In [ ]:
```
###necessary libaries###
import numpy as np
import pandas as pd
from seglearn.transform import FeatureRep, SegmentXYForecast, last
from subprocess import check_output
from keras.layers import Dense, Activation, Dropout, Input, LSTM, Flatten
from keras.models import Model
from sklearn.metrics import r2_score
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
import matplotlib.pyplot as plt
from numpy import newaxis
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
from math import sqrt
import glob
import os
from datetime import datetime
import math
from numpy.random import seed
import tensorflow as tf
import warnings
from sklearn.exceptions import DataConversionWarning
import xgboost as xgb
from sklearn.model_selection import ParameterSampler, ParameterGrid

model_seed = 100
# ensure same output results
seed(101)
tf.random.set_seed(model_seed)

# file where csv files lies
path = r'C:\Users\victo\Master_Thesis\merging_data\fiatchrysler\minutely\merged_fil
es'
all_files = glob.glob(os.path.join(path, "*.csv"))

# read files to pandas frame
list_of_files = []

for filename in all_files:
    list_of_files.append(pd.read_csv(filename,
                                     sep=',',
                                     )
                         )

# Concatenate all content of files into one DataFrames
concatenate_dataframe = pd.concat(list_of_files,
                                  ignore_index=True,
                                  axis=0,
                                  )

### analysis with flair sentiment content
new_df_flair_content = concatenate_dataframe[['Date',
                                              'OPEN',
                                              'HIGH',
                                              'LOW',
                                              'CLOSE',
                                              'VOLUME',
                                              'flair_sentiment_content_score']]

new_df_flair_content = new_df_flair_content.fillna(0)
# new_df_flair_content[['Date',
#                         'OPEN',
#                         'HIGH',
#                         'LOW',
#                         'CLOSE',
#                         'VOLUME',
#                         'flair_sentiment_content_score']].astype(np.float64)

new_df_flair_content['Year'] = pd.DatetimeIndex(new_df_flair_content['Date']).year
new_df_flair_content['Month'] = pd.DatetimeIndex(new_df_flair_content['Date']).mont
h
new_df_flair_content['Day'] = pd.DatetimeIndex(new_df_flair_content['Date']).day
new_df_flair_content['Hour'] = pd.DatetimeIndex(new_df_flair_content['Date']).hour
new_df_flair_content['Minute'] = pd.DatetimeIndex(new_df_flair_content['Date']).min
ute
new_df_flair_content['Second'] = pd.DatetimeIndex(new_df_flair_content['Date']).sec
ond

new_df_flair_content = new_df_flair_content.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_flair_content = 0.1

X_train_flair_content, \
X_else_flair_content,\
y_train_flair_content, \
y_else_flair_content = train_test_split(new_df_flair_content,
                                        new_df_flair_content['OPEN'],
                                        test_size=valid_test_size_split_flair_conte
nt*2,
                                        shuffle=False)

X_valid_flair_content, \
X_test_flair_content, \
y_valid_flair_content, \
y_test_flair_content = train_test_split(X_else_flair_content,
                                        y_else_flair_content,
                                        test_size=0.5,
                                        shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)


# normalize data
def minmax_scale_flair_content(df_x, series_y, normalizers_flair_content = None):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME',
                          'flair_sentiment_content_score']

    if not normalizers_flair_content:
        normalizers_flair_content = {}

    for feat in features_to_minmax:
        if feat not in normalizers_flair_content:
            normalizers_flair_content[feat] = MinMaxScaler()
            normalizers_flair_content[feat].fit(df_x[feat].values.reshape(-1, 1))
        df_x[feat] = normalizers_flair_content[feat].transform(df_x[feat].values.re
shape(-1, 1))

    series_y = normalizers_flair_content['OPEN'].transform(series_y.values.reshape
(-1, 1))

    return df_x, series_y, normalizers_flair_content

X_train_norm_flair_content, \
y_train_norm_flair_content, \
normalizers_flair_content = minmax_scale_flair_content(X_train_flair_content,
                                                       y_train_flair_content
                                                       )

X_valid_norm_flair_content, \
y_valid_norm_flair_content, \
_ = minmax_scale_flair_content(X_valid_flair_content,
                               y_valid_flair_content,
                               normalizers_flair_content=normalizers_flair_content
                               )

X_test_norm_flair_content, \
y_test_norm_flair_content, \
_ = minmax_scale_flair_content(X_test_flair_content,
                               y_test_flair_content,
                               normalizers_flair_content=normalizers_flair_content
                               )

def encode_cyclicals_flair_content(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)

    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)


    return df_x

X_train_norm_flair_content = encode_cyclicals_flair_content(X_train_norm_flair_cont
ent)
X_valid_norm_flair_content = encode_cyclicals_flair_content(X_valid_norm_flair_cont
ent)
X_test_norm_flair_content = encode_cyclicals_flair_content(X_test_norm_flair_conten
t)
```

```
# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_flair_content = 60
FORECAST_DISTANCE_flair_content = 30

segmenter_flair_content = SegmentXYForecast(width=TIME_WINDOW_flair_content,
                                            step=1,
                                            y_func=last,
                                            forecast=FORECAST_DISTANCE_flair_conten
t
                                            )

X_train_rolled_flair_content, \
y_train_rolled_flair_content, \
_ = segmenter_flair_content.fit_transform([X_train_norm_flair_content.values],
                                          [y_train_norm_flair_content.flatten()]
                                          )

X_valid_rolled_flair_content, \
y_valid_rolled_flair_content, \
_ = segmenter_flair_content.fit_transform([X_valid_norm_flair_content.values],
                                          [y_valid_norm_flair_content.flatten()]
                                          )

X_test_rolled_flair_content, \
y_test_rolled_flair_content, \
_ = segmenter_flair_content.fit_transform([X_test_norm_flair_content.values],
                                          [y_test_norm_flair_content.flatten()]
                                          )

shape_flair_content = X_train_rolled_flair_content.shape
X_train_flattened_flair_content = X_train_rolled_flair_content.reshape(shape_flair_
content[0],
                                                                       shape_flair_
content[1]*shape_flair_content[2]
                                                                       )

X_train_flattened_flair_content.shape
shape_flair_content = X_valid_rolled_flair_content.shape
X_valid_flattened = X_valid_rolled_flair_content.reshape(shape_flair_content[0],
                                                         shape_flair_content[1]*sha
pe_flair_content[2]
                                                         )

# Random Forest
N_ESTIMATORS_flair_content = 30
RANDOM_STATE_flair_content = 452543634

RF_base_model_flair_content = RandomForestRegressor(random_state=RANDOM_STATE_flair
_content,
                                                    n_estimators=N_ESTIMATORS_flair
_content,
                                                    n_jobs=-1,
                                                    verbose=100
                                                    )

RF_base_model_flair_content.fit(X_train_flattened_flair_content, y_train_rolled_fla
ir_content)
print(' ')
print("------------------------------------------------------------------")
print(' ')
RF_base_model_predictions_flair_content = RF_base_model_flair_content.predict(X_val
id_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_flair_content = sqrt(mean_squared_error(y_valid_rolled_flair_content,
                                                 RF_base_model_predictions_flair_co
ntent
                                                 )
                              )

print("Root mean squared error on valid:",rms_base_flair_content)
print("Root mean squared error on valid inverse transformed from normalization:",no
rmalizers_flair_content["OPEN"]
