Epoch 54/100
26/26 [=====] - 0s 5ms/step - loss: 135.8474 -
mean_absolute_error: 8.6971 - val_loss: 248.7760 - val_mean_absolute_error:
11.2760
Epoch 55/100
26/26 [=====] - 0s 5ms/step - loss: 141.7928 -
mean_absolute_error: 9.0761 - val_loss: 179.8068 - val_mean_absolute_error:
9.2702
Epoch 56/100
26/26 [=====] - 0s 5ms/step - loss: 131.4438 -
mean_absolute_error: 8.4372 - val_loss: 185.8106 - val_mean_absolute_error:
8.7765
Epoch 57/100
26/26 [=====] - 0s 5ms/step - loss: 164.7827 -
mean_absolute_error: 9.4662 - val_loss: 208.1839 - val_mean_absolute_error:
9.3379
Epoch 58/100
26/26 [=====] - 0s 5ms/step - loss: 120.1789 -
mean_absolute_error: 8.0302 - val_loss: 370.0333 - val_mean_absolute_error:
16.0233
Epoch 59/100
26/26 [=====] - 0s 5ms/step - loss: 146.5215 -
mean_absolute_error: 9.0395 - val_loss: 204.1731 - val_mean_absolute_error:
10.5463
Epoch 60/100

26/26 [=====] - 0s 6ms/step - loss: 127.8752 -
mean_absolute_error: 8.2559 - val_loss: 177.2521 - val_mean_absolute_error:
8.4993
Epoch 61/100
26/26 [=====] - 0s 5ms/step - loss: 127.3617 -
mean_absolute_error: 8.0215 - val_loss: 170.1115 - val_mean_absolute_error:
8.3845
Epoch 62/100
26/26 [=====] - 0s 6ms/step - loss: 121.1888 -
mean_absolute_error: 7.9866 - val_loss: 186.3846 - val_mean_absolute_error:
8.6958
Epoch 63/100
26/26 [=====] - 0s 5ms/step - loss: 128.4416 -
mean_absolute_error: 8.4885 - val_loss: 256.2840 - val_mean_absolute_error:
12.2310
Epoch 64/100
26/26 [=====] - 0s 6ms/step - loss: 117.2819 -
mean_absolute_error: 7.7766 - val_loss: 174.5816 - val_mean_absolute_error:
8.4774
Epoch 65/100
26/26 [=====] - 0s 6ms/step - loss: 119.6409 -
mean_absolute_error: 7.9041 - val_loss: 249.5272 - val_mean_absolute_error:
12.0038
Epoch 66/100
26/26 [=====] - 0s 5ms/step - loss: 138.1725 -
mean_absolute_error: 8.8449 - val_loss: 174.5928 - val_mean_absolute_error:
9.3277
Epoch 67/100
26/26 [=====] - 0s 5ms/step - loss: 120.9965 -
mean_absolute_error: 7.8849 - val_loss: 333.8981 - val_mean_absolute_error:
15.2993
Epoch 68/100
26/26 [=====] - 0s 5ms/step - loss: 173.5322 -
mean_absolute_error: 9.8697 - val_loss: 273.9355 - val_mean_absolute_error:
13.6015
Epoch 69/100
26/26 [=====] - 0s 6ms/step - loss: 169.2060 -
mean_absolute_error: 9.8738 - val_loss: 192.1674 - val_mean_absolute_error:
8.8002
Epoch 70/100
26/26 [=====] - 0s 6ms/step - loss: 127.6288 -
mean_absolute_error: 8.3910 - val_loss: 161.2447 - val_mean_absolute_error:
8.1436
Epoch 71/100
26/26 [=====] - 0s 6ms/step - loss: 116.2683 -
mean_absolute_error: 7.7953 - val_loss: 169.5639 - val_mean_absolute_error:
8.6035
Epoch 72/100

