

Anpfiff zur zweiten Hälfte der Energiewende:

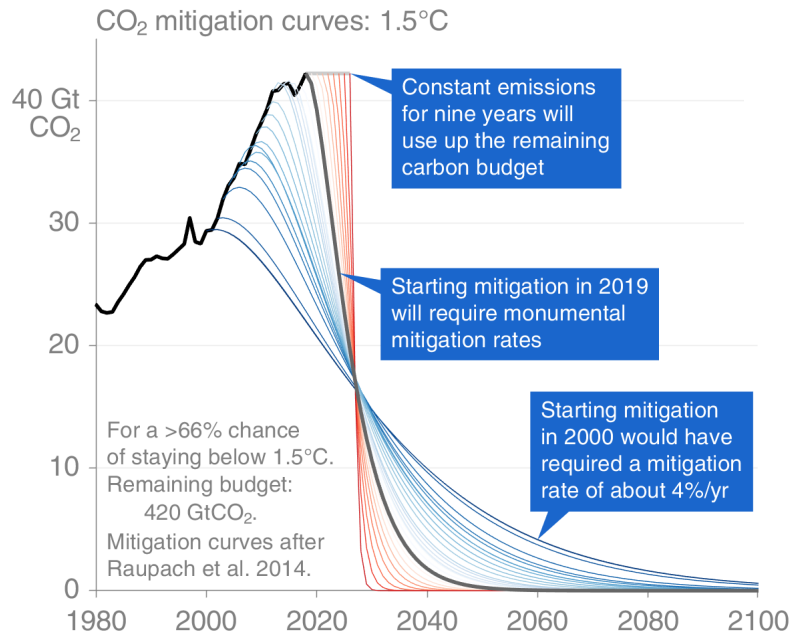
***Grundlegende Gedanken aus den
Natur- und Technikwissenschaften***

***“Mehr als die Vergangenheit und die Gegenwart
interessiert mich die Zukunft,
denn in ihr gedenke ich zu leben.”***

(Albert Einstein)