      .inverse_transform(np.array([rms_base_flair_content]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_flair_content = normalizers_flair_content['OPEN']\
                                          .inverse_transform(np.array(RF_base_model
_predictions_flair_content).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

### analysis with flair header
new_df_flair_header = concatenate_dataframe[['Date',
                                             'OPEN',
                                             'HIGH',
                                             'LOW',
                                             'CLOSE',
                                             'VOLUME',
                                             'flair_sentiment_header_score']]

new_df_flair_header = new_df_flair_header.fillna(0)
# new_df_flair_header[['Date',
#                      'OPEN',
#                      'HIGH',
#                      'LOW',
#                      'CLOSE',
#                      'VOLUME',
#                      'flair_sentiment_header_score']].astype(np.float64)

new_df_flair_header['Year'] = pd.DatetimeIndex(new_df_flair_header['Date']).year
new_df_flair_header['Month'] = pd.DatetimeIndex(new_df_flair_header['Date']).month
new_df_flair_header['Day'] = pd.DatetimeIndex(new_df_flair_header['Date']).day
new_df_flair_header['Hour'] = pd.DatetimeIndex(new_df_flair_header['Date']).hour
new_df_flair_header['Minute'] = pd.DatetimeIndex(new_df_flair_header['Date']).minut
e
new_df_flair_header['Second'] = pd.DatetimeIndex(new_df_flair_header['Date']).secon
d

new_df_flair_header = new_df_flair_header.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_flair_header = 0.1

X_train_flair_header, \
X_else_flair_header,\
y_train_flair_header, \
y_else_flair_header = train_test_split(new_df_flair_header,
                                       new_df_flair_header['OPEN'],
                                        test_size=valid_test_size_split_flair_header
*2,
                                        shuffle=False)

X_valid_flair_header, \
X_test_flair_header, \
y_valid_flair_header, \
y_test_flair_header = train_test_split(X_else_flair_header,
                                        y_else_flair_header,
                                        test_size=0.5,
                                        shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)


# normalize data
def minmax_scale_flair_header(df_x, series_y, normalizers_flair_header = None):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME',
                          'flair_sentiment_header_score']

    if not normalizers_flair_header:
        normalizers_flair_header = {}

    for feat in features_to_minmax:
        if feat not in normalizers_flair_header:
            normalizers_flair_header[feat] = MinMaxScaler()
            normalizers_flair_header[feat].fit(df_x[feat].values.reshape(-1, 1))

        df_x[feat] = normalizers_flair_header[feat].transform(df_x[feat].values.res
hape(-1, 1))

    series_y = normalizers_flair_header['OPEN'].transform(series_y.values.reshape(-
1, 1))

    return df_x, series_y, normalizers_flair_header

X_train_norm_flair_header, \
y_train_norm_flair_header, \
normalizers_flair_header = minmax_scale_flair_header(X_train_flair_header,
                                                     y_train_flair_header
                                                     )

X_valid_norm_flair_header, \
y_valid_norm_flair_header, \
_ = minmax_scale_flair_header(X_valid_flair_header,
                              y_valid_flair_header,
                              normalizers_flair_header=normalizers_flair_header
                              )

X_test_norm_flair_header, \
y_test_norm_flair_header, \
_ = minmax_scale_flair_header(X_test_flair_header,
                              y_test_flair_header,
                              normalizers_flair_header=normalizers_flair_header
                              )

def encode_cyclicals_flair_header(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)

    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)


    return df_x

X_train_norm_flair_header = encode_cyclicals_flair_header(X_train_norm_flair_heade
r)
X_valid_norm_flair_header = encode_cyclicals_flair_header(X_valid_norm_flair_heade
r)
X_test_norm_flair_header = encode_cyclicals_flair_header(X_test_norm_flair_header)

# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_flair_header = 60
FORECAST_DISTANCE_flair_header = 30

segmenter_flair_header = SegmentXYForecast(width=TIME_WINDOW_flair_header,
                                           step=1,
                                           y_func=last,
                                           forecast=FORECAST_DISTANCE_flair_header
                                           )

X_train_rolled_flair_header, \
y_train_rolled_flair_header, \
_ = segmenter_flair_header.fit_transform([X_train_norm_flair_header.values],
                                         [y_train_norm_flair_header.flatten()]
                                         )

X_valid_rolled_flair_header, \
y_valid_rolled_flair_header, \
_ = segmenter_flair_header.fit_transform([X_valid_norm_flair_header.values],
                                         [y_valid_norm_flair_header.flatten()]
                                         )

X_test_rolled_flair_header, \
y_test_rolled_flair_header, \
_ = segmenter_flair_header.fit_transform([X_test_norm_flair_header.values],
                                         [y_test_norm_flair_header.flatten()]
                                         )
```

```
shape_flair_header = X_train_rolled_flair_header.shape
X_train_flattened_flair_header = X_train_rolled_flair_header.reshape(shape_flair_he
ader[0],
                                                                      shape_flair_he
ader[1]*shape_flair_header[2]
                                                                      )

X_train_flattened_flair_header.shape
shape_flair_header = X_valid_rolled_flair_header.shape
X_valid_flattened = X_valid_rolled_flair_header.reshape(shape_flair_header[0],
                                                        shape_flair_header[1]*shape
_flair_header[2]
                                                        )

# Random Forest
N_ESTIMATORS_flair_header = 30
RANDOM_STATE_flair_header = 452543634

RF_base_model_flair_header = RandomForestRegressor(random_state=RANDOM_STATE_flair_
header,
                                                   n_estimators=N_ESTIMATORS_flair_
header,
                                                   n_jobs=-1,
                                                   verbose=100
                                                   )

RF_base_model_flair_header.fit(X_train_flattened_flair_header, y_train_rolled_flair
_header)
print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_flair_header = RF_base_model_flair_header.predict(X_valid
_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_flair_header = sqrt(mean_squared_error(y_valid_rolled_flair_header,
                                                RF_base_model_predictions_flair_hea
der
                                                )
                             )

print("Root mean squared error on valid:",rms_base_flair_header)
print("Root mean squared error on valid inverse transformed from normalization:",no
rmalizers_flair_header["OPEN"]
      .inverse_transform(np.array([rms_base_flair_header]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_flair_header = normalizers_flair_header['OPEN']\
                                          .inverse_transform(np.array(RF_base_model
_predictions_flair_header).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

### analysis with textblob sentiment content
new_df_textblob_content = concatenate_dataframe[['Date',
                                                 'OPEN',
                                                 'HIGH',
                                                 'LOW',
                                                 'CLOSE',
                                                 'VOLUME',
                                                 'polarity_textblob_sentiment_conte
nt']]

new_df_textblob_content = new_df_textblob_content.fillna(0)
# new_df_textblob_content[['Date',
#                            'OPEN',
#                            'HIGH',
#                            'LOW',
#                            'CLOSE',
#                            'VOLUME',
#                            'polarity_textblob_sentiment_content']].astype(np.float6
4)

new_df_textblob_content['Year'] = pd.DatetimeIndex(new_df_textblob_content['Date
']).year
new_df_textblob_content['Month'] = pd.DatetimeIndex(new_df_textblob_content['Date
']).month
new_df_textblob_content['Day'] = pd.DatetimeIndex(new_df_textblob_content['Date']).