26/26 [=====] - 0s 6ms/step - loss: 140.9998 -
mean_absolute_error: 8.9246 - val_loss: 235.6805 - val_mean_absolute_error:
12.4371
Epoch 73/100
26/26 [=====] - 0s 6ms/step - loss: 147.0506 -
mean_absolute_error: 9.1186 - val_loss: 260.6047 - val_mean_absolute_error:
13.4693
Epoch 74/100
26/26 [=====] - 0s 6ms/step - loss: 147.9171 -
mean_absolute_error: 9.1503 - val_loss: 334.0437 - val_mean_absolute_error:
13.3708
Epoch 75/100
26/26 [=====] - 0s 6ms/step - loss: 132.3647 -
mean_absolute_error: 8.4266 - val_loss: 173.8935 - val_mean_absolute_error:
8.5375
Epoch 76/100
26/26 [=====] - 0s 6ms/step - loss: 116.6217 -
mean_absolute_error: 7.7895 - val_loss: 160.1826 - val_mean_absolute_error:
8.1590
Epoch 77/100
26/26 [=====] - 0s 6ms/step - loss: 119.7691 -
mean_absolute_error: 7.8931 - val_loss: 164.9925 - val_mean_absolute_error:
8.2591
Epoch 78/100
26/26 [=====] - 0s 7ms/step - loss: 117.7961 -
mean_absolute_error: 7.8479 - val_loss: 158.4522 - val_mean_absolute_error:
8.1553
Epoch 79/100
26/26 [=====] - 0s 6ms/step - loss: 124.6997 -
mean_absolute_error: 8.1172 - val_loss: 158.8931 - val_mean_absolute_error:
8.1078
Epoch 80/100
26/26 [=====] - 0s 6ms/step - loss: 172.3972 -
mean_absolute_error: 9.8032 - val_loss: 206.4727 - val_mean_absolute_error:
9.2698
Epoch 81/100
26/26 [=====] - 0s 6ms/step - loss: 139.3251 -
mean_absolute_error: 8.7868 - val_loss: 206.3963 - val_mean_absolute_error:
10.2427
Epoch 82/100
26/26 [=====] - 0s 6ms/step - loss: 123.9896 -
mean_absolute_error: 8.2499 - val_loss: 270.0038 - val_mean_absolute_error:
13.6096
Epoch 83/100
26/26 [=====] - 0s 7ms/step - loss: 138.7762 -
mean_absolute_error: 8.9411 - val_loss: 184.7938 - val_mean_absolute_error:
9.8043
Epoch 84/100

26/26 [=====] - 0s 6ms/step - loss: 157.5565 -
mean_absolute_error: 9.3807 - val_loss: 181.1399 - val_mean_absolute_error:
8.6562
Epoch 85/100
26/26 [=====] - 0s 6ms/step - loss: 113.8167 -
mean_absolute_error: 7.7927 - val_loss: 429.5895 - val_mean_absolute_error:
18.1913
Epoch 86/100
26/26 [=====] - 0s 6ms/step - loss: 136.2842 -
mean_absolute_error: 8.5810 - val_loss: 187.9289 - val_mean_absolute_error:
9.4659
Epoch 87/100
26/26 [=====] - 0s 7ms/step - loss: 124.2476 -
mean_absolute_error: 8.0992 - val_loss: 189.0924 - val_mean_absolute_error:
10.4338
Epoch 88/100
26/26 [=====] - 0s 7ms/step - loss: 120.9756 -
mean_absolute_error: 7.9098 - val_loss: 190.0829 - val_mean_absolute_error:
8.8320
Epoch 89/100
26/26 [=====] - 0s 6ms/step - loss: 141.3713 -
mean_absolute_error: 9.0235 - val_loss: 195.8291 - val_mean_absolute_error:
8.9304
Epoch 90/100
26/26 [=====] - 0s 9ms/step - loss: 119.0125 -
mean_absolute_error: 8.0502 - val_loss: 232.4824 - val_mean_absolute_error:
12.3532
Epoch 91/100
26/26 [=====] - 0s 6ms/step - loss: 150.5720 -
mean_absolute_error: 9.1180 - val_loss: 189.7423 - val_mean_absolute_error:
8.7747
Epoch 92/100
26/26 [=====] - 0s 6ms/step - loss: 129.7334 -
mean_absolute_error: 8.3602 - val_loss: 270.1428 - val_mean_absolute_error:
14.0312
Epoch 93/100
26/26 [=====] - 0s 6ms/step - loss: 161.9862 -
mean_absolute_error: 9.7016 - val_loss: 244.2034 - val_mean_absolute_error:
10.7154
Epoch 94/100
26/26 [=====] - 0s 6ms/step - loss: 128.9421 -
mean_absolute_error: 8.3233 - val_loss: 168.5362 - val_mean_absolute_error:
9.2309
Epoch 95/100
26/26 [=====] - 0s 6ms/step - loss: 143.4112 -
mean_absolute_error: 8.8294 - val_loss: 202.5568 - val_mean_absolute_error:
10.6321
Epoch 96/100

```

26/26 [=====] - 0s 7ms/step - loss: 131.5090 -
mean_absolute_error: 8.5372 - val_loss: 174.9052 - val_mean_absolute_error:
8.7091
Epoch 97/100
26/26 [=====] - 0s 6ms/step - loss: 121.0583 -
mean_absolute_error: 8.0119 - val_loss: 197.3560 - val_mean_absolute_error:
11.1168
Epoch 98/100
26/26 [=====] - 0s 6ms/step - loss: 141.8681 -
mean_absolute_error: 8.8961 - val_loss: 203.8097 - val_mean_absolute_error:
9.6278
Epoch 99/100
26/26 [=====] - 0s 5ms/step - loss: 133.8441 -
mean_absolute_error: 8.5034 - val_loss: 176.0943 - val_mean_absolute_error:
9.6984
Epoch 100/100
26/26 [=====] - 0s 6ms/step - loss: 132.1107 -
mean_absolute_error: 8.3304 - val_loss: 289.9989 - val_mean_absolute_error:
14.7398