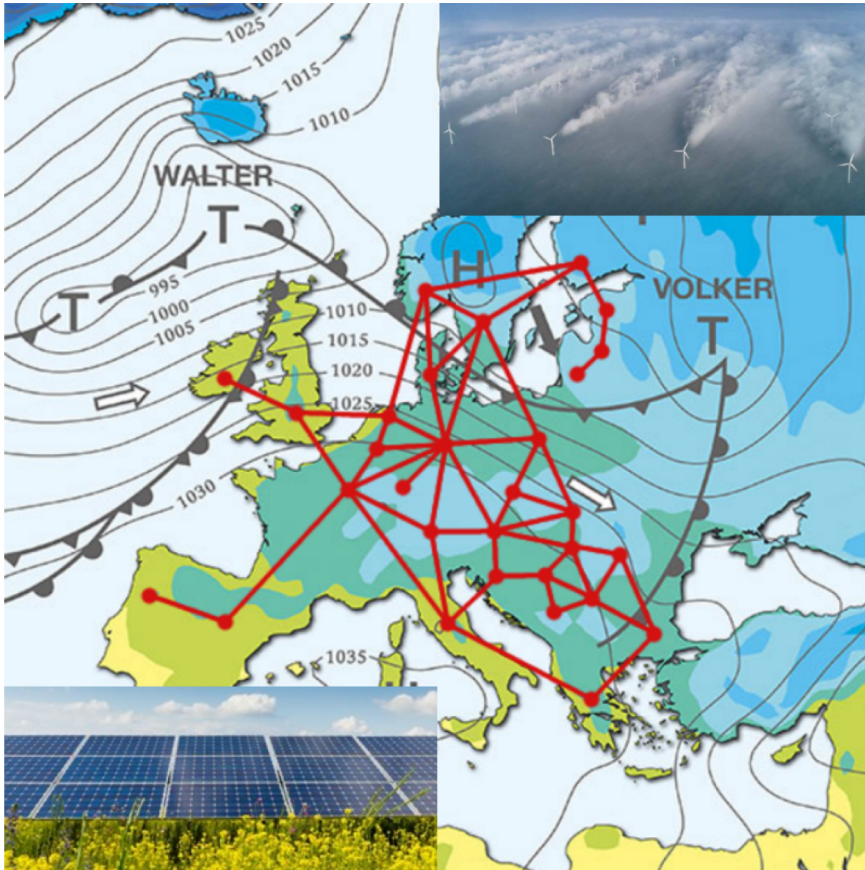
„A goal is a dream with a deadline“



© 2019 @robbie_andrew • Data: GCP • Emissions budget from IPCC SR1.5



Decarbonisation of the networked European Energy System

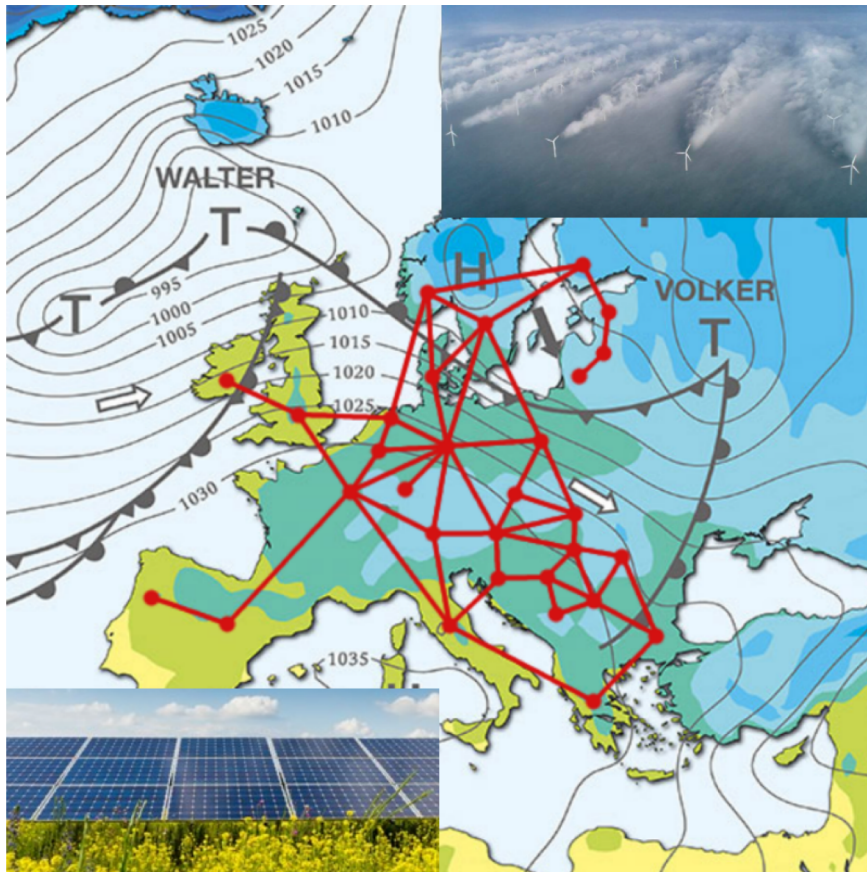


- I. simple modeling: **Renewable European electricity network**
 - weather-driven optimal design
 - value of cooperation

- II. advanced modeling: **electricity → energy network**
 - constraints
 - + (unavoidable) consequences

**capture / extract
general system dynamics
+ meaningful insights
+ inspirational results**

I. Simple Modeling: Renewable European electricity network



**Design of energy systems with
a high share of renewables:**

Let the weather decide!

