W12 NeuralNetwork-1uur

January 12, 2021

1 Een simpel Neuraal Netwerk in PyTorch

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aangemaakt: 25/11/2020

1.1 importeer alle modules:

```
[1]: import random
     import time
     import torch
     import torchvision
     import torchvision.transforms as transforms
     import numpy as np
     import pandas as pd
     import math
     import random
     import torch.nn as nn
     import torch.nn.functional as F
     import torch.optim as optim
     import seaborn as sns
     import matplotlib.pyplot as plt
     from tqdm import tqdm
     from IPython.display import clear_output
     from sklearn.metrics import r2_score
     from sklearn.metrics import mean_squared_error
     from sklearn.preprocessing import StandardScaler
     from sklearn.feature_selection import SelectKBest, mutual_info_classif
     from sklearn.metrics import mean_absolute_error
     import wandb
     # hier random seeds mee geven
     random.seed(1337)
```

```
torch.manual_seed(1337)

%matplotlib inline
%config InlineBackend.print_figure_kwargs={'facecolor' : "w"}

# CUDA initialisation
```

[2]: # CUDA initialisation

ngpu = torch.cuda.device_count() # number of available gpus

device = torch.device("cuda:4") if (torch.cuda.is_available() and ngpu > 0)

→else "cpu" #cuda:0 for gpu 0, cuda:4 for gpu 5

torch.backends.cudnn.benchmark=True # Uses cudnn auto-tuner to find the best

→algorithm to use for your hardware

1.2 Laad de dataframe in:

```
[3]: df = pd.read_pickle('/home/18005152/notebooks/zero/Data:/modelData/_v01_1') df = df['2019-09-02':'2019-11-29']
```

[4]: df.std()

[4]: consumption 0.370114
 cons_T-24 0.370994
 cons_T-48 0.372632
 cons_T-72 0.372919
 cons_T-168 0.364826
 day_mean 0.130623
 week_mean 0.104796
 dtype: float64

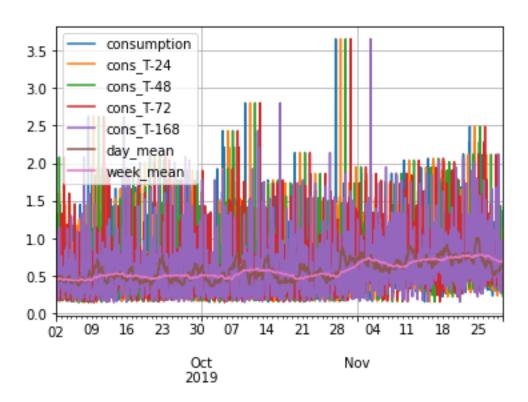
1.3 Laat de data zien:

[5]: %matplotlib inline
 df.plot()
 plt.grid()
 df.head()

[5]: consumption cons_T-24 cons_T-48 cons_T-72 cons_T-168 \ 2019-09-02 00:00:00 0.183 0.170 0.1790 0.604 0.227 2019-09-02 01:00:00 0.171 0.2090 0.239 0.563 0.214 2019-09-02 02:00:00 0.290 0.246 1.3335 0.780 1.099 2019-09-02 03:00:00 0.302 0.422 0.5510 1.101 0.768 2019-09-02 04:00:00 0.234 0.485 0.744 1.188 0.7450

day_mean week_mean 2019-09-02 00:00:00 0.51006 0.460259

```
2019-09-02 01:00:00 0.50974 0.459872
2019-09-02 02:00:00 0.51122 0.460093
2019-09-02 03:00:00 0.47476 0.455424
2019-09-02 04:00:00 0.50024 0.458321
```



1.4 Train Validate Test split (ok)

```
[6]: # lijst = ["consumption"]
# for i in range(0,24):
# lijst.append("hour_"+str(i))
# print(lijst)
```

```
[7]: #scale the data
#X:
scalerx = StandardScaler()
lijst = ["consumption"]
# for i in range(0,24):
# lijst.append("hour_"+str(i))
scalerx.fit(df.loc[:,~df.columns.isin(lijst)])
scaled_dataX = scalerx.transform(df.loc[:,~df.columns.isin(lijst)]).tolist()
# for j in range(0,24):
# for i in range(0,len(df["hour_"+str(j)].tolist())):
```

```
scaled\_dataX[i][j] = df["hour\_"+str(j)].tolist()[i]
#Y:
scalery = StandardScaler()
scalery.fit(df.loc[:,df.columns.isin(["consumption"])])
datay = scalery.transform(df.loc[:,df.columns.isin(["consumption"])])
start = 0
week = 7*1*24
end train = 7*4*24
end valid = 7*5*24
end_test = 7*6*24
#split the data
train_X = scaled_dataX[start:end_train]
train_y = datay[start:end_train].reshape(-1,1)
valid_X = scaled_dataX[end_train+week:end_valid+week]
valid_y = datay[end_train+week:end_valid+week].reshape(-1,1)
test_X = scaled_dataX[end_valid+week:end_test+week]
test_y = datay[end_valid+week:end_test+week].reshape(-1,1)
#Make tensors from the numpy arrays.
train X t = torch.from numpy(np.array(train X)).to(device).float()
train_y_t = torch.from_numpy(np.array(train_y)).to(device).float()
valid_X_t = torch.from_numpy(np.array(valid_X)).to(device).float()
valid_y_t = torch.from_numpy(np.array(valid_y)).to(device).float()
test_X_t = torch.from_numpy(np.array(test_X)).to(device).float()
test_y_t = torch.from_numpy(np.array(test_y)).to(device).float()
#Tensor Datasets
train_set = torch.utils.data.TensorDataset(train_X_t, train_y_t)
test_set = torch.utils.data.TensorDataset(valid_y_t, valid_X_t)
#Tensor DataLoaders
train_loader = torch.utils.data.DataLoader(train_set, batch_size=64,_u
⇒shuffle=False, num_workers = 0)#, pin_memory=True)
test_loader = torch.utils.data.DataLoader(test_set, batch_size=64,__
 →shuffle=False, num_workers = 0)#, pin_memory=True)
```