```

[16]: <keras.callbacks.History at 0x7b739c205090>

```

[17]: train_predictions = model.predict(X_train).flatten()

plt.plot(dates_train, train_predictions)
plt.plot(dates_train, y_train)
plt.legend(['Training Predictions', 'Training Observations'])

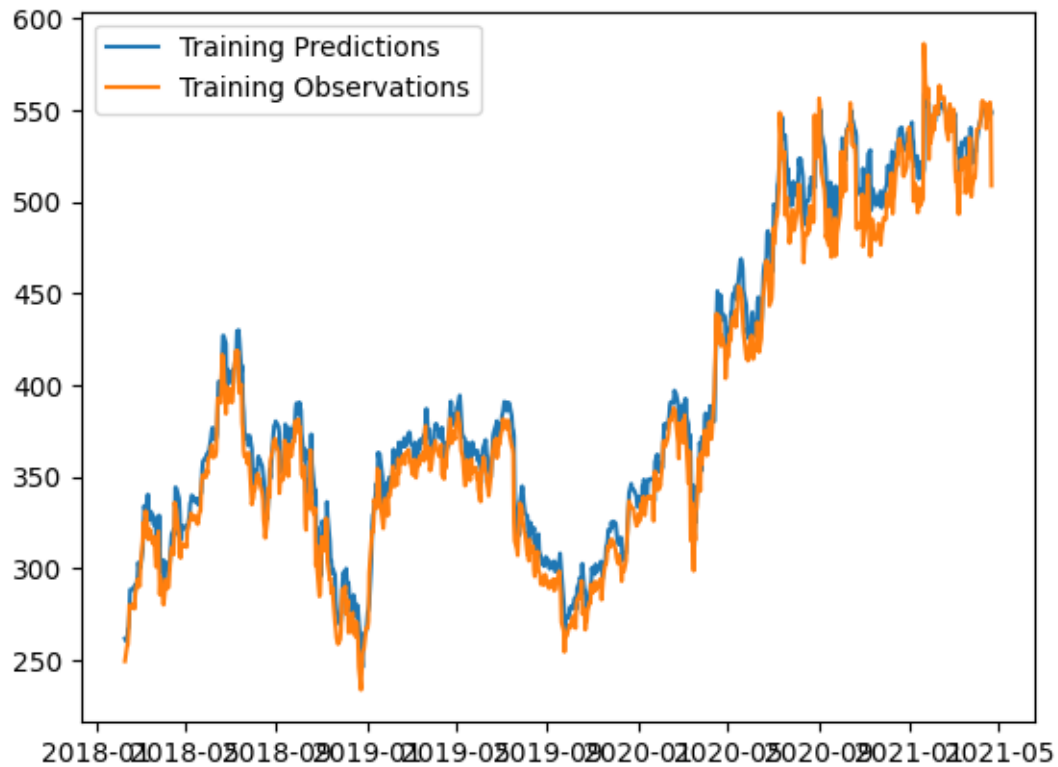
```

```

26/26 [=====] - 1s 3ms/step