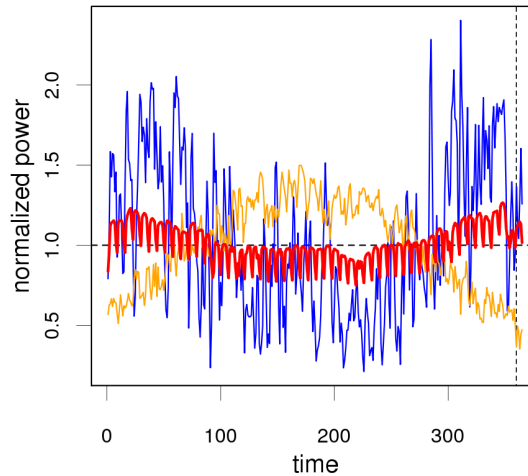
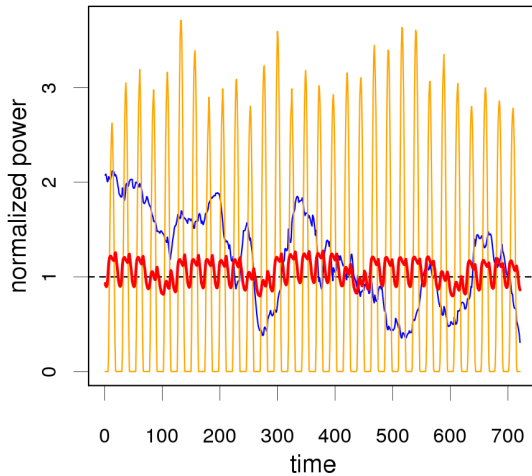
Renewable European electricity network + fluctuating „weather forces“

$$G_n^R(t) = G_n^W(t) + G_n^S(t)$$

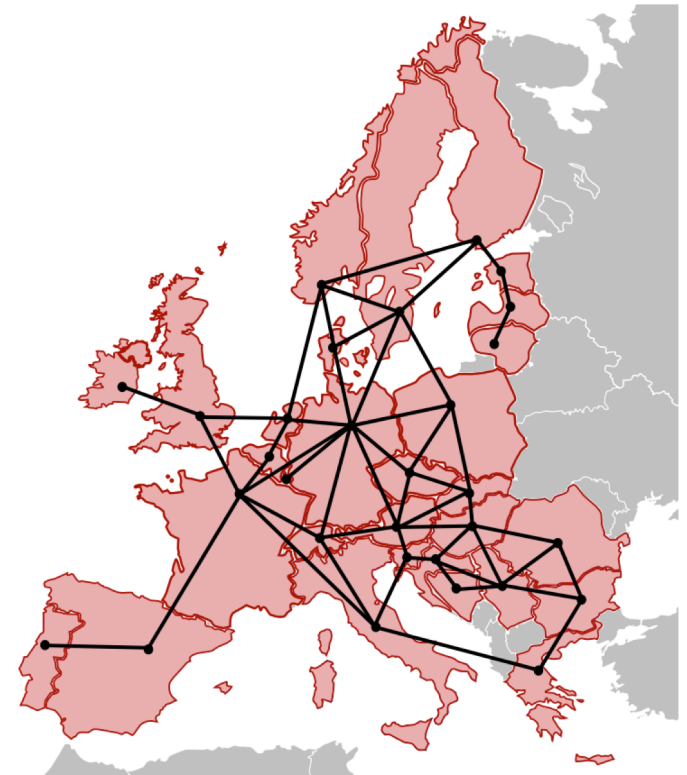
$$\langle G_n^R \rangle = \gamma_n \langle L_n \rangle$$

$$\langle G_n^W \rangle = \alpha_n \langle G_n^R \rangle$$

1980 – 2018: 1h, 30x30km²
Renewable Energy Atlas



3 TIME SCALES:
diurnal (1h-1d)
synoptic (2-10d)
seasonal (1y)



Renewable European electricity network + fluctuating „weather forces“

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$$\langle G_n^R \rangle = \gamma_n \langle L_n \rangle$$