day
new_df_textblob_content['Hour'] = pd.DatetimeIndex(new_df_textblob_content['Date
']).hour
new_df_textblob_content['Minute'] = pd.DatetimeIndex(new_df_textblob_content['Date
']).minute
new_df_textblob_content['Second'] = pd.DatetimeIndex(new_df_textblob_content['Date
']).second

new_df_textblob_content = new_df_textblob_content.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_textblob_content = 0.1

X_train_textblob_content, \
X_else_textblob_content,\
y_train_textblob_content, \
y_else_textblob_content = train_test_split(new_df_textblob_content,
                                           new_df_textblob_content['OPEN'],
                                           test_size=valid_test_size_split_textblob
_content*2,
                                           shuffle=False)

X_valid_textblob_content, \
X_test_textblob_content, \
y_valid_textblob_content, \
y_test_textblob_content = train_test_split(X_else_textblob_content,
                                           y_else_textblob_content,
                                           test_size=0.5,
                                           shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)


# normalize data
def minmax_scale_textblob_content(df_x, series_y, normalizers_textblob_content = No
ne):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME',
                          'polarity_textblob_sentiment_content']

    if not normalizers_textblob_content:
        normalizers_textblob_content = {}

    for feat in features_to_minmax:
        if feat not in normalizers_textblob_content:
            normalizers_textblob_content[feat] = MinMaxScaler()
            normalizers_textblob_content[feat].fit(df_x[feat].values.reshape(-1,
1))

        df_x[feat] = normalizers_textblob_content[feat].transform(df_x[feat].value
s.reshape(-1, 1))

    series_y = normalizers_textblob_content['OPEN'].transform(series_y.values.resha
pe(-1, 1))

    return df_x, series_y, normalizers_textblob_content

X_train_norm_textblob_content, \
y_train_norm_textblob_content, \
normalizers_textblob_content = minmax_scale_textblob_content(X_train_textblob_conte
nt,
                                                              y_train_textblob_conte
nt
                                                              )

X_valid_norm_textblob_content, \
y_valid_norm_textblob_content, \
_ = minmax_scale_textblob_content(X_valid_textblob_content,
                                  y_valid_textblob_content,
                                  normalizers_textblob_content=normalizers_textblob
_content
                                  )

X_test_norm_textblob_content, \
y_test_norm_textblob_content, \
_ = minmax_scale_textblob_content(X_test_textblob_content,
                                  y_test_textblob_content,
                                  normalizers_textblob_content=normalizers_textblob
_content
                                  )

def encode_cyclicals_textblob_content(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)

    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)


    return df_x

X_train_norm_textblob_content = encode_cyclicals_textblob_content(X_train_norm_text
blob_content)
X_valid_norm_textblob_content = encode_cyclicals_textblob_content(X_valid_norm_text
blob_content)
X_test_norm_textblob_content = encode_cyclicals_textblob_content(X_test_norm_textbl
ob_content)

# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_textblob_content = 60
FORECAST_DISTANCE_textblob_content = 30

segmenter_textblob_content = SegmentXYForecast(width=TIME_WINDOW_textblob_content,
                                               step=1,
                                               y_func=last,
                                               forecast=FORECAST_DISTANCE_textblob_
content
                                               )

X_train_rolled_textblob_content, \
y_train_rolled_textblob_content, \
_ = segmenter_textblob_content.fit_transform([X_train_norm_textblob_content.value
s],
                                             [y_train_norm_textblob_content.flatten
()]
                                             )

X_valid_rolled_textblob_content, \
y_valid_rolled_textblob_content, \
_ = segmenter_textblob_content.fit_transform([X_valid_norm_textblob_content.value
s],
                                             [y_valid_norm_textblob_content.flatten
()]
                                             )

X_test_rolled_textblob_content, \
y_test_rolled_textblob_content, \
_ = segmenter_textblob_content.fit_transform([X_test_norm_textblob_content.values],
                                             [y_test_norm_textblob_content.flatten
()]
                                             )

shape_textblob_content = X_train_rolled_textblob_content.shape
X_train_flattened_textblob_content = X_train_rolled_textblob_content.reshape(shape_
textblob_content[0],
                                                                             shape_
textblob_content[1]*shape_textblob_content[2]
                                                                             )

X_train_flattened_textblob_content.shape
shape_textblob_content = X_valid_rolled_textblob_content.shape
```

```
X_valid_flattened = X_valid_rolled_textblob_content.reshape(shape_textblob_content
[0],
                                                            shape_textblob_content
[1]*shape_textblob_content[2]
                                                            )

# Random Forest
N_ESTIMATORS_textblob_content = 30
RANDOM_STATE_textblob_content = 452543634

RF_base_model_textblob_content = RandomForestRegressor(random_state=RANDOM_STATE_te
xtblob_content,
                                                       n_estimators=N_ESTIMATORS_te
xtblob_content,
                                                       n_jobs=-1,
                                                       verbose=100
                                                       )

RF_base_model_textblob_content.fit(X_train_flattened_textblob_content, y_train_roll
ed_textblob_content)
print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_textblob_content = RF_base_model_textblob_content.predict
(X_valid_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_textblob_content = sqrt(mean_squared_error(y_valid_rolled_textblob_conten
t,
                                                    RF_base_model_predictions_textb
lob_content
                                                    )
                                 )

print("Root mean squared error on valid:",rms_base_textblob_content)
print("Root mean squared error on valid inverse transformed from normalization:",no
rmalizers_textblob_content["OPEN"]
      .inverse_transform(np.array([rms_base_textblob_content]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_textblob_content = normalizers_textblob_content['OPEN']\
                                          .inverse_transform(np.array(RF_base_model
_predictions_textblob_content).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

### analysis with textblob header
new_df_textblob_header = concatenate_dataframe[['Date',
                                                'OPEN',
                                                'HIGH',
                                                'LOW',
                                                'CLOSE',
                                                'VOLUME',
                                                'polarity_textblob_sentiment_header
']]
```

```
new_df_textblob_header = new_df_textblob_header.fillna(0)
# new_df_textblob_header[['Date',
#                          'OPEN',
#                          'HIGH',
#                          'LOW',
#                          'CLOSE',
#                          'VOLUME',
#                          'polarity_textblob_sentiment_header']].astype(np.float64)

new_df_textblob_header['Year'] = pd.DatetimeIndex(new_df_textblob_header['Date']).y
ear
new_df_textblob_header['Month'] = pd.DatetimeIndex(new_df_textblob_header['Date']).
