W11_SVR_Consumption

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1 SVR model Energy Consumption

Een SVR model ter referentie voor bij het Multi Layer Perceptron model.

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1.0.1 Importeer modules

```
import numpy as np
import pandas as pd
from tqdm import tqdm
import matplotlib.pyplot as plt
import glob
from sklearn.metrics import r2_score
import sklearn
from sklearn.model_selection import train_test_split
from sklearn.model_selection import cross_validate
import seaborn as sns
from sklearn import linear_model
from sklearn.svm import SVR
from sklearn.preprocessing import StandardScaler
```

1.1 Data klaarmaken

```
[57]: %matplotlib notebook
df = pd.read_pickle('28_df_consumption_weekday')

#scale the data
scaler = StandardScaler()
scaler.fit(df.loc[:,~df.columns.isin(["consumption"])])
scaled_dataX = scaler.transform(df.loc[:,~df.columns.isin(["consumption"])]).

+tolist()
```

```
datay = np.array(df["consumption"].tolist())
      #split the data
      train_X = np.array(scaled_dataX[0:24*9])
      train_y = datay[0:24*9].reshape(-1,1)
      valid_X = np.array(scaled_dataX[24*9:24*10])
      valid_y = datay[24*9:24*10].reshape(-1,1)
      ## sccaler op train fitten
      # test X = scaled dataX[24*10:24*11]
      \# test_y = datay[24*10:24*11].reshape(-1,1)
      # scaler op test fitten
[58]: df.head()
[58]:
                           consumption day_0 day_1
                                                       day_2 day_3 day_4
                                                                             day_5 \setminus
      2019-01-01 23:00:00
                                 0.209
                                             0
                                                    1
                                                    0
                                                                  0
                                                                                 0
      2019-01-02 00:00:00
                                 0.253
                                             0
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      2019-01-02 01:00:00
                                 0.332
                                             0
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                                                           1
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      2019-01-02 02:00:00
                                 0.597
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                                                           1
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      2019-01-02 03:00:00
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                                                                          0.810
                           cons_T-17 cons_T-18 cons_T-19 cons_T-20
                                                                        cons_T-21 \
                               0.733
                                           0.799
                                                      0.274
      2019-01-01 23:00:00
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      2019-01-02 00:00:00
                                                                             1.526
      2019-01-02 01:00:00
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      2019-01-02 02:00:00
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                                                                 0.733
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      2019-01-02 03:00:00
                               0.435
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                                                      0.255
                                                                 1.041
                                                                             0.733
                           cons_T-22 cons_T-23 cons_T-24
      2019-01-01 23:00:00
                               0.264
                                           0.402
                                                      0.622
      2019-01-02 00:00:00
                               0.163
                                           0.264
                                                      0.402
```

[5 rows x 32 columns]

2019-01-02 01:00:00

2019-01-02 02:00:00

2019-01-02 03:00:00

0.163

1.526

0.274

0.264

0.163

1.526

1.526

0.274

0.799

1.2 SVR Trainen

```
[59]: #train the model:
    svr = SVR(C=1)
    svr.fit(train_X,train_y)

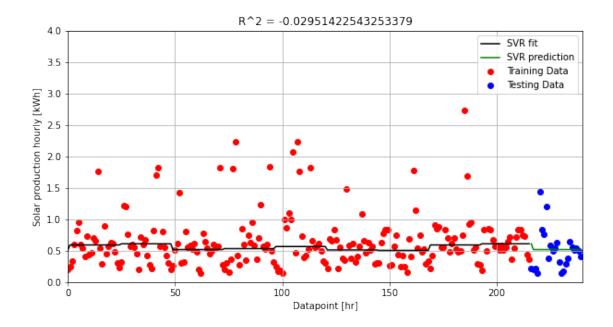
#make data plot ready:
    valid = np.arange(train_X.shape[0], train_X.shape[0]+valid_X.shape[0])
    train = np.arange(train_X.shape[0])

/opt/jupyterhub/anaconda/lib/python3.7/site-
    packages/sklearn/utils/validation.py:72: DataConversionWarning: A column-vector
    y was passed when a 1d array was expected. Please change the shape of y to
    (n_samples, ), for example using ravel().
```

1.2.1 Plotje

return f(**kwargs)

```
[62]: %matplotlib inline
      plt.subplots(figsize=(10,5))
      plt.plot(train,svr.predict(train_X),color='black',label="SVR fit")
      plt.plot(valid,svr.predict(valid X),color='g',label="SVR prediction")
      plt.scatter(train,train_y,color="r",label="Training Data")
      plt.scatter(valid,valid_y,color="b",label="Testing Data")
      #Nice layout:
      plt.xlabel("Datapoint [hr]")
      plt.ylabel("Solar production hourly [kWh]")
      plt.legend(loc="upper right")
      plt.title("R^2 = "+str(svr.score(valid_X,valid_y)))
      plt.ylim([0,4])
      #plt.xlim([X train.shape[0]-(24*5), X train.shape[0]+X test.shape[0]])
      plt.xlim([0,train_X.shape[0]+valid_X.shape[0]])
      plt.grid()
      plt.show()
      #print de score:
      print("R2-score test data:")
      print(svr.score(valid X, valid y))
```



R2-score test data: -0.02951422543253379

1.3 Evaluatie

```
[61]: %matplotlib notebook
  plt.scatter(valid_y,svr.predict(valid_X))
  plt.xlabel("y [kWh]")
  plt.ylabel("y_hat [kWh]")
  #plt.legend(loc="upper right")
  plt.title("$R^2$ = %.2f" % (r2_score(valid_y,svr.predict(valid_X))))
  plt.plot(plt.xlim(), plt.xlim(), ls="--", c='r', label="$y$=$\hat{y}$")
  plt.grid()
  plt.show()

<IPython.core.display.Javascript object>
  <IPython.core.display.HTML object>
```