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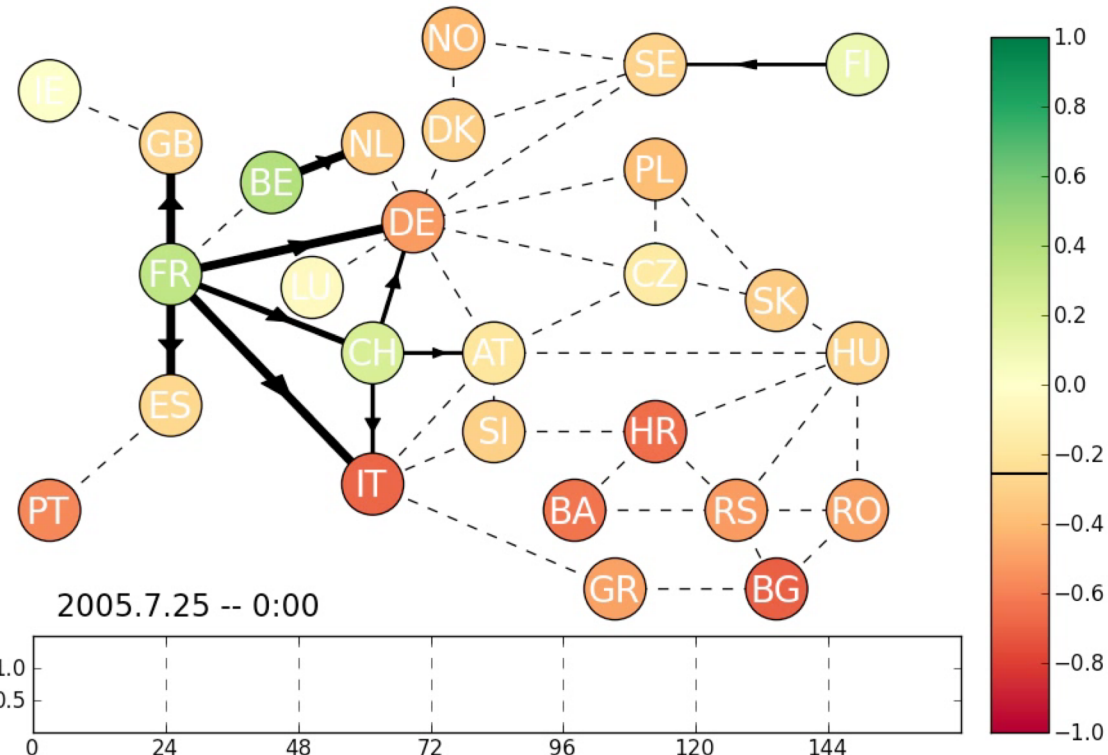
$$G_n^R(t) - L_n(t) = B_n(t) + P_n(t) + \dots$$