1.5 Neuraal Netwerk:

```
[8]: #Parameters:
     layerSize = 128
     outputSize = 1
     featureSize = train_X_t.shape[1]
     relu = nn.ReLU()
     #class maken voor NN
     class Net(nn.Module):
         def __init__(self):
             super(Net, self).__init__()
             self.fc1 = nn.Linear(featureSize, layerSize)
             self.fc2 = nn.Linear(layerSize, outputSize)
         def forward(self, x):
             x = relu(self.fc1(x))
             x = self.fc2(x)
             return x
[9]: #make model:
     model = Net().to(device)
    model.float()
```

```
[9]: Net(
          (fc1): Linear(in_features=6, out_features=128, bias=True)
          (fc2): Linear(in_features=128, out_features=1, bias=True)
)
```

1.6 Train het Neuraal Netwerk:

```
[13]: #Training parameters:
    train_for = 100
    show_every = 2
    optimizer = optim.Adam(model.parameters(),lr=3e-4)
    criterion = nn.SmoothL1Loss()
    criterion0 = nn.MSELoss()

#initialize:
    ite = 0
    epochs = range(1,train_for)
    alijst = []; blijst = []

#Learning loop:
    for i in (epochs):
        btime = time.time()
```

```
model.train()
   for batch_idx, data_target in enumerate(train_loader):
       data = data_target[0]
       target = data_target[1]
       data = data.view(-1, data.shape[1])
       optimizer.zero_grad()
       output = model(data)
       loss = 5*criterion(output, target)# + 3*criterion0(output, target)
       print(input)
       loss.backward()
       optimizer.step()
   etime = time.time()
   model.eval()
   if (ite%show_every) == 0:
       y = valid_y_t.cpu().detach().numpy()
       yhat = model(valid_X_t.float()).cpu().detach().numpy()
       # hier loss appenden ipv nog een losse berekening
       alijst.append(mean_absolute_error(train_y_t.cpu().detach().
→numpy(),model(train_X_t.float()).cpu().detach().numpy()))
       blijst.append(mean_absolute_error(y,yhat))
       y = train_y_t.cpu().detach().numpy()
       yhat = model(train_X_t.float()).cpu().detach().numpy()
       # R^2-score:
       r2 = r2_score(yhat, y)
       # RMSE:
       rmse = np.sqrt(mean_squared_error(yhat, y))
       # MAPE:
       actual, pred = np.array(y), np.array(yhat)
       mape = np.mean(np.abs((actual - pred) / actual)) * 100
       # MAE:
       mae = mean_absolute_error(yhat, y)
       #print:
       print('Epoch: %d\t R\u00b2: %.2f\t RMSE: %.2f\t MAPE: %.2f\t MAE: %.
→2f\t Looptime: %.3f s' % (ite,r2,rmse,mape,mae,etime-btime))
   ite+=1
```

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Epoch: 0 R²: -2.82 RMSE: 0.81 MAPE: 107.92 MAE: 0.47

Looptime: 0.035 s

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Epoch: 2 R²: -2.82 RMSE: 0.81 MAPE: 108.38 MAE: 0.48

Looptime: 0.030 s

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Epoch: 4 R²: -2.80 RMSE: 0.81 MAPE: 108.35 MAE: 0.48

Looptime: 0.030 s

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Epoch: 6 R²: -2.80 RMSE: 0.81 MAPE: 108.37 MAE: 0.48

Looptime: 0.035 s

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Epoch: 8 R²: -2.79 RMSE: 0.81 MAPE: 108.38 MAE: 0.48

Looptime: 0.028 s

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Epoch: 10 R²: -2.79 RMSE: 0.81 MAPE: 108.40 MAE: 0.48

Looptime: 0.029 s

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Epoch: 12 R²: -2.78 RMSE: 0.81 MAPE: 108.38 MAE: 0.48

Looptime: 0.029 s

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Epoch: 14 R²: -2.78 RMSE: 0.81 MAPE: 108.37 MAE: 0.48

Looptime: 0.029 s

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Epoch: 16 R²: -2.77 RMSE: 0.81 MAPE: 108.35 MAE: 0.48