```

[17]: <matplotlib.legend.Legend at 0x7b73977763e0>

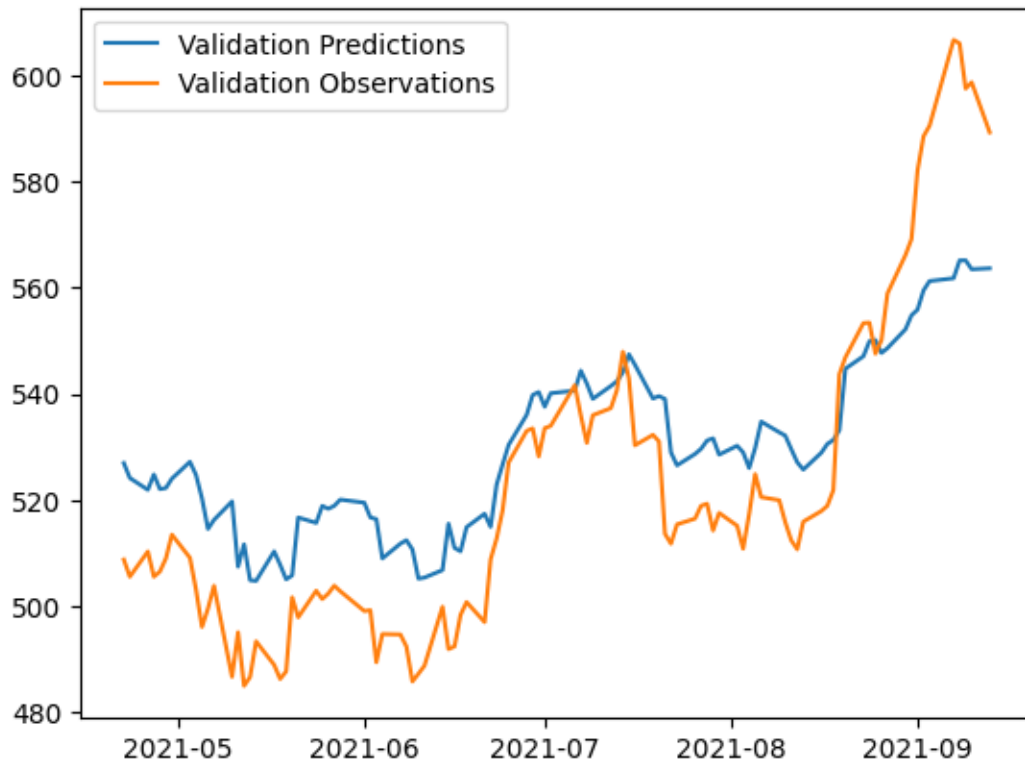


```
[18]: val_predictions = model.predict(X_val).flatten()

plt.plot(dates_val, val_predictions)
plt.plot(dates_val, y_val)
plt.legend(['Validation Predictions', 'Validation Observations'])
```

4/4 [=====] - 0s 4ms/step

```
[18]: <matplotlib.legend.Legend at 0x7b7397626c50>
```