$$G_n^B(t) = (B_n(t))_-$$

$$C_n(t) = (B_n(t))_+$$

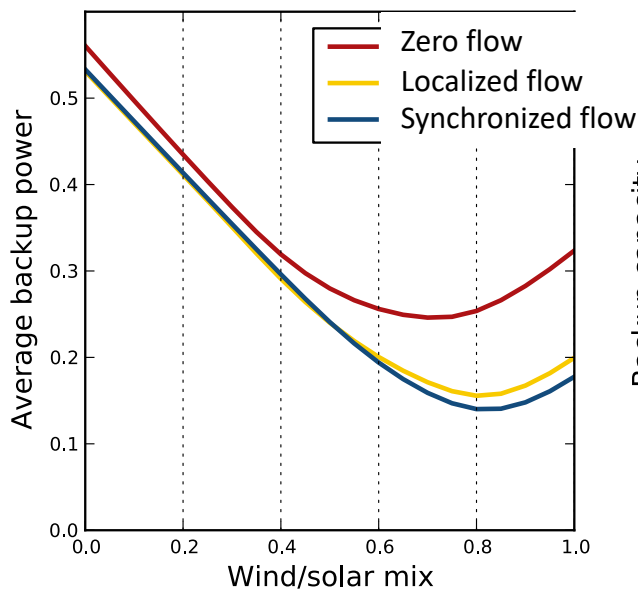
$$\sum_n P_n(t) = 0$$

$$F_l(t) = \sum_n H_{ln} P_n(t)$$



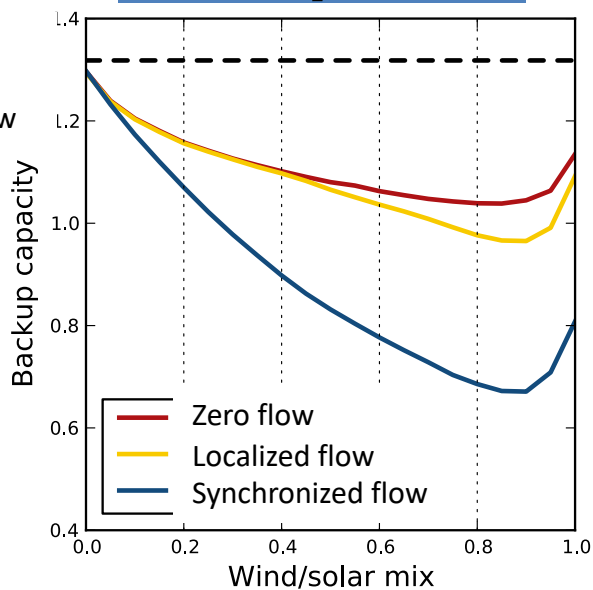
backup energy

$$\langle G_n^B \rangle$$



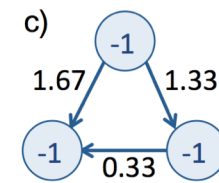
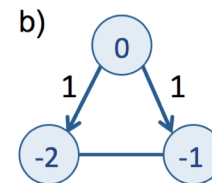
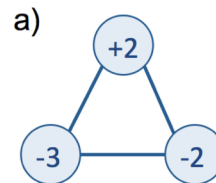
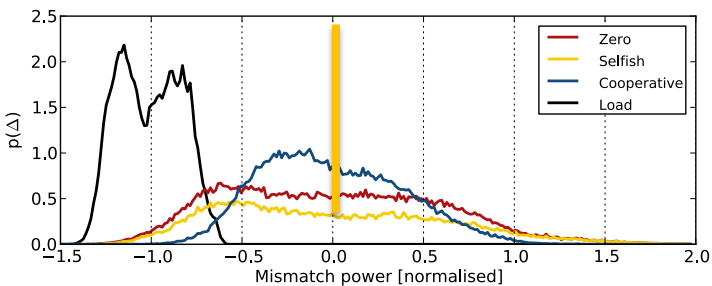
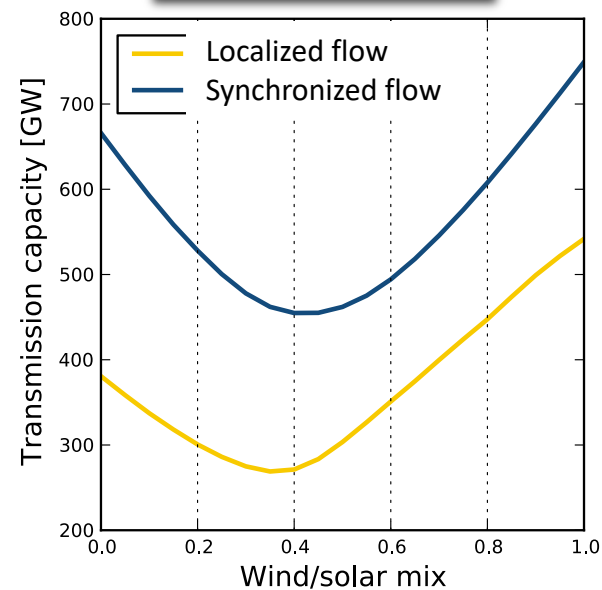
backup capacity