Looptime: 0.028 s

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Epoch: 18 R^2 : -2.77 RMSE: 0.81 MAPE: 108.33 MAE: 0.48

Looptime: 0.029 s

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Epoch: 20 R²: -2.77 RMSE: 0.81 MAPE: 108.33 MAE: 0.48

Looptime: 0.028 s

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Epoch: 22 R²: -2.76 RMSE: 0.81 MAPE: 108.32 MAE: 0.47

Looptime: 0.029 s

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Epoch: 24 R²: -2.76 RMSE: 0.81 MAPE: 108.31 MAE: 0.47

Looptime: 0.028 s

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Epoch: 26 R²: -2.75 RMSE: 0.81 MAPE: 108.31 MAE: 0.47

Looptime: 0.029 s

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Epoch: 28 R²: -2.75 RMSE: 0.81 MAPE: 108.30 MAE: 0.47

Looptime: 0.028 s

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Epoch: 30 R²: -2.74 RMSE: 0.81 MAPE: 108.29 MAE: 0.47

Looptime: 0.028 s

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Epoch: 32 R²: -2.74 RMSE: 0.81 MAPE: 108.26 MAE: 0.47

Looptime: 0.028 s

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Epoch: 34 R²: -2.73 RMSE: 0.81 MAPE: 108.26 MAE: 0.47

Looptime: 0.028 s

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Epoch: 36 R²: -2.73 RMSE: 0.81 MAPE: 108.26 MAE: 0.47

Looptime: 0.028 s

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Epoch: 38 R²: -2.72 RMSE: 0.81 MAPE: 108.24 MAE: 0.47

Looptime: 0.028 s

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Epoch: 40 R²: -2.72 RMSE: 0.81 MAPE: 108.23 MAE: 0.47

Looptime: 0.031 s

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Epoch: 42 R²: -2.72 RMSE: 0.81 MAPE: 108.22 MAE: 0.47

Looptime: 0.028 s

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Epoch: 44 R²: -2.71 RMSE: 0.81 MAPE: 108.22 MAE: 0.47

Looptime: 0.028 s

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Epoch: 46 R²: -2.71 RMSE: 0.81 MAPE: 108.20 MAE: 0.47

Looptime: 0.030 s

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Epoch: 48 R²: -2.70 RMSE: 0.81 MAPE: 108.19 MAE: 0.47

Looptime: 0.028 s

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Epoch: 50 R²: -2.70 RMSE: 0.80 MAPE: 108.17 MAE: 0.47

Looptime: 0.028 s

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Epoch: 52 R²: -2.70 RMSE: 0.80 MAPE: 108.14 MAE: 0.47

Looptime: 0.028 s

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Epoch: 54 R²: -2.69 RMSE: 0.80 MAPE: 108.13 MAE: 0.47

Looptime: 0.028 s

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Epoch: 56 R²: -2.69 RMSE: 0.80 MAPE: 108.11 MAE: 0.47

Looptime: 0.028 s

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Epoch: 58 R²: -2.68 RMSE: 0.80 MAPE: 108.10 MAE: 0.47

Looptime: 0.028 s

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Epoch: 60 R²: -2.68 RMSE: 0.80 MAPE: 108.10 MAE: 0.47

Looptime: 0.029 s

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Epoch: 62 R²: -2.67 RMSE: 0.80 MAPE: 108.08 MAE: 0.47

Looptime: 0.027 s

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Epoch: 64 R²: -2.67 RMSE: 0.80 MAPE: 108.06 MAE: 0.47

Looptime: 0.028 s

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Epoch: 66 R²: -2.66 RMSE: 0.80 MAPE: 108.03 MAE: 0.47

Looptime: 0.028 s

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Epoch: 68 R²: -2.66 RMSE: 0.80 MAPE: 108.01 MAE: 0.47

Looptime: 0.028 s

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Epoch: 70 R²: -2.66 RMSE: 0.80 MAPE: 108.00 MAE: 0.47

Looptime: 0.028 s

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Epoch: 72 R²: -2.65 RMSE: 0.80 MAPE: 107.95 MAE: 0.47

Looptime: 0.028 s

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0x7fb3b5b73ac8>>

Epoch: 74 R²: -2.65 RMSE: 0.80 MAPE: 107.94 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 76 R²: -2.64 RMSE: 0.80 MAPE: 107.90 MAE: 0.47

Looptime: 0.029 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 78 R²: -2.64 RMSE: 0.80 MAPE: 107.86 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 80 R²: -2.63 RMSE: 0.80 MAPE: 107.86 MAE: 0.47

Looptime: 0.029 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 82 R²: -2.63 RMSE: 0.80 MAPE: 107.83 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 84 R²: -2.62 RMSE: 0.80 MAPE: 107.81 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 86 R²: -2.62 RMSE: 0.80 MAPE: 107.80 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 88 R²: -2.61 RMSE: 0.80 MAPE: 107.79 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 90 R²: -2.61 RMSE: 0.80 MAPE: 107.74 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 92 R²: -2.61 RMSE: 0.80 MAPE: 107.71 MAE: 0.47

Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 94 R²: -2.60 RMSE: 0.80 MAPE: 107.67 MAE: 0.47