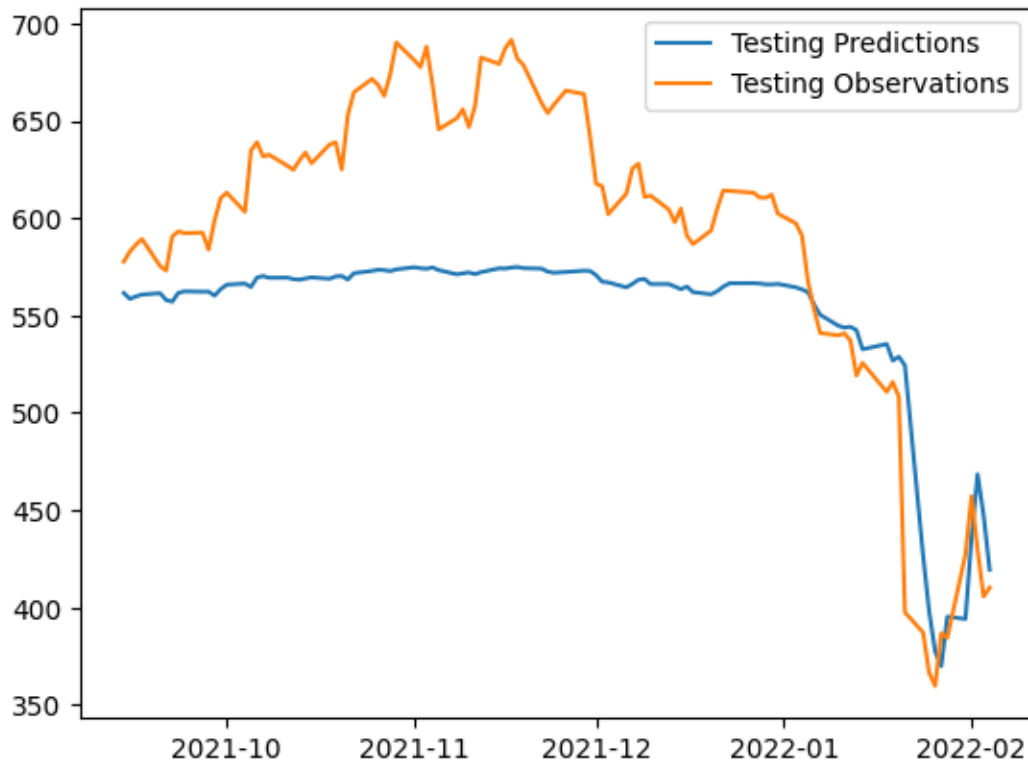


```
[19]: test_predictions = model.predict(X_test).flatten()

plt.plot(dates_test, test_predictions)
plt.plot(dates_test, y_test)
plt.legend(['Testing Predictions', 'Testing Observations'])
```

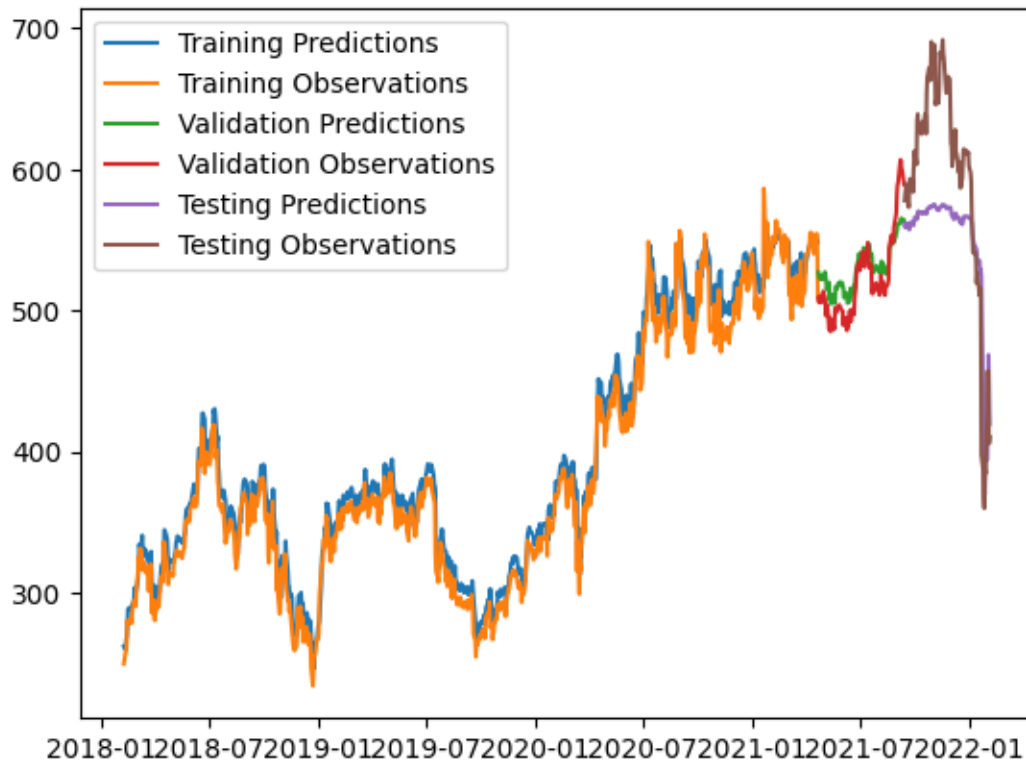
4/4 [=====] - 0s 4ms/step

```
[19]: <matplotlib.legend.Legend at 0x7b7397624310>
```



```
[20]: plt.plot(dates_train, train_predictions)
plt.plot(dates_train, y_train)
plt.plot(dates_val, val_predictions)
plt.plot(dates_val, y_val)
plt.plot(dates_test, test_predictions)
plt.plot(dates_test, y_test)
plt.legend(['Training Predictions',
            'Training Observations',
            'Validation Predictions',
            'Validation Observations',
            'Testing Predictions',
            'Testing Observations'])
```

```
[20]: <matplotlib.legend.Legend at 0x7b7397519900>
```



```
[21]: from copy import deepcopy

recursive_predictions = []
recursive_dates = np.concatenate([dates_val, dates_test])

for target_date in recursive_dates:
    last_window = deepcopy(X_train[-1])
    next_prediction = model.predict(np.array([last_window])).flatten()
    recursive_predictions.append(next_prediction)
    last_window[-1] = next_prediction
```