month
new_df_textblob_header['Day'] = pd.DatetimeIndex(new_df_textblob_header['Date']).da
y
new_df_textblob_header['Hour'] = pd.DatetimeIndex(new_df_textblob_header['Date']).h
our
new_df_textblob_header['Minute'] = pd.DatetimeIndex(new_df_textblob_header['Date
']).minute
new_df_textblob_header['Second'] = pd.DatetimeIndex(new_df_textblob_header['Date
']).second

new_df_textblob_header = new_df_textblob_header.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_textblob_header = 0.1

X_train_textblob_header, \
X_else_textblob_header,\
y_train_textblob_header, \
y_else_textblob_header = train_test_split(new_df_textblob_header,
                                          new_df_textblob_header['OPEN'],
                                          test_size=valid_test_size_split_textblob_
header*2,
                                          shuffle=False)

X_valid_textblob_header, \
X_test_textblob_header, \
y_valid_textblob_header, \
y_test_textblob_header = train_test_split(X_else_textblob_header,
                                          y_else_textblob_header,
                                          test_size=0.5,
                                          shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)


# normalize data
def minmax_scale_textblob_header(df_x, series_y, normalizers_textblob_header = Non
e):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME',
                          'polarity_textblob_sentiment_header']

    if not normalizers_textblob_header:
        normalizers_textblob_header = {}

    for feat in features_to_minmax:
        if feat not in normalizers_textblob_header:
            normalizers_textblob_header[feat] = MinMaxScaler()
            normalizers_textblob_header[feat].fit(df_x[feat].values.reshape(-1, 1))

        df_x[feat] = normalizers_textblob_header[feat].transform(df_x[feat].values.
reshape(-1, 1))

    series_y = normalizers_textblob_header['OPEN'].transform(series_y.values.reshap
e(-1, 1))

    return df_x, series_y, normalizers_textblob_header

X_train_norm_textblob_header, \
y_train_norm_textblob_header, \
normalizers_textblob_header = minmax_scale_textblob_header(X_train_textblob_header,
                                                           y_train_textblob_header
                                                           )

X_valid_norm_textblob_header, \
y_valid_norm_textblob_header, \
_ = minmax_scale_textblob_header(X_valid_textblob_header,
                                 y_valid_textblob_header,
                                 normalizers_textblob_header=normalizers_textblob_h
eader
                                 )

X_test_norm_textblob_header, \
y_test_norm_textblob_header, \
_ = minmax_scale_textblob_header(X_test_textblob_header,
                                 y_test_textblob_header,
                                 normalizers_textblob_header=normalizers_textblob_h
eader
                                 )

def encode_cyclicals_textblob_header(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)

    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)
    return df_x

X_train_norm_textblob_header = encode_cyclicals_textblob_header(X_train_norm_textbl
ob_header)
X_valid_norm_textblob_header = encode_cyclicals_textblob_header(X_valid_norm_textbl
ob_header)
X_test_norm_textblob_header = encode_cyclicals_textblob_header(X_test_norm_textblob
_header)

# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_textblob_header = 60
FORECAST_DISTANCE_textblob_header = 30

segmenter_textblob_header = SegmentXYForecast(width=TIME_WINDOW_textblob_header,
                                              step=1,
                                              y_func=last,
                                              forecast=FORECAST_DISTANCE_textblob_h
eader
                                              )

X_train_rolled_textblob_header, \
y_train_rolled_textblob_header, \
_ = segmenter_textblob_header.fit_transform([X_train_norm_textblob_header.values],
                                            [y_train_norm_textblob_header.flatten
()]
                                            )

X_valid_rolled_textblob_header, \
y_valid_rolled_textblob_header, \
_ = segmenter_textblob_header.fit_transform([X_valid_norm_textblob_header.values],
                                            [y_valid_norm_textblob_header.flatten
()]
                                            )

X_test_rolled_textblob_header, \
y_test_rolled_textblob_header, \
_ = segmenter_textblob_header.fit_transform([X_test_norm_textblob_header.values],
                                            [y_test_norm_textblob_header.flatten()]
                                            )

shape_textblob_header = X_train_rolled_textblob_header.shape
X_train_flattened_textblob_header = X_train_rolled_textblob_header.reshape(shape_te
xtblob_header[0],
                                                                           shape_te
xtblob_header[1]*shape_textblob_header[2]
                                                                           )

X_train_flattened_textblob_header.shape
shape_textblob_header = X_valid_rolled_textblob_header.shape
X_valid_flattened = X_valid_rolled_textblob_header.reshape(shape_textblob_header
[0],
                                                           shape_textblob_header[1]
*shape_textblob_header[2]
                                                           )

# Random Forest
N_ESTIMATORS_textblob_header = 30
RANDOM_STATE_textblob_header = 452543634

RF_base_model_textblob_header = RandomForestRegressor(random_state=RANDOM_STATE_tex
tblob_header,
                                                      n_estimators=N_ESTIMATORS_tex
tblob_header,
                                                      n_jobs=-1,
                                                       verbose=100
                                                       )

RF_base_model_textblob_header.fit(X_train_flattened_textblob_header, y_train_rolled
_textblob_header)
print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_textblob_header = RF_base_model_textblob_header.predict(X
_valid_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_textblob_header = sqrt(mean_squared_error(y_valid_rolled_textblob_header,
                                                   RF_base_model_predictions_textbl
ob_header
                                                   )
                                )

print("Root mean squared error on valid:",rms_base_textblob_header)
print("Root mean squared error on valid inverse transformed from normalization:",no
rmalizers_textblob_header["OPEN"]
      .inverse_transform(np.array([rms_base_textblob_header]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_textblob_header = normalizers_textblob_header['OPEN']\
                                          .inverse_transform(np.array(RF_base_model
_predictions_textblob_header).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

### analysis with vader sentiment content
new_df_vader_content = concatenate_dataframe[['Date',