Looptime: 0.029 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 96 R²: -2.60 RMSE: 0.80 MAPE: 107.64 MAE: 0.47

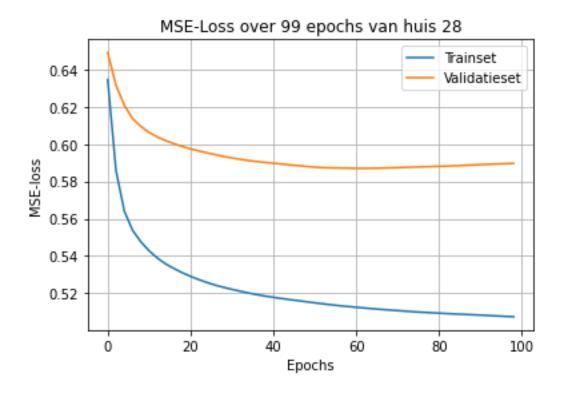
Looptime: 0.028 s

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

```
<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>
```

<bound method Kernel.raw_input of <ipykernel.ipkernel.IPythonKernel object at
0x7fb3b5b73ac8>>

Epoch: 98 R²: -2.60 RMSE: 0.80 MAPE: 107.59 MAE: 0.47 Looptime: 0.028 s



1.7 Test het Neuraal Netwerk:

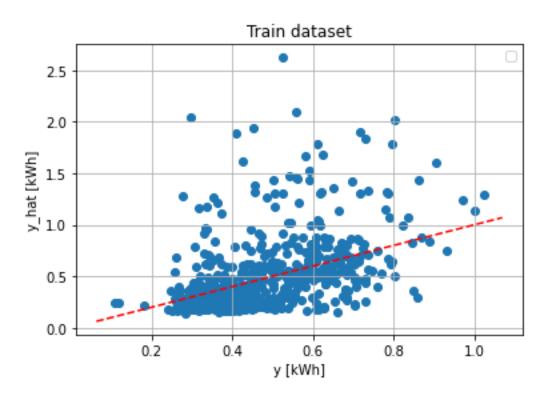
Geïmplementeerde validatie methoden: - R^2 - RMSE - MAPE - MAE (L1-Loss)

Train evaluation

```
#zet de voorspelde tegen de werkelijke waarde uit.
%matplotlib inline
plt.scatter(yhat,y)
plt.xlabel("y [kWh]")
plt.ylabel("y_hat [kWh]")
plt.legend(loc="upper right")
plt.title('Train dataset')
plt.plot(plt.xlim(), plt.xlim(), ls="--", c='r', label="$y$=$\hat{y}$")
# plt.xlim([0.1,1])
# plt.ylim([0.1,1])
print('R\u00b2: \t'.2f \nRMSE: \t'.2f \nMAPE: \t'.2f \nMAE: \t'.2f' %_\u00d3
\u00c4(r2,rmse,mape,mae))
plt.grid()
plt.savefig("im2.png")
plt.show()
```

No handles with labels found to put in legend.

R²: -3.19 RMSE: 0.30 MAPE: 42.48 MAE: 0.19

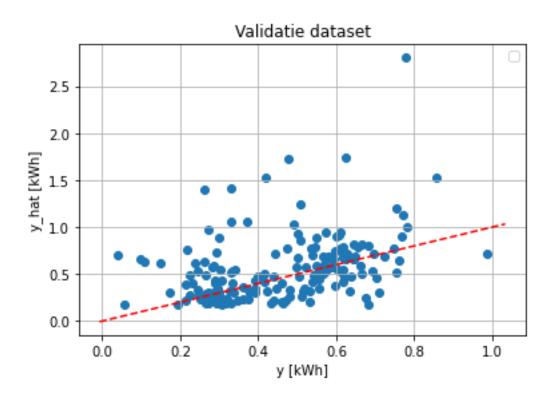


Validation evaluation

```
[30]: model.eval()
      y = scalery.inverse_transform(valid_y_t.cpu().detach().numpy())
      yhat = scalery.inverse transform(model(valid X t.float()).cpu().detach().
       →numpy())
      # R^2-score:
      r2 = r2_score(yhat, y)
      # RMSE:
      rmse = np.sqrt(mean_squared_error(yhat, y))
      # MAPE:
      actual, pred = np.array(y), np.array(yhat)
      mape = np.mean(np.abs((actual - pred) / actual)) * 100
      # MAE:
      mae = mean_absolute_error(yhat, y)
      #zet de voorspelde tegen de werkelijke waarde uit.
      %matplotlib inline
      plt.scatter(yhat,y)
      plt.xlabel("y [kWh]")
      plt.ylabel("y_hat [kWh]")
      plt.legend(loc="upper right")
      plt.title("Validatie dataset")
      plt.plot(plt.xlim(), plt.xlim(), ls="--", c='r', label="$y$=$\hat{y}$")
      # plt.xlim([0.1,1])
      # plt.ylim([0.1,1])
      print('R\u00b2: \t%.2f \nRMSE: \t%.2f \nMAPE: \t%.2f \nMAE: \t%.2f' %_
      \rightarrow (r2,rmse,mape,mae))
      plt.grid()
      plt.savefig("im3.png")
      plt.show()
```

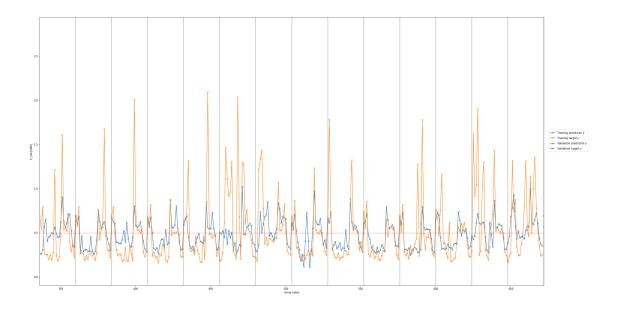
No handles with labels found to put in legend.

