```
import numpy as np
import pandas as pd
from datetime import datetime
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.svm import SVR
from sklearn.metrics import mean_squared_error, mean_absolute_error
# read and merge datasets
data_ohlc = pd.read_csv('sp500.csv')
# data_ohlc['Date'] = pd.to_datetime(data_ohlc['Date'])
# data_ohlc['Date'] = data_ohlc['Date'].dt.strftime("%Y-%m-%d")
data_index = pd.read_csv('sp500_index.csv')
merged = pd.merge(data_ohlc, data_index, on='Date', how='inner')
merged[['Open', 'High', 'Low', 'Close', 'Volume', 'Adj Close']] = merged[['Open', 'High', 'Low', 'Close*', 'Volume', 'Adj Close**']].replace(
# merged[['Close', 'Volume']] = merged[['Close*', 'Volume']].astype(float)
# merged = merged.drop(['Close*', 'S&P500', 'Adj Close**'], axis='columns')
merged['Date'] = [int(datetime.strptime(date, '%Y-%m-%d').timestamp()) for date in merged['Date']]
# merged.dropna(inplace=True)
merged.to csv('form1.csv',index=False)
# pd.read_csv('form1.csv').info()
set = pd.read_csv('form1.csv')
set.head(10)
```

	Date	0pen	High	Low	Close*	Adj Close**	Volu
0	1672704000	3853.29	3878.46	3794.33	3,824.14	3,824.14	3.959140e+
1	1672358400	3829.06	3839.85	3800.34	3,839.50	3,839.50	2.979870e+
2	1672272000	3805.45	3858.19	3805.45	3,849.28	3,849.28	3.003680e+
3	1672185600	3829.56	3848.32	3780.78	3,783.22	3,783.22	3.083520e+
4	1672099200	3843.34	3846.65	3813.22	3,829.25	3,829.25	3.030300e+
5	1671753600	3815.11	3845.80	3797.01	3,844.82	3,844.82	2.819280e+
6	1671667200	3853.26	3853.26	3764.49	3,822.39	3,822.39	3.956950e+
4							<b>+</b>

```
<ipython-input-6-9f83690982b6>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version,
    set.corr()['Close']
Date    0.865539
```

0.998632 0pen 0.999240 High Low 0.999337 Volume 0.157841 0.027479 Return S&P500 1.000000 Close 1.000000 Adi Close 1.000000 Name: Close, dtype: float64

set.corr()['Close']

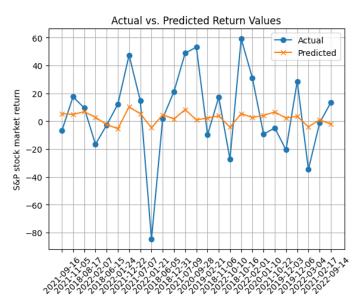
```
X = set[['Date', 'Open', 'High', 'Low', 'Volume', 'Close', 'Adj Close']]
# X = set[['Volume', 'S&P500', 'Close', 'Day', 'Month', 'Year']]
y = set['Return']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4)
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size:
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

# svm_model = SVR(kernel='poly', degree=2, C=100)# best parameter
svm_model = SVR(kernel='linear')
```

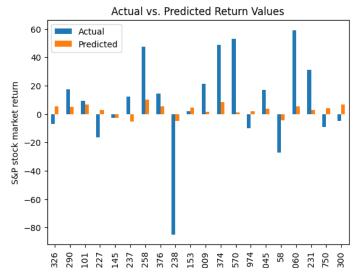
	Actual	Predicted
326	-6.95	5.397390
290	17.47	4.769143
1101	9.44	6.794012
227	-16.66	2.703139
1145	-2.83	-2.534731
237	12.19	-5.416264
258	47.33	10.202859
376	14.59	5.290810
238	-84.79	-4.735621
1153	1.93	4.509782
1009	21.11	1.577971
374	48.73	8.213565
570	53.14	0.934381
974	-9.82	2.155136
1045	17.14	3.749649
58	-27.27	-4.379720
1060	59.13	5.350856

```
merged_df = quick_test.join(X_test)
merged_df = merged_df.head(25)
plt.plot(merged_df['Date'], merged_df['Actual'], label='Actual', marker='o')
plt.plot(merged_df['Date'], merged_df['Predicted'], label='Predicted', marker='x')
plt.xlabel('Date')
plt.ylabel('S&P stock market return')
plt.title('Actual vs. Predicted Return Values')
plt.legend()
plt.xticks(rotation=45)
plt.grid()
plt.show()
```



```
\# X_{\text{test}}['Date'] = [datetime.fromtimestamp(timestamp).strftime('%Y-%m-%d') for timestamp in X_test['Date']]
# merged_df = quick_test.join(X_test)
# merged_df.set_index('Date', inplace=True)
# graph = merged_df.head(25)
# graph.plot(kind='bar')
# plt.xlabel('Date')
# plt.ylabel('S&P stock market return')
# plt.title('Actual vs. Predicted Return Values')
# plt.legend()
# plt.xticks(rotation=45)
# plt.grid()
# plt.show()
graph = quick_test.head(20)
graph.plot(kind='bar')
plt.xlabel('Date')
plt.ylabel('S&P stock market return')
plt.title('Actual vs. Predicted Return Values')
```

