

FYS375 Modelling av solfanger

Gruppe 3 - Høst 2021

Modellering av solfanger. Her lages en model basert på data fra mars til juli 2020. Med gjeldene krav fra ASHRAE-93. De predikerte verdiene er plottet mot målte verdier, og samlignet med en benchmarkmodell med 50% virkningsgrad.

Nøkkeltall:

Datapunkter i 2020: 586

$RMSE_{model} = 16.206$

$RMSE_{benchmark} = 28.095$

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import math

from sklearn.metrics import mean_squared_error
from sklearn.linear_model import LinearRegression

In [2]: data = pd.read_csv("Måledata mars_juli_2020.csv")
data.Time = pd.to_datetime(data.Time)
data

Out[2]:
```

| | Time | Irradiance (W/m2) | Ambient Temperature (C) | Inlet Temperature (C) | Mass flow (g/s) | Outlet Temperature (C) |
|--|--------|---------------------|-------------------------|-----------------------|-----------------|------------------------|
| | 0 | 2020-03-01 00:01:00 | 0.0 | 0.4 | 15.6 | 107.0 |
| | 1 | 2020-03-01 00:02:00 | 0.0 | 0.4 | 15.6 | 107.0 |
| | 2 | 2020-03-01 00:03:00 | 0.0 | 0.4 | 15.6 | 107.0 |
| | 3 | 2020-03-01 00:04:00 | 0.0 | 0.4 | 15.6 | 107.0 |
| | 4 | 2020-03-01 00:05:00 | 0.0 | 0.4 | 15.6 | 107.0 |
| | ... | ... | ... | ... | ... | ... |
| | 220314 | 2020-07-31 23:55:00 | -1.0 | 17.7 | 25.5 | 0.0 |
| | 220315 | 2020-07-31 23:56:00 | -1.0 | 17.8 | 25.5 | 0.0 |
| | 220316 | 2020-07-31 23:57:00 | -1.0 | 17.8 | 25.5 | 0.0 |
| | 220317 | 2020-07-31 23:58:00 | -1.0 | 17.8 | 25.5 | 0.0 |
| | 220318 | 2020-07-31 23:59:00 | -1.0 | 17.8 | 25.5 | 0.0 |

220319 rows x 6 columns

```
In [3]: def find_irradiance(data):
    try:
        data_irra = data[data["Irradiance (W/m2)"] > 790]
    except:
        data_irra = data[data["Irradiance (W/m2)"] > 790]
    return data_irra.to_numpy()

In [4]: def find_15_min_intervall(np_data):
    info = []
    for i, (tid, irradiance, Ta, Ti, ms, To) in enumerate(np_data[15:]):
        if tid - np_data[i][0] == pd.Timedelta("15m"):
            # save all 15 minutes innside infomration
            info.append(np_data[i:i+15])
    info = np.array(info)
    return info
```

Sette opp kravene våres til steady-state

$Irradiance \pm 32$

$T_a \pm 1.5$

$T_i \pm 1$

```
In [5]: def steady_state_krav(info):
    innenfor_krav = []

    for intervall15min in info:
        ir_mean, Ta_mean, Ti_mean, _, To_mean = (np.mean(intervall15min[:,1:], axis=0))
        if (np.max(abs(ir_mean - intervall15min[:,1])) <= 32 and
            np.max(abs(Ta_mean - intervall15min[:,2])) <= 1.5 and
            np.max(abs(To_mean - intervall15min[:,5])) <= 1 and
            np.min(intervall15min[:,4]) >= 50):
            innenfor_krav.append(intervall15min)
    innenfor_krav = np.array(innenfor_krav)
    return innenfor_krav

In [6]: def delete_duplicates(innenfor_krav):
    femtenmin_split = np.array(innenfor_krav[0:2])

    for i,v in enumerate(innenfor_krav[2:]):
        print(f"{i} av {len(innenfor_krav[2:])}",end="\r")
        if not np.isin(v[:,0], femtenmin_split[:, :, 0]).any():
            #print("legger til")
            femtenmin_split = np.append(femtenmin_split, [v], axis=0)

    femtenmin_split = femtenmin_split[1:]
    return femtenmin_split
```

Filtere massestrømmer: mean \pm 1std

Beholder massestrømmen som er mest like ved å finne snittet og standardavviket.