R²: -2.84 RMSE: 0.35 MAPE: 38.99 MAE: 0.22



```
[31]: A = np.sum(scalery.inverse_transform(model(train_X_t.float()).cpu().detach().
       →numpy()))
      B = np.sum(scalery.inverse_transform(train_y_t.cpu().detach().numpy()))
      C = (A-B)/B*100
      D = np.sum(scalery.inverse_transform(model(valid_X_t.float()).cpu().detach().
      →numpy()))
      E = np.sum(scalery.inverse_transform(valid_y_t.cpu().detach().numpy()))
      F = (D-E)/E*100
      print("Train:\t\t Model = %.2f kWh \t Actual = %.2f kWh \t percentage = %.2f" __
       \rightarrow% (A,B,C) +"%")
      print("Validation:\t Model = %.2f kWh \t Actual = %.2f kWh \t percentage = %.
       \rightarrow2f" % (D,E,F) +"%")
     Train:
                        Model = 321.65 \text{ kWh}
                                                  Actual = 333.30 \text{ kWh}
                                                                           percentage =
     -3.50%
     Validation:
                        Model = 77.69 \text{ kWh}
                                                 Actual = 95.46 \text{ kWh}
                                                                           percentage =
     -18.62%
[32]: kaas = []
      \#kaas.append(([i[0] for i in train_y_t.detach().cpu().numpy().tolist()])[-1:
       →][0])
```

```
[33]: # Training
     %matplotlib inline
     plt.subplots(figsize=(30,15))
     plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
      →inverse_transform(model(train_X_t.float()).detach().cpu().numpy()),
               "x-", label="Training prediction $\hat{y}$")
     plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
      →inverse_transform(train_y_t.detach().cpu().numpy()),
               "x-", label="Training target $y$")
     plt.grid()
     plt.ylabel("E_out [kWh]")
     plt.legend(loc=(1.01, 0.5))
     plt.plot((np.arange(len(train_y_t.detach().cpu().numpy()),len(train_y_t.
      →detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(),__
      →kaas,
               "x-", label="Validation prediction $\hat{y}$")
     plt.plot((np.arange(len(train_y_t.detach().cpu().numpy()),len(train_y_t.
      →detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(), ___
      →kaas1,
               "x-", label="Validation target $y$")
      [plt.axvline(i,color="black", alpha=0.4) for i in list(range(0,len(valid_y_t.
      →detach().cpu().numpy())+len(train_y_t.detach().cpu().numpy())+24,24))]
     plt.axhline(np.mean(scalery.inverse_transform(train_y_t.detach().cpu().numpy().
      # layout
      # plt.ylim([-1,1.5])
      #plt.xlim([672,840])
     plt.xlim([336,672])
     plt.xlabel("Array index")
     plt.ylabel("E out [kWh]")
     plt.legend(loc=(1.01, 0.5))
     plt.grid()
     plt.tight_layout()
     plt.savefig("im4.png")
     plt.show()
```