                                              'OPEN',
                                              'HIGH',
                                              'LOW',
                                              'CLOSE',
                                              'VOLUME',
                                              'compound_vader_articel_content']]

new_df_vader_content = new_df_vader_content.fillna(0)
# new_df_vader_content[['Date',
#                       'OPEN',
#                       'HIGH',
#                       'LOW',
#                       'CLOSE',
#                       'VOLUME',
#                       'compound_vader_articel_content']].astype(np.float64)

new_df_vader_content['Year'] = pd.DatetimeIndex(new_df_vader_content['Date']).year
new_df_vader_content['Month'] = pd.DatetimeIndex(new_df_vader_content['Date']).mont
h
new_df_vader_content['Day'] = pd.DatetimeIndex(new_df_vader_content['Date']).day
new_df_vader_content['Hour'] = pd.DatetimeIndex(new_df_vader_content['Date']).hour
new_df_vader_content['Minute'] = pd.DatetimeIndex(new_df_vader_content['Date']).min
ute
new_df_vader_content['Second'] = pd.DatetimeIndex(new_df_vader_content['Date']).sec
```

```
ond

new_df_vader_content = new_df_vader_content.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_vader_content = 0.1

X_train_vader_content, \
X_else_vader_content,\
y_train_vader_content, \
y_else_vader_content = train_test_split(new_df_vader_content,
                                        new_df_vader_content['OPEN'],
                                        test_size=valid_test_size_split_vader_conte
nt*2,
                                        shuffle=False)

X_valid_vader_content, \
X_test_vader_content, \
y_valid_vader_content, \
y_test_vader_content = train_test_split(X_else_vader_content,
                                        y_else_vader_content,
                                        test_size=0.5,
                                        shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)


# normalize data
def minmax_scale_vader_content(df_x, series_y, normalizers_vader_content = None):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME',
                          'compound_vader_articel_content']

    if not normalizers_vader_content:
        normalizers_vader_content = {}

    for feat in features_to_minmax:
        if feat not in normalizers_vader_content:
            normalizers_vader_content[feat] = MinMaxScaler()
            normalizers_vader_content[feat].fit(df_x[feat].values.reshape(-1, 1))

        df_x[feat] = normalizers_vader_content[feat].transform(df_x[feat].values.re
shape(-1, 1))

    series_y = normalizers_vader_content['OPEN'].transform(series_y.values.reshape
(-1, 1))

    return df_x, series_y, normalizers_vader_content

X_train_norm_vader_content, \
y_train_norm_vader_content, \
normalizers_vader_content = minmax_scale_vader_content(X_train_vader_content,
                                                       y_train_vader_content
                                                       )
```

```
X_valid_norm_vader_content, \
y_valid_norm_vader_content, \
_ = minmax_scale_vader_content(X_valid_vader_content,
                               y_valid_vader_content,
                               normalizers_vader_content=normalizers_vader_content
                               )

X_test_norm_vader_content, \
y_test_norm_vader_content, \
_ = minmax_scale_vader_content(X_test_vader_content,
                               y_test_vader_content,
                               normalizers_vader_content=normalizers_vader_content
                               )

def encode_cyclicals_vader_content(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)

    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)


    return df_x

X_train_norm_vader_content = encode_cyclicals_vader_content(X_train_norm_vader_cont
ent)
X_valid_norm_vader_content = encode_cyclicals_vader_content(X_valid_norm_vader_cont
ent)
X_test_norm_vader_content = encode_cyclicals_vader_content(X_test_norm_vader_conten
t)

# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_vader_content = 60
FORECAST_DISTANCE_vader_content = 30

segmenter_vader_content = SegmentXYForecast(width=TIME_WINDOW_vader_content,
                                            step=1,
                                            y_func=last,
                                            forecast=FORECAST_DISTANCE_vader_conten
t
                                            )

X_train_rolled_vader_content, \
y_train_rolled_vader_content, \
_ = segmenter_vader_content.fit_transform([X_train_norm_vader_content.values],
                                          [y_train_norm_vader_content.flatten()]
                                          )

X_valid_rolled_vader_content, \
y_valid_rolled_vader_content, \
_ = segmenter_vader_content.fit_transform([X_valid_norm_vader_content.values],
                                          [y_valid_norm_vader_content.flatten()]
                                          )

X_test_rolled_vader_content, \
y_test_rolled_vader_content, \
_ = segmenter_vader_content.fit_transform([X_test_norm_vader_content.values],
                                          [y_test_norm_vader_content.flatten()]
                                          )

shape_vader_content = X_train_rolled_vader_content.shape
X_train_flattened_vader_content = X_train_rolled_vader_content.reshape(shape_vader_
content[0],
                                                                       shape_vader_
content[1]*shape_vader_content[2]
                                                                       )

X_train_flattened_vader_content.shape
shape_vader_content = X_valid_rolled_vader_content.shape
X_valid_flattened = X_valid_rolled_vader_content.reshape(shape_vader_content[0],
                                                         shape_vader_content[1]*sha
pe_vader_content[2]
                                                         )

# Random Forest
N_ESTIMATORS_vader_content = 30
RANDOM_STATE_vader_content = 452543634

RF_base_model_vader_content = RandomForestRegressor(random_state=RANDOM_STATE_vader
_content,
                                                    n_estimators=N_ESTIMATORS_vader
_content,
                                                    n_jobs=-1,
                                                    verbose=100
                                                    )

RF_base_model_vader_content.fit(X_train_flattened_vader_content, y_train_rolled_vad
er_content)
print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_vader_content = RF_base_model_vader_content.predict(X_val
id_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_vader_content = sqrt(mean_squared_error(y_valid_rolled_vader_content,
                                                 RF_base_model_predictions_vader_co
ntent
                                                 )
                              )