$$\max_q (G_n^B)$$



transmission capacity

$$\sum_l \max_q |F_l|$$



Infrastructure measures

backup energy

$$E_n^B = \langle G_n^B \rangle$$

backup capacity

$$K_n^B = \max_q (G_n^B)$$

transmission capacity

$$K_l^T = \max_q |F_l| \cdot d_l$$

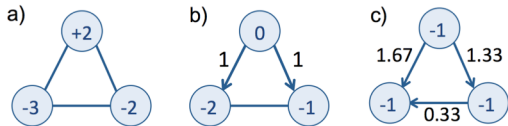
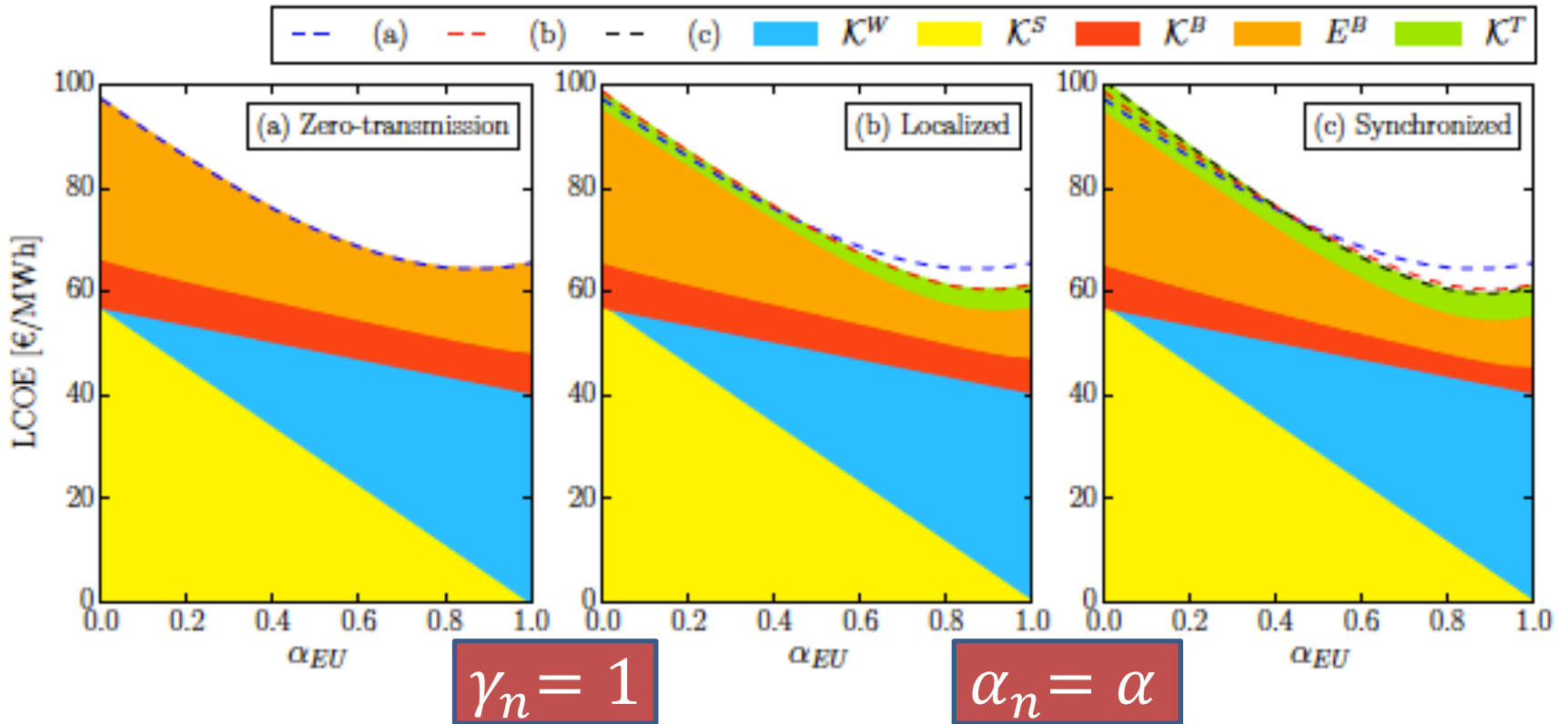
wind capacity