Text(0.5, 1.0, 'Actual vs. Predicted Return Values')



```
****WITH OTHER DATA****
```

```
# read and merge datasets
data_ohlc = pd.read_csv('sp500.csv')
# data_ohlc['Date'] = pd.to_datetime(data_ohlc['Date'])
```

```
# data_ohlc['Date'] = data_ohlc['Date'].dt.strftime("%Y-%m-%d")
data_index = pd.read_csv('sp500_index.csv')
merged = pd.merge(data_ohlc, data_index, on='Date', how='inner')
merged[['Open','High', 'Low', 'Adj Close', 'Close', 'Volume']] = merged[['Open','High', 'Low', 'Adj Close**', 'Volume']].replace({'
# merged[['Open','High', 'Low', 'Adj Close**', 'Close', 'Volume']] = merged[['Close*', 'Volume']].astype(float)
merged = merged.drop(['Adj Close**', 'Close*', 'S&P500'], axis='columns')
# merged.dropna(inplace=True)
print(merged)
                                                         Volume Return Adj Close
                 Date
                          0pen
                                   High
                                             Low
     0
           2023-01-03 3853.29 3878.46 3794.33 3.959140e+09 -15.36
                                                                           3824.14
     1
           2022-12-30 3829.06 3839.85 3800.34 2.979870e+09
                                                                  -9.78
                                                                           3839.50
           2022-12-29 3805.45 3858.19 3805.45 3.003680e+09
     2
                                                                  66.06
                                                                           3849,28
     3
           2022-12-28 3829.56 3848.32 3780.78 3.083520e+09 -46.03
                                                                           3783.22
     4
           2022-12-27
                       3843.34 3846.65 3813.22 3.030300e+09 -15.57
                                                                           3829.25
     1255 2018-01-08 2742.67 2748.51 2737.60 3.246160e+09
                                                                   4.56
                                                                           2747.71
          2018-01-05
                       2731.33
                                2743.45
                                         2727.92
                                                                  19.16
                                                                           2743.15
     1257 2018-01-04 2719.31 2729.29 2719.07 3.697340e+09
                                                                  10.93
                                                                           2723.99
     1258 2018-01-03 2697.85 2714.37 2697.77 3.544030e+09
                                                                  17.25
                                                                           2713.06
     1259 2018-01-02 2683.73 2695.89 2682.36 3.397430e+09
                                                                  22.20
                                                                           2695.81
             Close
     a
           3824.14
     1
           3839.50
     2
           3849.28
     3
           3783,22
     4
           3829.25
     1255 2747.71
     1256 2743.15
     1257 2723.99
     1258 2713.06
     1259 2695.81
     [1260 rows x 8 columns]
data_news = pd.read_csv('sp500_news.csv')
merged_data = pd.merge(merged, data_news, on='Date', how='inner')
# merged_data['Date'] = pd.to_datetime(merged_data['Date'])
# merged_data['Day'] = merged_data['Date'].dt.day
# merged_data['Month'] = merged_data['Date'].dt.month
# merged_data['Year'] = merged_data['Date'].dt.year
merged_data['Date'] = [int(datetime.strptime(date, '%Y-%m-%d').timestamp()) for date in merged_data['Date']]
print(merged_data)
merged_data.info()
merged_data.isnull().sum()
                                  High
                                                        Volume Return Adi Close \
                Date
                         0pen
                                            I ow
     0
          1591833600 3123.53 3123.53 2999.49 7.037320e+09 -188.04
                                                                          3002.10
          1591747200 3213.42 3223.27
                                        3181.49
                                                6.598870e+09
                                                               -17.04
                                                                          3190.14
          1591660800 3213.32 3222.71 3193.11 6.410930e+09 -25.21
                                                                          3207.18
     2
          1591574400 3199.92 3233.13 3196.00 8.498300e+09 38.46
     3
                                                                          3232.39
     4
          1591315200 3163.84 3211.72 3163.84 8.639300e+09
                                                                 81.58
                                                                          3193.93
     610 1515369600 2742.67 2748.51
                                        2737.60 3.246160e+09
                                                                          2747.71
                                                                 4.56
     611 1515110400 2731.33 2743.45
                                        2727.92
                                                 3.239280e+09
                                                                 19.16
                                                                          2743.15
     612 1515024000 2719.31 2729.29 2719.07 3.697340e+09
                                                                          2723.99
     613 1514937600 2697.85 2714.37
                                        2697.77 3.544030e+09
                                                                 17.25
                                                                          2713.06
     614 1514851200 2683.73 2695.89 2682.36 3.397430e+09
                                                                 22.20
                                                                          2695.81
            Close Positive Negative
                                        Neutral Price_change
     a
          3002.10 0.067369 0.743126 0.189504
                                                           -1
     1
          3190.14 0.390972 0.240399 0.368628
                                                            -1
          3207.18 0.296441 0.337487 0.366072
     2
                                                            -1
          3232.39 0.523481 0.124646 0.351873
     3
                                                             1
     4
          3193.93 0.567891 0.070503 0.361606
                                                             1
     610 2747.71 0.281969 0.176601 0.541430
                                                             1
     611 2743.15 0.134174 0.150020 0.715806
                                                             1
     612 2723.99 0.201664 0.272806 0.525530
                                                             1
     613 2713.06 0.203532 0.295476 0.500992
     614 2695.81 0.255685 0.267352 0.476963
     [615 rows x 12 columns]
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 615 entries, 0 to 614
```

```
Data columns (total 12 columns):
                                        Dtype
         Column
                       Non-Null Count
     0
                        615 non-null
                                        int64
         Date
     1
          0pen
                        615 non-null
                                        float64
     2
                        615 non-null
                                        float64
         High
     3
                        615 non-null
                                        float64
         Low
     4
         Volume
                        615 non-null
                                        float64
          Return
                        615 non-null
                                        float64
     6
         Adj Close
                        615 non-null
                                        float64
          Close
                        615 non-null
                                        float64
     8
         Positive
                        615 non-null
                                        float64
                        615 non-null
         Negative
                                        float64
     10 Neutral
                        615 non-null
                                        float64
     11 Price_change 615 non-null
                                        int64
     dtypes: float64(10), int64(2)
    memory usage: 62.5 KB
    Date
                    a
    0pen
                    0
    High
                     0
                     0
    Low
    Volume
                    0
    Return
    Adj Close
                     0
     Close
                     0
    Positive
    Negative
                     0
    Neutral
                    a
     Dnice change
merged_data.loc[:, 'ma20'] = merged_data.Close.rolling(20).mean()
merged_data.dropna(inplace=True)
# merged_data.loc[:, "rsi"] = talib.RSI(merged_data.Close, 14)
# print(merged_data)
merged_data.to_csv('form1.csv',index=False)
# pd.read_csv('form1.csv').info()
set = pd.read_csv('form1.csv')
# set = set.drop(['Price_change'], axis='columns')
set.head(10)
```

```
Ad
                         High
        Date
                 0pen
                                  Low
                                             Volume Return
                                                             Close
0 1589414400 2794.54 2852.80 2766.64 5.651130e+09
                                                      32.50 2852.50
1 1589328000 2865.86 2874.14 2793.15 6.151650e+09
                                                     -50.12 2820.00
2 1589241600 2939.50 2945.82 2869.59 5.119630e+09
                                                     -60.07 2870.12
3 1589155200 2915.46 2944.25 2903.44 4.819730e+09
                                                       0.39 2930.19
  1588896000 2908.83 2932.16 2902.88 4.876030e+09
                                                      48.61 2929.80
  1588809600 2878.26 2901.92 2876.48 5.178790e+09
                                                      32.77 2881.19
  1588723200 2883.14 2891.11 2847.65 4.892570e+09
                                                     -20.02 2848.43
```

```
# merged['Date'] = pd.to_datetime(merged['Date'])
# merged['Day'] = merged['Date'].dt.day
# merged['Month'] = merged['Date'].dt.month
# merged['Year'] = merged['Date'].dt.year
# print(merged)

# merged.to_csv('form1.csv',index=False)
# # pd.read_csv('form1.csv').info()
# set = pd.read_csv('form1.csv')
# set.head()

X = set[['Date', 'Open', 'High', 'Low', 'Volume', 'Adj Close', 'Positive', 'Negative', 'Price_change', 'Neutral', 'ma20', 'Close']]
# X = set[['Volume', 'S&P500', 'Close', 'Day', 'Month', 'Year']]
y = set['Return']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4)

scaler = StandardScaler()
```

```
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
# svm_model = SVR(kernel='poly', degree=3, C=100)# best parameter
svm_model = SVR(kernel='linear')
# svm_model = SVR(kernel='rbf', C=1.0, epsilon=0.1)
svm_model.fit(X_train_scaled, y_train)
              SVR
     SVR(kernel='linear')
y_pred = svm_model.predict(X_test_scaled)
mse = mean_squared_error(y_test, y_pred)
mae = mean_absolute_error(y_test, y_pred)
print(f"Mean Squared Error (MSE): {mse:.2f}")
print(f"Mean Absolute Error (MAE): {mae:.2f}")
    Mean Squared Error (MSE): 415.98
    Mean Absolute Error (MAE): 13.93
quick_test = pd.DataFrame({'Actual': y_test, 'Predicted' : y_pred})
quick_test.head(20)
```