print("Root mean squared error on valid:",rms_base_vader_content)
print("Root mean squared error on valid inverse transformed from normalization:",no
rmalizers_vader_content["OPEN"]
      .inverse_transform(np.array([rms_base_textblob_content]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_vader_content = normalizers_vader_content['OPEN']\
                                          .inverse_transform(np.array(RF_base_model
_predictions_vader_content).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

### analysis with vader header
new_df_vader_header = concatenate_dataframe[['Date',
                                             'OPEN',
                                             'HIGH',
                                             'LOW',
                                             'CLOSE',
                                             'VOLUME',
                                             'compound_vader_header']]

new_df_vader_header = new_df_vader_header.fillna(0)
# new_df_vader_header[['Date',
#                       'OPEN',
#                       'HIGH',
#                       'LOW',
#                       'CLOSE',
#                       'VOLUME',
#                       'compound_vader_header']].astype(np.float64)

new_df_vader_header['Year'] = pd.DatetimeIndex(new_df_vader_header['Date']).year
new_df_vader_header['Month'] = pd.DatetimeIndex(new_df_vader_header['Date']).month
new_df_vader_header['Day'] = pd.DatetimeIndex(new_df_vader_header['Date']).day
new_df_vader_header['Hour'] = pd.DatetimeIndex(new_df_vader_header['Date']).hour
new_df_vader_header['Minute'] = pd.DatetimeIndex(new_df_vader_header['Date']).minut
e
new_df_vader_header['Second'] = pd.DatetimeIndex(new_df_vader_header['Date']).secon
d

new_df_vader_header = new_df_vader_header.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_vader_header = 0.1

X_train_vader_header, \
X_else_vader_header,\
y_train_vader_header, \
y_else_vader_header = train_test_split(new_df_vader_header,
                                       new_df_vader_header['OPEN'],
                                       test_size=valid_test_size_split_vader_header
*2,
                                       shuffle=False)

X_valid_vader_header, \
X_test_vader_header, \
y_valid_vader_header, \
y_test_vader_header = train_test_split(X_else_vader_header,
                                       y_else_vader_header,
                                       test_size=0.5,
                                       shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)
```

```
# normalize data
def minmax_scale_vader_header(df_x, series_y, normalizers_vader_header = None):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME',
                          'compound_vader_header']

    if not normalizers_vader_header:
        normalizers_vader_header = {}

    for feat in features_to_minmax:
        if feat not in normalizers_vader_header:
            normalizers_vader_header[feat] = MinMaxScaler()
            normalizers_vader_header[feat].fit(df_x[feat].values.reshape(-1, 1))

        df_x[feat] = normalizers_vader_header[feat].transform(df_x[feat].values.res
hape(-1, 1))

    series_y = normalizers_vader_header['OPEN'].transform(series_y.values.reshape(-
1, 1))

    return df_x, series_y, normalizers_vader_header

X_train_norm_vader_header, \
y_train_norm_vader_header, \
normalizers_vader_header = minmax_scale_vader_header(X_train_vader_header,
                                                     y_train_vader_header
                                                     )

X_valid_norm_vader_header, \
y_valid_norm_vader_header, \
_ = minmax_scale_vader_header(X_valid_vader_header,
                              y_valid_vader_header,
                              normalizers_vader_header=normalizers_vader_header
                              )

X_test_norm_vader_header, \
y_test_norm_vader_header, \
_ = minmax_scale_vader_header(X_test_vader_header,
                              y_test_vader_header,
                              normalizers_vader_header=normalizers_vader_header
                              )

def encode_cyclicals_vader_header(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)
    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)

    return df_x

X_train_norm_vader_header = encode_cyclicals_vader_header(X_train_norm_vader_heade
r)
X_valid_norm_vader_header = encode_cyclicals_vader_header(X_valid_norm_vader_heade
r)
X_test_norm_vader_header = encode_cyclicals_vader_header(X_test_norm_vader_header)

# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_vader_header = 60
FORECAST_DISTANCE_vader_header = 30

segmenter_vader_header = SegmentXYForecast(width=TIME_WINDOW_vader_header,
                                           step=1,
                                           y_func=last,
                                           forecast=FORECAST_DISTANCE_vader_header
                                           )

X_train_rolled_vader_header, \
y_train_rolled_vader_header, \
_ = segmenter_vader_header.fit_transform([X_train_norm_vader_header.values],
                                         [y_train_norm_vader_header.flatten()]
                                         )

X_valid_rolled_vader_header, \
y_valid_rolled_vader_header, \
_ = segmenter_vader_header.fit_transform([X_valid_norm_vader_header.values],
                                         [y_valid_norm_vader_header.flatten()]
                                         )

X_test_rolled_vader_header, \
y_test_rolled_vader_header, \
_ = segmenter_vader_header.fit_transform([X_test_norm_vader_header.values],
                                         [y_test_norm_vader_header.flatten()]
                                         )

shape_vader_header = X_train_rolled_vader_header.shape
X_train_flattened_vader_header = X_train_rolled_vader_header.reshape(shape_vader_he
ader[0],
                                                                    shape_vader_he
ader[1]*shape_vader_header[2]
                                                                    )

X_train_flattened_vader_header.shape
shape_vader_header = X_valid_rolled_vader_header.shape
X_valid_flattened = X_valid_rolled_vader_header.reshape(shape_vader_header[0],
                                                       shape_vader_header[1]*shape
_vader_header[2]
                                                       )

# Random Forest
```

```
N_ESTIMATORS_vader_header = 30
RANDOM_STATE_vader_header = 452543634

RF_base_model_vader_header = RandomForestRegressor(random_state=RANDOM_STATE_vader_
header,
                                                   n_estimators=N_ESTIMATORS_vader_
header,
                                                   n_jobs=-1,
                                                   verbose=100
                                                   )

RF_base_model_vader_header.fit(X_train_flattened_vader_header, y_train_rolled_vader
_header)