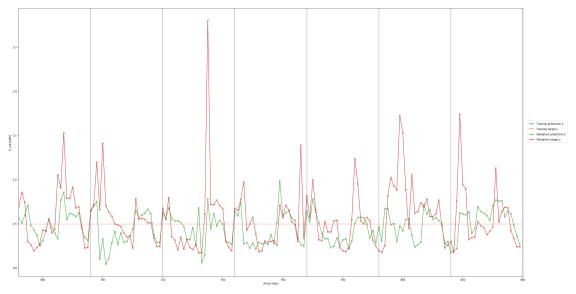


```
[34]: # Training
      %matplotlib inline
      plt.subplots(figsize=(30,15))
      plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
       →inverse_transform(model(train_X_t.float()).detach().cpu().numpy()),
               "x-", label="Training prediction $\hat{y}$")
      plt.plot(np.arange(0,len(train y t.detach().cpu().numpy())), scalery.
       →inverse_transform(train_y_t.detach().cpu().numpy()),
               "x-", label="Training target $y$")
      plt.grid()
      plt.ylabel("E_out [kWh]")
      plt.legend(loc=(1.01, 0.5))
      plt.plot((np.arange(len(train_y_t.detach().cpu().numpy()),len(train_y_t.
       →detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(), ___
       ⊸kaas,
               "x-", label="Validation prediction $\hat{y}$")
      plt.plot((np.arange(len(train y_t.detach().cpu().numpy()),len(train y_t.

→detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(),

       ⊸kaas1,
               "x-", label="Validation target $y$")
      [plt.axvline(i,color="black", alpha=0.4) for i in list(range(0,len(valid_y_t.
      \rightarrowdetach().cpu().numpy())+len(train_y_t.detach().cpu().numpy())+24,24))]
      plt.axhline(np.mean(scalery.inverse_transform(train_y_t.detach().cpu().numpy().
       →tolist())),color='red', alpha=0.4)
      # layout
      # plt.ylim([-1,1.5])
      plt.xlim([672,840])
      # plt.xlim([336,672])
```

```
plt.xlabel("Array index")
plt.ylabel("E_out [kWh]")
plt.legend(loc=(1.01, 0.5))
plt.grid()
plt.tight_layout()
plt.savefig("im5.png")
plt.show()
```



```
[75]: #Save the figs:
      #export to pdf:
      from fpdf import FPDF
      pdf = FPDF()
      pdf.add_page()
      #write the doc:
      info = ["Dataset 1 is hierbij gebruikt, met een batchsize van 64.",
              "8 Layers in network, tussen iedere layer zit een dropout van 0.16.",
              "Iedere Layersize is 256 nodes diep. Er is een LeakyReLU als_
      →activatiefunctie gebruikt."]
      pdf.set_font("Arial","B",size=15)
      pdf.cell(200, 10, txt = "Model 1: Simple Neural Network",
               ln = 1, align = 'L')
      pdf.set_font("Arial",size=9)
      [pdf.cell(200, 5, txt = i, ln = 2, align = 'L') for i in info]
      pdf.image("im"+str(1)+".png",x=125,y=15, w=700/5*0.5, h=450/5*0.5)
```

```
#Training plot:
pdf.set_font("Arial","B",size=15)
pdf.text(10, 70, txt = "Training:")
pdf.set_font("Arial",size=9)
pdf.image("im"+str(2)+".png",x=10,y=75, w=700/5*0.5, h=450/5*0.5)
pdf.image("im"+str(4)+".png",x=80,y=75, w=700/5*0.9, h=450/5*0.5)

#Validation plot:
pdf.set_font("Arial","B",size=15)
pdf.text(10, 145, txt = "Validation:")
pdf.set_font("Arial",size=9)
pdf.image("im"+str(3)+".png",x=10,y=150, w=700/5*0.5, h=450/5*0.5)
pdf.image("im"+str(5)+".png",x=80,y=150, w=700/5*0.9, h=450/5*0.5)

#output the file:
pdf.output("Evaluation.pdf")
```

[75]: ''

```
ValueError
                                           Traceback (most recent call last)
<ipython-input-22-c5af4bb3c82b> in <module>()
      5 df = px.data.iris()
      6 pltx = px.scatter(y)
---> 7 plotly.io.write_image(pltx,file='pltx.png',format='png',width=700,u
\rightarrowheight=450)
      8 pltx=(os.getcwd()+'/'+"pltx.png")
      9 ### define a method
/home/18005152/.local/lib/python3.7/site-packages/plotly/io/_kaleido.py in_u
→write_image(fig, file, format, scale, width, height, validate, engine)
    250
               height=height,
                validate=validate,
    251
--> 252
                engine=engine,
```

```
253
            )
    254
/home/18005152/.local/lib/python3.7/site-packages/plotly/io/_kaleido.py in_
 →to image(fig, format, width, height, scale, validate, engine)
    107
                    height=height,
    108
                    scale=scale,
--> 109
                    validate=validate,
                )
    110
    111
            elif engine != "kaleido":
/home/18005152/.local/lib/python3.7/site-packages/plotly/io/_orca.py in_
→to_image(fig, format, width, height, scale, validate)
   1533
            # Make sure orca sever is running
            # -----
   1534
-> 1535
            ensure_server()
   1536
           # Handle defaults
   1537
/home/18005152/.local/lib/python3.7/site-packages/plotly/io/ orca.py in ...
→ensure server()
                # Validate orca executable only if server url is not provided
   1388
                if status.state == "unvalidated":
   1389
-> 1390
                    validate_executable()
   1391
                # Acquire lock to make sure that we keep the properties of \Box
\hookrightarroworca_state
               # consistent across threads
   1392
/home/18005152/.local/lib/python3.7/site-packages/plotly/io/_orca.py inu
 →validate_executable()
   1085
                        executable=config.executable,
   1086
                        formatted_path=formatted_path,
-> 1087
                        instructions=install_location_instructions,
   1088
                    )
   1089
                )
ValueError:
The orca executable is required to export figures as static images,
but it could not be found on the system path.
Searched for executable 'orca' on the following path:
    /home/hub/bin
    /usr/local/cuda/bin
    /home/hub/bin
    /opt/jupyterhub/anaconda/bin
    /opt/jupyterhub/spark-2.2.0/bin
    /usr/local/sbin
    /usr/local/bin
```

```
/usr/sbin
    /usr/bin
    /sbin
    /bin
    /usr/games
    /usr/local/games
    /snap/bin
    /usr/lib/jvm/java-8-oracle/bin
    /usr/lib/jvm/java-8-oracle/db/bin
    /usr/lib/jvm/java-8-oracle/jre/bin
If you haven't installed orca yet, you can do so using conda as follows:
    $ conda install -c plotly plotly-orca
Alternatively, see other installation methods in the orca project README at
https://github.com/plotly/orca
After installation is complete, no further configuration should be needed.
If you have installed orca, then for some reason plotly.py was unable to
locate it. In this case, set the `plotly.io.orca.config.executable`
property to the full path of your orca executable. For example:
    >>> plotly.io.orca.config.executable = '/path/to/orca'
After updating this executable property, try the export operation again.
If it is successful then you may want to save this configuration so that it
will be applied automatically in future sessions. You can do this as follows:
    >>> plotly.io.orca.config.save()
If you're still having trouble, feel free to ask for help on the forums at
https://community.plot.ly/c/api/python
```