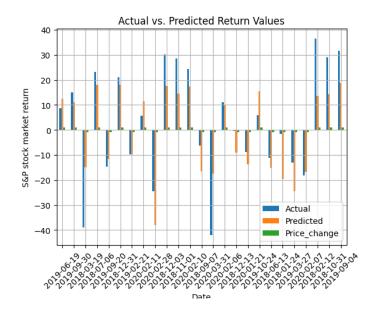
	Actual	Predicted
228	8.71	12.470199
157	14.95	10.933563
543	-39.09	-14.925421
467	23.21	18.101486
163	-14.72	-11.529882
345	21.11	17.920783
310	-9.82	-9.871308
65	5.66	11.605153
53	-24.54	-38.064555
363	30.20	17.454317
384	28.63	14.505847
66	24.38	17.287955
423	-6.37	-16.435005
31	-42.06	-17.389006
68	11.09	9.929528
356	-0.53	-9.092536
80	-8.83	-13.788133

```
merged_df = quick_test.join(X_test)
merged_df = merged_df.drop([ 'Open', 'High', 'Low', 'Volume', 'Adj Close', 'Positive', 'Negative', 'Neutral', 'ma20', 'Close'], axis='columns'
merged_df['Date'] = [datetime.fromtimestamp(timestamp).strftime('%Y-%m-%d') for timestamp in merged_df['Date']]
merged_df
```