print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_vader_header = RF_base_model_vader_header.predict(X_valid
_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_vader_header = sqrt(mean_squared_error(y_valid_rolled_vader_header,
                                                RF_base_model_predictions_vader_hea
der
                                                )
                             )

print("Root mean squared error on valid:", rms_base_vader_header)
print("Root mean squared error on valid inverse transformed from normalization:", n
ormalizers_vader_header["OPEN"]
      .inverse_transform(np.array([rms_base_vader_header]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_vader_header = normalizers_vader_header['OPEN']\
                                         .inverse_transform(np.array(RF_base_model
_predictions_vader_header).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

### analysis with without semantics
new_df_without_semantics = concatenate_dataframe[['Date',
                                                  'OPEN',
                                                  'HIGH',
                                                  'LOW',
                                                  'CLOSE',
                                                  'VOLUME']]

new_df_without_semantics = new_df_without_semantics.fillna(0)
# new_df_without_semantics[['Date',
#                            'OPEN',
#                            'HIGH',
#                            'LOW',
#                            'CLOSE',
#                            'VOLUME']].astype(np.float64)

new_df_without_semantics['Year'] = pd.DatetimeIndex(new_df_without_semantics['Date
']).year
```

```
new_df_without_semantics['Month'] = pd.DatetimeIndex(new_df_without_semantics['Date
']).month
new_df_without_semantics['Day'] = pd.DatetimeIndex(new_df_without_semantics['Date
']).day
new_df_without_semantics['Hour'] = pd.DatetimeIndex(new_df_without_semantics['Date
']).hour
new_df_without_semantics['Minute'] = pd.DatetimeIndex(new_df_without_semantics['Dat
e']).minute
new_df_without_semantics['Second'] = pd.DatetimeIndex(new_df_without_semantics['Dat
e']).second

new_df_without_semantics = new_df_without_semantics.drop(['Date'], axis=1)

# train, valid, test split
valid_test_size_split_without_semantics = 0.1

X_train_without_semantics, \
X_else_without_semantics,\
y_train_without_semantics, \
y_else_without_semantics = train_test_split(new_df_without_semantics,
                                            new_df_without_semantics['OPEN'],
                                            test_size=valid_test_size_split_without
_semantics*2,
                                            shuffle=False)

X_valid_without_semantics, \
X_test_without_semantics, \
y_valid_without_semantics, \
y_test_without_semantics = train_test_split(X_else_without_semantics,
                                            y_else_without_semantics,
                                            test_size=0.5,
                                            shuffle=False)

warnings.filterwarnings(action='ignore', category=DataConversionWarning)


# normalize data
def minmax_scale_without_semantics(df_x, series_y, normalizers_without_semantics =
None):
    features_to_minmax = ['Year',
                          'Month',
                          'Day',
                          'Hour',
                          'Minute',
                          'Second',
                          'OPEN',
                          'HIGH',
                          'LOW',
                          'CLOSE',
                          'VOLUME']

    if not normalizers_without_semantics:
        normalizers_without_semantics = {}

    for feat in features_to_minmax:
        if feat not in normalizers_without_semantics:
            normalizers_without_semantics[feat] = MinMaxScaler()
            normalizers_without_semantics[feat].fit(df_x[feat].values.reshape(-1,
1))

        df_x[feat] = normalizers_without_semantics[feat].transform(df_x[feat].value
s.reshape(-1, 1))

    series_y = normalizers_without_semantics['OPEN'].transform(series_y.values.resh
ape(-1, 1))

    return df_x, series_y, normalizers_without_semantics

X_train_norm_without_semantics, \
y_train_norm_without_semantics, \
normalizers_without_semantics = minmax_scale_without_semantics(X_train_without_sema
ntics,
                                                               y_train_without_sema
ntics
                                                               )

X_valid_norm_without_semantics, \
y_valid_norm_without_semantics, \
_ = minmax_scale_without_semantics(X_valid_without_semantics,
                                   y_valid_without_semantics,
                                   normalizers_without_semantics=normalizers_withou
t_semantics
                                   )

X_test_norm_without_semantics, \
y_test_norm_without_semantics, \
_ = minmax_scale_without_semantics(X_test_without_semantics,
                                   y_test_without_semantics,
                                   normalizers_without_semantics=normalizers_withou
t_semantics
                                   )

def encode_cyclicals_without_semantics(df_x):
    # "month","day","hour", "cdbw", "dayofweek"

    #DIRECTIONS = {"N": 1.0, "NE": 2.0, "E": 3.0, "SE": 4.0, "S": 5.0, "SW": 6.0, "
W": 7.0, "NW": 8.0, "cv": np.nan}

    df_x['month_sin'] = np.sin(2 * np.pi * df_x.Month / 12)
    df_x['month_cos'] = np.cos(2 * np.pi * df_x.Month / 12)
    df_x.drop('Month', axis=1, inplace=True)

    df_x['day_sin'] = np.sin(2 * np.pi * df_x.Day / 31)
    df_x['day_cos'] = np.cos(2 * np.pi * df_x.Day / 31)
    df_x.drop('Day', axis=1, inplace=True)

    df_x['hour_sin'] = np.sin(2 * np.pi * df_x.Hour / 24)
    df_x['hour_cos'] = np.cos(2 * np.pi * df_x.Hour / 24)
    df_x.drop('Hour', axis=1, inplace=True)

    df_x['min_sin'] = np.sin(2 * np.pi * df_x.Minute / 60)
    df_x['min_cos'] = np.cos(2 * np.pi * df_x.Minute / 60)
    df_x.drop('Minute', axis=1, inplace=True)

    df_x['sec_sin'] = np.sin(2 * np.pi * df_x.Second / 60)
    df_x['sec_cos'] = np.cos(2 * np.pi * df_x.Second / 60)
    df_x.drop('Second', axis=1, inplace=True)


    return df_x

X_train_norm_without_semantics = encode_cyclicals_without_semantics(X_train_norm_wi
thout_semantics)
X_valid_norm_without_semantics = encode_cyclicals_without_semantics(X_valid_norm_wi
thout_semantics)
X_test_norm_without_semantics = encode_cyclicals_without_semantics(X_test_norm_with
out_semantics)
```

```
# Creating target (y) and "windows" (X) for modeling
TIME_WINDOW_without_semantics = 60
FORECAST_DISTANCE_without_semantics = 30

segmenter_without_semantics = SegmentXYForecast(width=TIME_WINDOW_without_semantic
s,
                                                step=1,
                                                y_func=last,
                                                forecast=FORECAST_DISTANCE_without_
semantics
                                                )

X_train_rolled_without_semantics, \