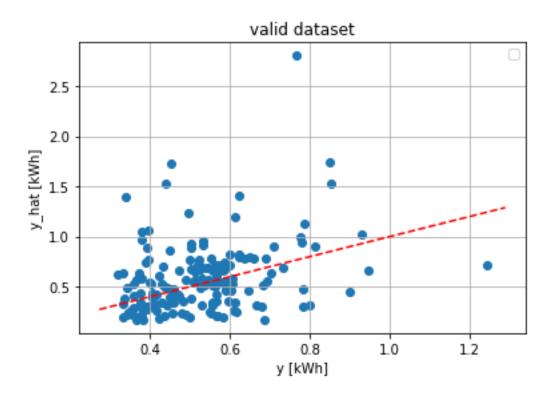
2 MVLR

```
[17]: from sklearn import linear_model
  regr = linear_model.LinearRegression()
  regr.fit(train_X,train_y)
  yhat = scalery.inverse_transform(regr.predict(valid_X))
  y = scalery.inverse_transform(valid_y)
```

```
[18]: r2 = r2\_score(yhat, y)
     rmse = np.sqrt(mean_squared_error(yhat, y))
     actual, pred = np.array(y), np.array(yhat)
     mape = np.mean(np.abs((actual - pred) / actual)) * 100
     mae = mean_absolute_error(yhat, y)
     #zet de voorspelde tegen de werkelijke waarde uit.
     %matplotlib inline
     plt.scatter(yhat,y)
     plt.xlabel("y [kWh]")
     plt.ylabel("y_hat [kWh]")
     plt.legend(loc="upper right")
     plt.title('valid dataset')
     plt.plot(plt.xlim(), plt.xlim(), ls="--", c='r', label="$y$=$\hat{y}$")
     # plt.xlim([0.1,1])
     # plt.ylim([0.1,1])
     print('R\u00b2: \t%.2f \nRMSE: \t%.2f \nMAPE: \t%.2f \nMAE: \t%.2f' %_
      plt.grid()
     plt.show()
```

No handles with labels found to put in legend.

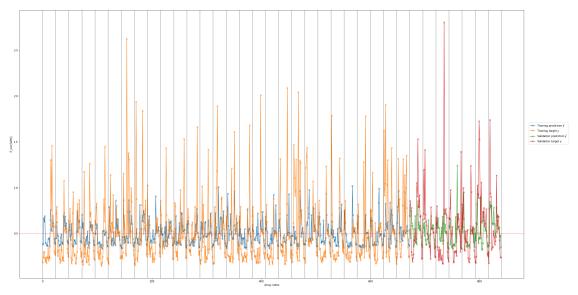
R²: -4.25 RMSE: 0.34 MAPE: 46.04 MAE: 0.22



```
[19]: # Training
      %matplotlib inline
      plt.subplots(figsize=(30,15))
      plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
      →inverse_transform(regr.predict(train_X)),
               "x-", label="Training prediction $\hat{y}$")
      plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
       →inverse_transform(train_y_t.detach().cpu().numpy()),
               "x-", label="Training target $y$")
      plt.grid()
      plt.ylabel("E_out [kWh]")
      plt.legend(loc=(1.01, 0.5))
      plt.plot((np.arange(len(train_y_t.detach().cpu().numpy()),len(train_y_t.
      →detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(),
      ⇔scalery.inverse_transform(regr.predict(valid_X)),
               "x-", label="Validation prediction $\hat{y}$")
      plt.plot((np.arange(len(train_y_t.detach().cpu().numpy()),len(train_y_t.

→detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(),

      →scalery.inverse_transform(valid_y),
               "x-", label="Validation target $y$")
      [plt.axvline(i,color="black", alpha=0.4) for i in list(range(0,len(valid_y_t.
       detach().cpu().numpy())+len(train y t.detach().cpu().numpy())+24,24))]
```