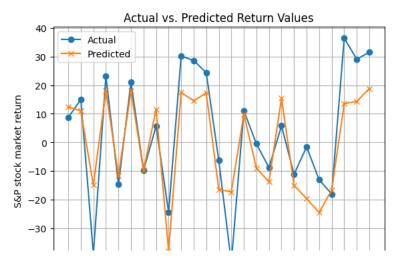
	Actual	Predicted	Date	Price_change
228	8.71	12.470199	2019-06-19	1
157	14.95	10.933563	2019-09-30	1
543	-39.09	-14.925421	2018-03-19	-1
467	23.21	18.101486	2018-07-06	1
163	-14.72	-11.529882	2019-09-20	-1
242	440.00	00 100001	2012 12 22	

merged\_df.set\_index('Date', inplace=True)

```
# Create the bar chart
graph = merged_df.head(25)
graph.plot(kind='bar')
plt.xlabel('Date')
plt.ylabel('S&P stock market return')
plt.title('Actual vs. Predicted Return Values')
plt.legend()
plt.xticks(rotation=45)
plt.grid()
plt.show()
```



```
X_test['Date'] = [datetime.fromtimestamp(timestamp).strftime('%Y-%m-%d') for timestamp in X_test['Date']]
merged_df = quick_test.join(X_test)
merged_df = merged_df.head(25)
plt.plot(merged_df['Date'], merged_df['Actual'], label='Actual', marker='o')
plt.plot(merged_df['Date'], merged_df['Predicted'], label='Predicted', marker='x')
plt.xlabel('Date')
plt.ylabel('S&P stock market return')
plt.title('Actual vs. Predicted Return Values')
plt.legend()
plt.xticks(rotation=45)
plt.grid()
plt.show()
```



4/20/2020 volume-5228630000 close-2823.16 index-2823.16 positi-0.160252 negtive-0.284604 neutral-0.555145 pricechange-(-1) day-20 month-4 year-2020 ma20-2856.8213 expected(-51.4)