```
In [7]: def remove_ms_std(femtenmin_split):
    snitt = np.mean(femtenmin_split[:, :, 4].flatten())
    std = np.std(femtenmin_split[:, :, 4].flatten())

    femtenmin_split = femtenmin_split[(abs(femtenmin_split[:, :, 4] - snitt) / std <= 1).any(axis=1)]
    print(f"snitt: {snitt:.2f}, std:{std:.2f}")
    return femtenmin_split

In [8]: data_irra = find_irradiance(data)
info = find_15_min_intervall(data_irra)
innenfor_krav = steady_state_krav(info)
femtenmin_split = delete_duplicates(innenfor_krav)
ready_15_min_intervall = remove_ms_std(femtenmin_split)

snitt: 374.29, std:42.89
```

Modellering

Finne variabel (X) veridene for hvert momentanpunkt:

(1) $\frac{T_i - T_a}{G_T} [Tm2/W]$

Finne virkningsgrad ved å se max effekt og Q

(2) $\eta = \frac{Q}{GA}$

Effekt i vannet

$c = 4.183 \text{ j/g K}$

(3) $Q = \dot{m}c\Delta T [W]$

Max effekt

Solfanger er $A=67.2 \text{ m}^2$

(4) $G_A = G * A [W]$

```
In [10]: def formulas(femtenmin_split):
    X = []
    Q = []
    GA = []
    for i in femtenmin_split:
        ir_mean, Ta_mean, Ti_mean, m_s_mean, To_mean = np.mean(i[:, 1:], axis=0)

        # X formelen (1)
        x = (Ti_mean - Ta_mean)/ir_mean
        X.append(x)

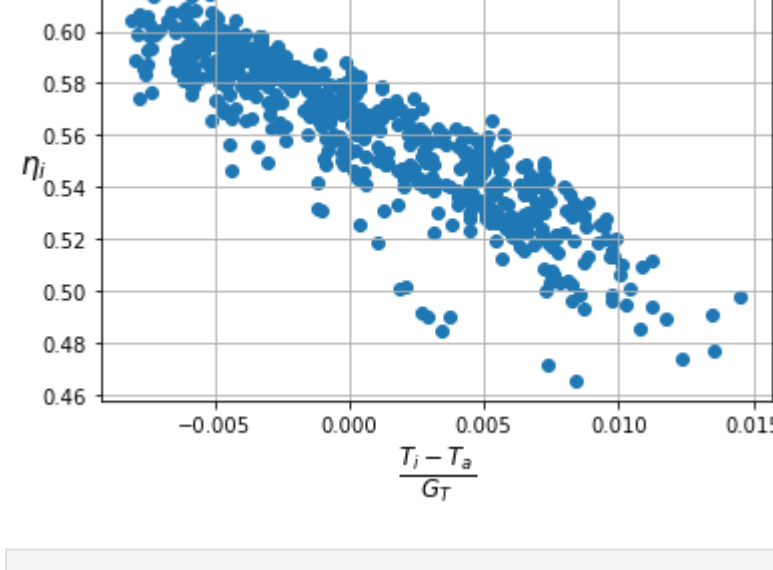
        # Q formelen (3)
        q = m_s_mean * 4.183 * (To_mean - Ti_mean)
        Q.append(q)

        # GA foremlen (4)
        ga = ir_mean * 67.2
        GA.append(ga)

    X = np.array(X)
    Q = np.array(Q)
    GA = np.array(GA)
    n = Q/GA # (2)
    return X, Q, GA, n

In [11]: X, Q, GA, n = formulas(ready_15_min_intervall)
```

```
In [12]: plt.scatter(X[n>0.2],n[n>0.2])
plt.grid()
plt.xlabel("$\\frac{T_i - T_a}{G_T}$", fontsize=17)
plt.ylabel("$\\eta$", fontsize=15).set_rotation(0)
plt.show()
```

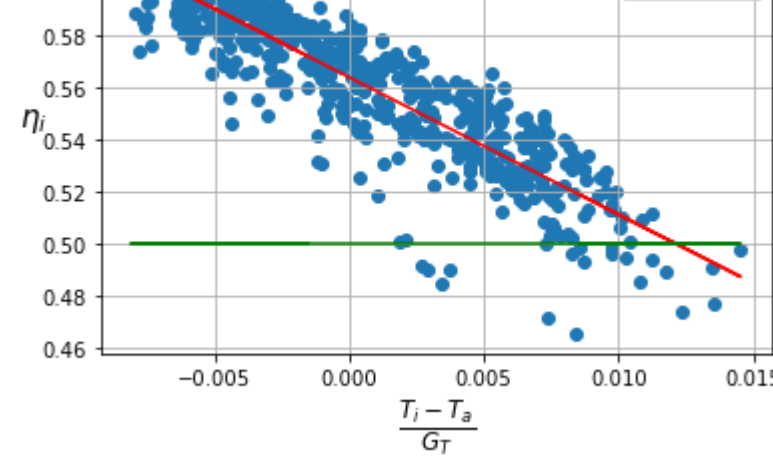


```
In [13]: reg = LinearRegression().fit(X[n>0.2].reshape(-1,1),n[n>0.2].reshape(-1,1))
print(f"r2score: {reg.score(X[n>0.2].reshape(-1,1),n[n>0.2].reshape(-1,1))}")

r2score: 0.7941282898548367
```

```
In [14]: pred = reg.predict(X[n>0.2].reshape(-1,1))
```

```
In [15]: plt.scatter(X[n>0.2],n[n>0.2])
plt.plot(X[n>0.2],pred, c="red", label="modell")
plt.plot(X[n>0.2],[0.5 for i in range(len(pred))], c="green", label="$\\eta_{50\\%}$")
plt.grid()
plt.xlabel("$\\frac{T_i - T_a}{G_T}$", fontsize=17)
plt.ylabel("$\\eta$", fontsize=15).set_rotation(0)
plt.legend()
plt.show()
```



