

Controlling wind-solar power in Germany

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- ▶ electric power consumption in Germany $\approx 60 \text{ GW}_{el}$
- ▶ but 80% power consumption are nonelectric, (energy for transportation, warm water, space heating and in part for process heating)
- ▶ therefore: 'Energiewende': $60 \text{ GW}_{el} \rightarrow 300 \text{ GW}_{el}$

Problem, needed: controlled wind-solar power generation of 60 GW_{el} { 300 GW_{el} }.

database: power data over 7 years (2015 - 2021) from ENTSO-E

01.01.2021 00:00 - 01.01.2022 00:00 - CET		
Time	Germany (DE)	
	Day-ahead Total Load Forecast	Actual Total Load
	[MW]	[MW]
00:00 - 00:15	43567	45170
00:15 - 00:30	43371	44835
00:30 - 00:45	42884	44519
00:45 - 01:00	42803	44265
01:00 - 01:15	42096	43713
01:15 - 01:30	41824	43157
01:30 - 01:45	41426	42556
01:45 - 02:00	41051	42326
02:00 - 02:15	40781	41798

Germany (DE) / 01.01.2021 00:00 - 01.01.2022 00:00 - CET			
MTU	Solar	Wind Offshore	Wind Onshore
	Actual Aggregated	Actual Aggregated	Actual Aggregated
	[MW]	[MW]	[MW]
	0	0	0
00:00 - 00:15	0	337	4232
00:15 - 00:30	0	352	4098
00:30 - 00:45	0	404	3819
00:45 - 01:00	0	433	3723
01:00 - 01:15	0	419	3771
01:15 - 01:30	0	410	3689
01:30 - 01:45	0	379	3559
01:45 - 02:00	0	361	3439
02:00 - 02:15	0	318	3372

averaged power (GW) and scaling

year	2015	2016	2017	2018	2019	2020	2021
el. power demand \bar{P}_d	57.1	57.2	57.7	57.9	56.4	55.1	57.7
w-s volatile power \hat{P}_v	12.8	12.7	15.8	17.1	18.9	20.0	18.3

Scaling factor s_f required!

Scaling: s_f selected such that $\bar{\tilde{P}}_v \cdot s_f = \bar{P}_v = \bar{P}_d$

- ▶ energy demand: $E_d(t) = \int_0^t P_d(t') dt'$
- ▶ volatile energy generation: $E_v(t) = \int_0^t P_v(t') dt'$
- ▶ stored energy: $E_s(t) = \int_0^t P_s(t') dt'_s$

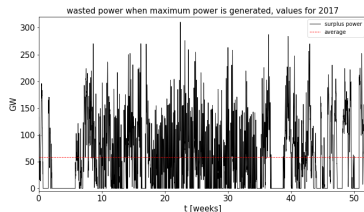
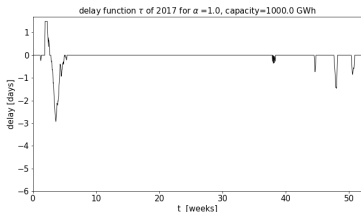
Together, Ansatz:

$$E_d(t + \tau) + E_s(t) = (1 + \alpha) \cdot E_v(t) - E_{disc}(t)$$

with

- ▶ τ = delay function, simulates smart meters
- ▶ $\alpha \cdot E_v(t)$ = surplus energy
- ▶ E_{disc} = 'discarded' energy

$$E_d(t + \tau) + E_s(t) = (1 + \alpha) \cdot E_v(t) - E_{disc}(t)$$



Three characteristic numbers

- ▶ n_λ =range of smart meters: $n_\lambda = 4.4$ [days]
- ▶ n_δ =days out of prescribed range: $n_\delta = 6$ [days]
- ▶ n_σ =days for which smart meters are active: $n_\sigma = 36$ [days]

Conclusion

based on i) precise (partly scaled) data over 7 years (2015-2021), ii) surplus of power, iii) smart meter and iv) characteristic numbers $(n_\lambda, n_\delta, n_\sigma)$ our approach predicts:

1. Germany's present electric power demand of $\approx 60\text{GW}$ can be supplied by wind-solar power alone.
2. Formidable volatility suppressed by storage capacity of $0.3\text{TWh} - 1\text{TWh}$.
3. $\approx 100\%$ surplus of wind-solar power devices required.

extrapolation to 300 GW case:

- I formidable volatility suppressed by storage capacity of $1.5\text{TWh} - 5\text{TWh}$.
- II almost every roof and possibly a significant part of facades have to be covered with solar cells.
- III 270 000 windturbines of type (6MW, 200m height) are required (area of Germany $360\ 000\ \text{km}^2$)