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```

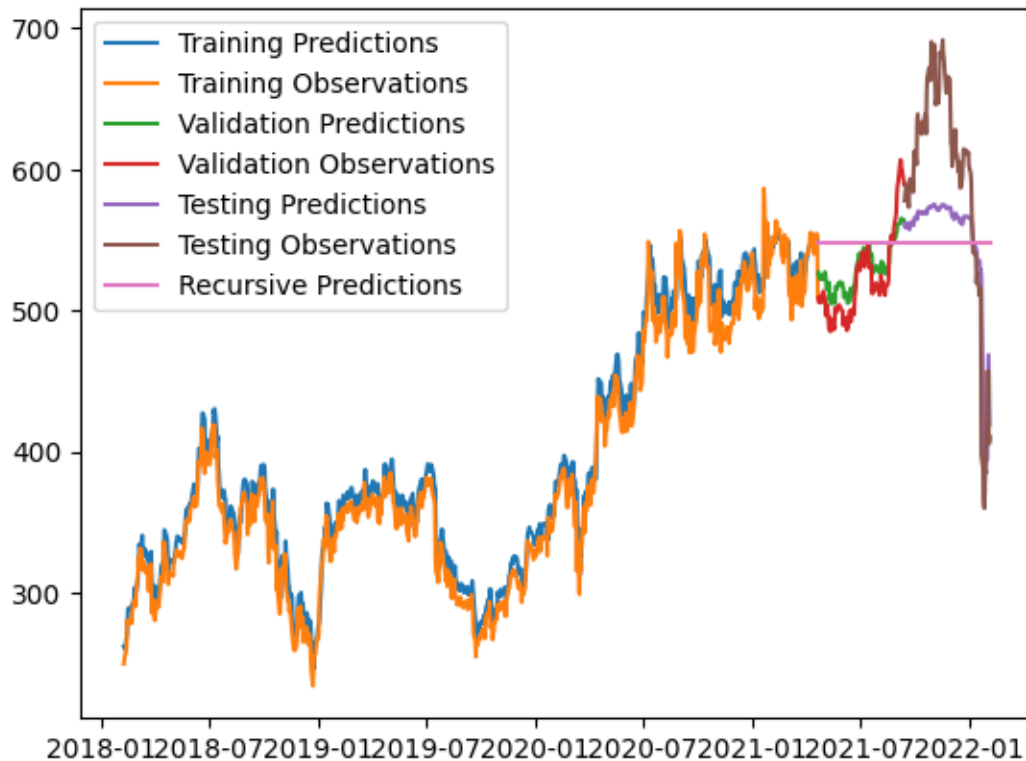
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```



```
[22]: plt.plot(dates_train, train_predictions)
plt.plot(dates_train, y_train)
plt.plot(dates_val, val_predictions)
plt.plot(dates_val, y_val)
plt.plot(dates_test, test_predictions)
plt.plot(dates_test, y_test)
plt.plot(recursive_dates, recursive_predictions)
plt.legend(['Training Predictions',
            'Training Observations',
            'Validation Predictions',
            'Validation Observations',
            'Testing Predictions',
            'Testing Observations',
            'Recursive Predictions'])
```

[22]: <matplotlib.legend.Legend at 0x7b73fb73e680>



[]: