```
globals().clear
import time
import math
import pandas as pd
import numpy as np
from matplotlib import pyplot
import matplotlib.pyplot as plt
import seaborn as sns
sns.set()
%matplotlib inline
from datetime import datetime
pd.options.display.max rows = 5000
pd.options.display.max_columns = 500
from sklearn.svm import SVR
from sklearn.metrics import mean_squared_error
from sklearn.metrics import mean_absolute_percentage_error
from sklearn.preprocessing import MinMaxScaler
from sklearn.ensemble import RandomForestRegressor
# Load dataset
#path = '/content/drive/MyDrive/CapstoneProject/code/02. Machine Learning/data/'
df = pd.read_excel('merged_onehot_test.xlsx')
df.shape
     (13564, 66)
t1=df
t1.index=t1['Date']
t1.drop(columns=t1.columns[0:2],
        axis=1,
        inplace=True)
t1.head()
```

TSLA_close TSLA_vol_4_ave TSLA_vwap_4_ave TSLA_trans_4_ave nasx_clos

```
Date
      2020-06-
         01
                   176,600
                                6531560.00
                                                  174.371825
                                                                       29927.25
                                                                                       952
      10:30:00
      2020-06-
                   176.748
                                4872685.00
                                                  175.236475
                                                                       22062.00
                                                                                       953
         01
      10:45:00
t1.shape
     (13564, 64)
      ∠∪∠∪=∪∪=
train df = t1.loc['2020-06-01 10:30:00':'2021-12-31 16:00:00']
test_df = t1.loc['2022-01-01 09:30:00':'2022-05-27 16:00:00']
                   47E 400
                                202445 00
                                                  470 4E44E0
                                                                      40476 7E
                                                                                       05.
import time
start = time.time()
predictions = list()
a=1
count_time=list()
for i in test_df['week_label'].unique():
    st = time.time()
    scale X = MinMaxScaler()
    test subset = test df[test df['week label']==i]
    print(train df.index[0])
    print(train df.index[-1])
    print(test subset.index[0])
    print(test subset.index[-1])
    train stand = train df.copy()
    test stand = test subset.copy()
    X train, y train = train stand.iloc[:,2:65], train stand.iloc[:,0]
    X_train = scale_X.fit_transform(X_train)
    rf = RandomForestRegressor(n estimators = 100, max depth = 100)
    rf.fit(X train,y train)
    X_test, y_test = test_stand.iloc[:,2:65], test_stand.iloc[:,0]
    X test = scale X.transform(X test)
    y_hat=rf.predict(X_test)
    predictions.append(y hat)
    et = time.time()
    used time=et-st
    count_time.append(used_time)
```

```
train df = train df.append(test df['week label']==i])
   train_df=train_df.drop(train_df[train_df['week_label']==a].index)
   a+=1
   print(train_df.index[0])
   print(train_df.index[-1])
   print('Time taken:'+str(used time))
   print('----')
end = time.time()
print("total used time"+str(end-start))
     lime taken:10.0/3/559/953/964
     -----
    2020-09-08 09:30:00
    2022-04-08 16:00:00
    2022-04-11 09:30:00
    2022-04-14 16:00:00
    2020-09-14 09:30:00
    2022-04-14 16:00:00
    Time taken: 9.748496055603027
    ______
    2020-09-14 09:30:00
    2022-04-14 16:00:00
    2022-04-18 09:30:00
    2022-04-22 16:00:00
    2020-09-21 09:30:00
    2022-04-22 16:00:00
    Time taken: 9.361459016799927
     ______
    2020-09-21 09:30:00
    2022-04-22 16:00:00
    2022-04-25 09:30:00
    2022-04-29 16:00:00
    2020-09-28 09:30:00
    2022-04-29 16:00:00
    Time taken: 9.398331880569458
    ______
    2020-09-28 09:30:00
    2022-04-29 16:00:00
    2022-05-02 09:30:00
    2022-05-06 16:00:00
    2020-10-05 09:30:00
    2022-05-06 16:00:00
    Time taken: 9.61611270904541
    2020-10-05 09:30:00
    2022-05-06 16:00:00
    2022-05-09 09:30:00
    2022-05-13 16:00:00
    2020-10-12 09:30:00
    2022-05-13 16:00:00
    Time taken: 9.35092544555664
```

```
2020-10-12 09:30:00
    2022-05-13 16:00:00
    2022-05-16 09:30:00
    2022-05-20 16:00:00
    2020-10-19 09:30:00
    2022-05-20 16:00:00
    Time taken:9.408068180084229
     -----
    2020-10-19 09:30:00
    2022-05-20 16:00:00
    2022-05-23 09:30:00
    2022-05-27 16:00:00
    2020-10-26 09:30:00
    2022-05-27 16:00:00
    Time taken:10.33316969871521
     -----
    +n+al usad +ima200 7070205551/1526
df_expe = pd.DataFrame(test_df.iloc[:,0])
pred_list= list()
for i in range(len(predictions)):
 pred_list=pred_list+predictions[i].tolist()
df_pred = pd.DataFrame(pred_list,index=test_df.index,columns= ['predict'])
df_Result = pd.concat([df_expe,df_pred],axis=1)
df_Result
```

2022-05-26 15:00:00	703.8000	709.123648
2022-05-26 15:15:00	704.0733	709.407120
2022-05-26 15:30:00	708.6400	706.171643
2022-05-26 15:45:00	707.5500	705.924214
2022-05-26 16:00:00	708.2200	705.595794
2022-05-27 09:30:00	735.1000	702.728110
2022-05-27 09:45:00	741.9400	709.437814
2022-05-27 10:00:00	740.2100	723.536763
2022-05-27 10:15:00	747.6900	720.922105
2022-05-27 10:30:00	750.4500	737.922416
2022-05-27 10:45:00	747.0500	743.238414
2022-05-27 11:00:00	751.5034	745.354393
2022-05-27 11:15:00	751.1136	750.779016
2022-05-27 11:30:00	751.4000	749.893828
2022-05-27 11:45:00	753.1350	748.903928
2022-05-27 12:00:00	753.3241	751.485871
2022-05-27 12:15:00	749.5700	754.491417
2022-05-27 12:30:00	750.3192	754.969176
2022-05-27 12:45:00	750.4500	751.604822
2022-05-27 13:00:00	753.1285	751.345382
2022-05-27 13:15:00	752.0998	751.370943
2022-05-27 13:30:00	750.2650	751.848696
2022-05-27 13:45:00	751.5700	751.901624
2022-05-27 14:00:00	751.8050	751.352857
2022-05-27 14:15:00	754.2700	751.491555
2022-05-27 14:30:00	755.1675	751.293295
2022-05-27 14:45:00	755.5150	754.363616
2022-05-27 15:00:00	756.9600	754.611151
2022-05-27 15:15:00	758.7000	757.740453
2022-05-27 15:30:00	757.6500	758.110167
2022-05-27 15:45:00	759.6600	758.770454
2022-05-27 16:00:00		757.395853