```
In [16]: print(f"b_1: {reg.coef_[:,0]} \t = Fr*U1")
print(f"b0: {reg.intercept_} \t = Fr(tau)")

Fr = reg.intercept_ / (0.85 * 0.95)
print(f"Fr er {Fr}")

U1 = -reg.coef_[0][0] / Fr
print(f"U1 er {U1}")

b_1: [[-5.27614314]]... = Fr*U1
b0: [0.5642758]... = Fr(tau)
Fr er [0.69879356]
U1 er [7.55036028]
```

Estimere energi for 2021

```
In [17]: data = pd.read_csv("Måledata mars_juli_2021.csv")
try:
    data = data.rename(columns={"Irradiance (W/m2)": "Irradiance (W/m2)"})
except:
    print("Heter Irradiance (W/m2) allerede")
```

```
In [18]: data.Time = pd.to_datetime(data.Time)
data.index = data.Time
data = data.drop(columns=["Time"])
```

Energi målt

$Q = \dot{m}c\Delta T$

```
In [19]: data["energi [kWh]"] = data["Mass flow (g/s)"] * 4.183 * (data["Outlet Temperature (C)"] - data["Inlet Tempe
```

Energi predikert

$\eta_{pred} = \frac{Q}{GA} = Fr * (\tau\alpha) - Fr * U_l \frac{T_i - T_a}{G}$

$Q_{pred} = Fr * (\tau\alpha) * GA - Fr * U_l (T_i - T_a) * A$

$\eta = \frac{Q}{GA} = \frac{\dot{m}c\Delta T}{GA}$

```
In [20]: Fr = 0.69718706
U1 = 7.55036028
tau = 0.85
alpha = 0.95
areal = 67.2

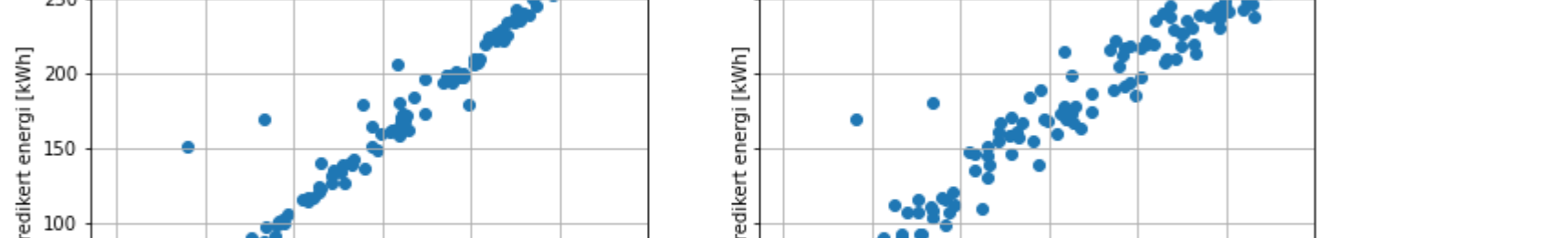
data["pred Q [W]"] = Fr * (tau * alpha) * data["Irradiance (W/m2)"] * areal - Fr * U1 * (data["Inlet Tempera
data["pred energi [kWh]"] = data["pred Q [W]"] * (1/60) * 1e-3
data["dårligmodell [kWh]"] = data["Irradiance (W/m2)"] * areal * 0.5 * (1/60) * 1e-3
```

Antar at alle predikerte verdier der Q blir mindre enn 0 ikke skjer i virkeligheten. Setter derfor alle negative verdier til 0

```
In [21]: t = data["pred energi [kWh]"].to_numpy()
t[t < 0] = 0
data["pred energi [kWh]"] = t
```

```
In [22]: # Samler det i dagsintervall og summer opp energien
data_per_dag = data.resample("1D").sum()
```

```
In [23]: fig, (ax1, ax2) = plt.subplots(nrows=1, ncols=2, sharey = True, figsize=(12,5))
ax1.scatter(data_per_dag["energi [kWh]"], data_per_dag["pred energi [kWh]"])
ax1.set_title("Modell", fontname="Times New Roman")
ax2.scatter(data_per_dag["energi [kWh]"], data_per_dag["dårligmodell [kWh]"])
ax2.set_title("Benchmark modell $\\eta_{50\\%}$", fontname="Times New Roman")
ax1.grid(); ax2.grid()
```



```
In [24]: print(f"RMSE for egen modell: {math.sqrt(mean_squared_error(data_per_dag['energi [kWh]'], data_per_dag['pred
print(f"RMSE for benchmark modell: {math.sqrt(mean_squared_error(data_per_dag['energi [kWh]'], data_per_dag[
```

RMSE for egen modell: 16.206
RMSE for benchmark modell: 28.095