3 SVR.

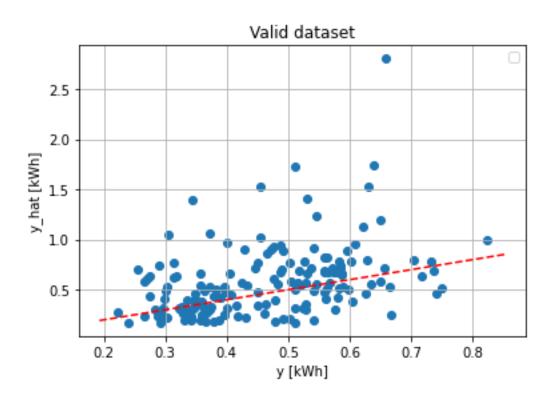
```
[20]: from sklearn.svm import SVR
regr = SVR()
regr.fit(train_X,train_y)
yhat = scalery.inverse_transform(regr.predict(valid_X))
y = scalery.inverse_transform(valid_y)
```

/opt/jupyterhub/anaconda/lib/python3.7/sitepackages/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().
return f(**kwargs)

```
[21]: r2 = r2\_score(yhat, y)
     rmse = np.sqrt(mean_squared_error(yhat, y))
     actual, pred = np.array(y), np.array(yhat)
     mape = np.mean(np.abs((actual - pred) / actual)) * 100
     mae = mean_absolute_error(yhat, y)
     #zet de voorspelde tegen de werkelijke waarde uit.
     %matplotlib inline
     plt.scatter(yhat,y)
     plt.xlabel("y [kWh]")
     plt.ylabel("y_hat [kWh]")
     plt.legend(loc="upper right")
     plt.title('Valid dataset')
     plt.plot(plt.xlim(), plt.xlim(), ls="--", c='r', label="$y$=$\hat{y}$")
     # plt.xlim([0.1,1])
     # plt.ylim([0.1,1])
     print('R\u00b2: \t%.2f \nRMSE: \t%.2f \nMAPE: \t%.2f \nMAE: \t%.2f' %_
      plt.grid()
     plt.show()
```

No handles with labels found to put in legend.

R²: -6.30 RMSE: 0.34 MAPE: 52.64 MAE: 0.22

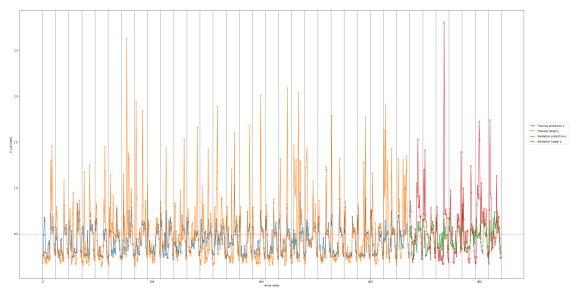


```
C = (A-B)/B*100
      D = np.sum(scalery.inverse_transform(regr.predict(valid_X)))
      E = np.sum(scalery.inverse_transform(valid_y))
      F = (D-E)/E*100
      print("Train:\t\t Model = %.2f kWh \t Actual = %.2f kWh \t percentage = %.2f" __
       \rightarrow% (A,B,C) +"%")
      print("Validation:\t Model = %.2f kWh \t Actual = %.2f kWh \t percentage = %.
       \rightarrow2f" % (D,E,F) +"%")
     Train:
                        Model = 284.43 \text{ kWh}
                                                  Actual = 333.30 \text{ kWh}
                                                                             percentage =
     -14.66%
     Validation:
                        Model = 77.38 \text{ kWh}
                                                  Actual = 95.46 \text{ kWh}
                                                                             percentage =
     -18.94%
[23]: # Training
      %matplotlib inline
      plt.subplots(figsize=(30,15))
      plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
       →inverse_transform(regr.predict(train_X)),
```

[22]: A = np.sum(scalery.inverse_transform(regr.predict(train_X)))

B = np.sum(scalery.inverse_transform(train_y))

```
"x-", label="Training prediction $\hat{y}$")
plt.plot(np.arange(0,len(train_y_t.detach().cpu().numpy())), scalery.
 →inverse_transform(train_y_t.detach().cpu().numpy()),
         "x-", label="Training target $y$")
plt.grid()
plt.ylabel("E out [kWh]")
plt.legend(loc=(1.01, 0.5))
plt.plot((np.arange(len(train_y_t.detach().cpu().numpy()),len(train_y_t.
→detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(),
⇒scalery.inverse_transform(regr.predict(valid_X)),
         "x-", label="Validation prediction $\hat{y}$")
plt.plot((np.arange(len(train y t.detach().cpu().numpy()),len(train y t.
→detach().cpu().numpy())+len(valid_y_t.detach().cpu().numpy()))).tolist(), ___
⇒scalery.inverse_transform(valid_y),
         "x-", label="Validation target $y$")
[plt.axvline(i,color="black", alpha=0.4) for i in list(range(0,len(valid_y_t.
→detach().cpu().numpy())+len(train_y_t.detach().cpu().numpy())+24,24))]
plt.axhline(np.mean(scalery.inverse_transform(train_y_t.detach().cpu().numpy().
→tolist())),color='red', alpha=0.4)
# lavout
#plt.ylim([-1,1.5])
#plt.xlim([672,840])
#plt.xlim([336,672])
plt.xlabel("Array index")
plt.ylabel("E_out [kWh]")
plt.legend(loc=(1.01, 0.5))
plt.grid()
plt.tight_layout()
plt.show()
```



3.1 Stop de notebook:

 $//\% javascript \ // Jupyter.notebook.session.delete()$