y_train_rolled_without_semantics, \
_ = segmenter_without_semantics.fit_transform([X_train_norm_without_semantics.value
s],
                                              [y_train_norm_without_semantics.flatt
en()]
                                              )

X_valid_rolled_without_semantics, \
y_valid_rolled_without_semantics, \
_ = segmenter_without_semantics.fit_transform([X_valid_norm_without_semantics.value
s],
                                              [y_valid_norm_without_semantics.flatt
en()]
                                              )

X_test_rolled_without_semantics, \
y_test_rolled_without_semantics, \
_ = segmenter_without_semantics.fit_transform([X_test_norm_without_semantics.value
s],
                                              [y_test_norm_without_semantics.flatte
n()]
                                              )

shape_without_semantics = X_train_rolled_without_semantics.shape
X_train_flattened_without_semantics = X_train_rolled_without_semantics.reshape(shap
e_without_semantics[0],
                                                                               shap
e_without_semantics[1]*shape_without_semantics[2]
                                                                               )

X_train_flattened_without_semantics.shape
shape_without_semantics = X_valid_rolled_without_semantics.shape
X_valid_flattened = X_valid_rolled_without_semantics.reshape(shape_without_semantic
s[0],
                                                             shape_without_semantic
s[1]*shape_without_semantics[2]
                                                             )

# Random Forest
N_ESTIMATORS_without_semantics = 30
RANDOM_STATE_without_semantics = 452543634

RF_base_model_without_semantics = RandomForestRegressor(random_state=RANDOM_STATE_w
ithout_semantics,
                                                        n_estimators=N_ESTIMATORS_w
ithout_semantics,
                                                        n_jobs=-1,
                                                        verbose=100
                                                        )

RF_base_model_without_semantics.fit(X_train_flattened_without_semantics, y_train_ro
lled_without_semantics)
print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_without_semantics = RF_base_model_without_semantics.predi
ct(X_valid_flattened)
print(' ')
print("----------------------------------------------------------------")
print(' ')
rms_base_without_semantics = sqrt(mean_squared_error(y_valid_rolled_without_semanti
cs,
                                                     RF_base_model_predictions_with
out_semantics
                                                     )
                                  )

print("Root mean squared error on valid:", rms_base_without_semantics)
print("Root mean squared error on valid inverse transformed from normalization:", n
ormalizers_without_semantics["OPEN"]
      .inverse_transform(np.array([rms_base_without_semantics]).reshape(-1, 1)))

print(' ')
print("----------------------------------------------------------------")
print(' ')
RF_base_model_predictions_without_semantics = normalizers_without_semantics['OPEN
']\
                                          .inverse_transform(np.array(RF_base_model
_predictions_without_semantics).reshape(-1, 1))
print(' ')
print("----------------------------------------------------------------")
print(' ')

print(' ')
print("----------------------------------------------------------------")
print(' ')

plt.figure(figsize=(10,5))
plt.plot(RF_base_model_predictions_flair_content, color='green', label='Predicted F
iatchrysler Stock Price with flair content analysis')
plt.plot(RF_base_model_predictions_flair_header, color='red', label='Predicted Fiat
chrysler Stock Price with flair header analysis')
plt.plot(RF_base_model_predictions_textblob_content, color='orange', label='Predict
ed Fiatchrysler Stock Price with textblob content analysis')
plt.plot(RF_base_model_predictions_textblob_header, color='blue', label='Predicted
Fiatchrysler Stock Price with textblob header analysis')
plt.plot(RF_base_model_predictions_vader_content, color='cyan', label='Predicted Fi
atchrysler Stock Price with vader content analysis')
plt.plot(RF_base_model_predictions_vader_header, color='magenta', label='Predicted
Fiatchrysler Stock Price with vader header analysis')
plt.plot(RF_base_model_predictions_without_semantics, color='yellow', label='Predic
ted Fiatchrysler Stock Price without semantics analysis')
plt.title('Fiatchrysler Stock Price Prediction')
plt.xlabel('Time')
plt.ylabel('Fiatchrysler Stock Price')
plt.legend(loc='upper center', bbox_to_anchor=(0.5, -0.005), borderaxespad=8)

date_today = str(datetime.now().strftime("%Y%m%d"))
#plt.savefig(r'C:\Users\victo\Master_Thesis\stockprice_prediction\RandomForest_base
_model\fiatchrysler\minutely\prediction_fiatchrysler_' + date_today + '.png',
#             bbox_inches="tight",
#             dpi=100,
#             pad_inches=1.5)
plt.show()
print("Root mean squared error flair content on valid:", rms_base_flair_content)
print("Root mean squared error on flair content valid inverse transformed from norm
alization:",
      normalizers_flair_content["OPEN"].inverse_transform(np.array([rms_base_flair_
content]).reshape(-1, 1)))
print("Root mean squared error on flair header valid:", rms_base_flair_header)
print("Root mean squared error on valid inverse transformed from normalization:",
      normalizers_flair_header["OPEN"].inverse_transform(np.array([rms_base_flair_h
eader]).reshape(-1, 1)))
print("Root mean squared error on textblob content valid:", rms_base_textblob_conte
nt)
print("Root mean squared error on valid inverse transformed from normalization:",
      normalizers_textblob_content["OPEN"].inverse_transform(np.array([rms_base_tex
tblob_content]).reshape(-1, 1)))
print("Root mean squared error on textblob header valid:", rms_base_textblob_heade
r)
print("Root mean squared error on valid inverse transformed from normalization:",
      normalizers_textblob_header["OPEN"].inverse_transform(np.array([rms_base_text
blob_header]).reshape(-1, 1)))
print("Root mean squared error on vader vader content valid:", rms_base_vader_conte
nt)
print("Root mean squared error on valid inverse transformed from normalization:",
      normalizers_vader_content["OPEN"].inverse_transform(np.array([rms_base_vader_
content]).reshape(-1, 1)))
print("Root mean squared error on vader header valid:", rms_base_vader_header)
print("Root mean squared error on valid inverse transformed from normalization:",
      normalizers_vader_header["OPEN"].inverse_transform(np.array([rms_base_vader_h
eader]).reshape(-1, 1)))
print("Root mean squared error on valid:", rms_base_without_semantics)
print("Root mean squared error on valid inverse transformed from normalization:",
      normalizers_without_semantics["OPEN"].inverse_transform(np.array([rms_base_wi
thout_semantics]).reshape(-1, 1)))

print('Run is finished and plot is saved!')
```