$$K_n^W = \frac{\alpha_n \gamma_n \langle L_n \rangle}{CF_n^W}$$

solar capacity

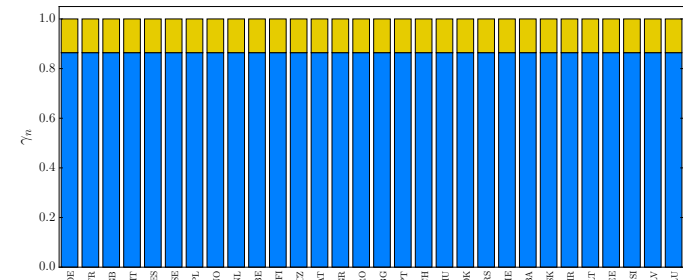
$$K_n^S = \frac{(1 - \alpha_n) \gamma_n \langle L_n \rangle}{CF_n^S}$$

Levelized Cost of SYSTEM Energy

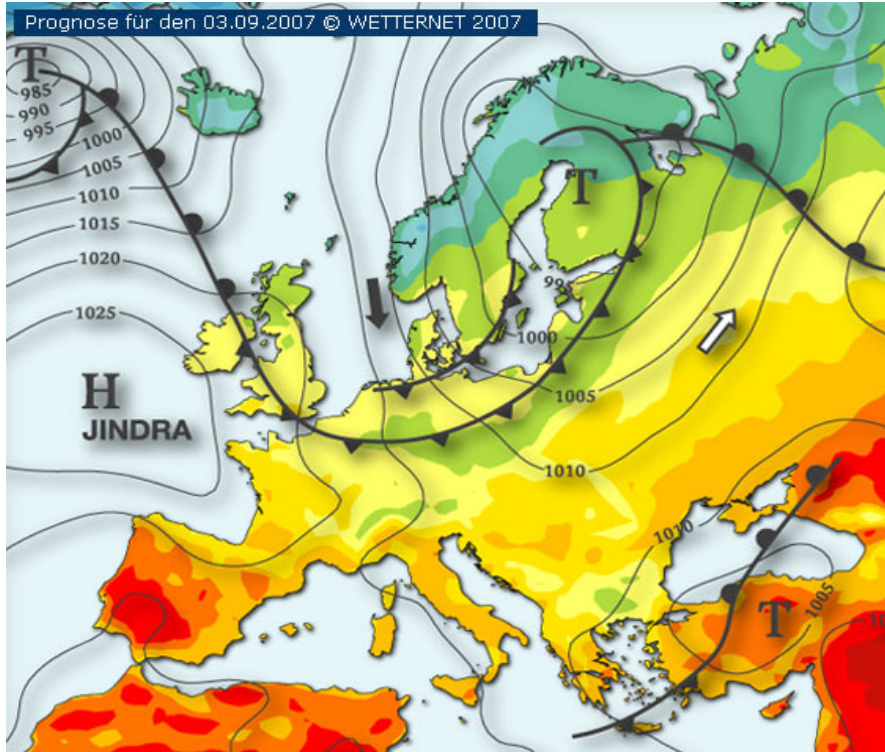


Coupling objectives: **B** ↔ **P**

$$G_n^R(t) - L_n(t) = B_n(t) + P_n(t)$$



wind and solar power capacities



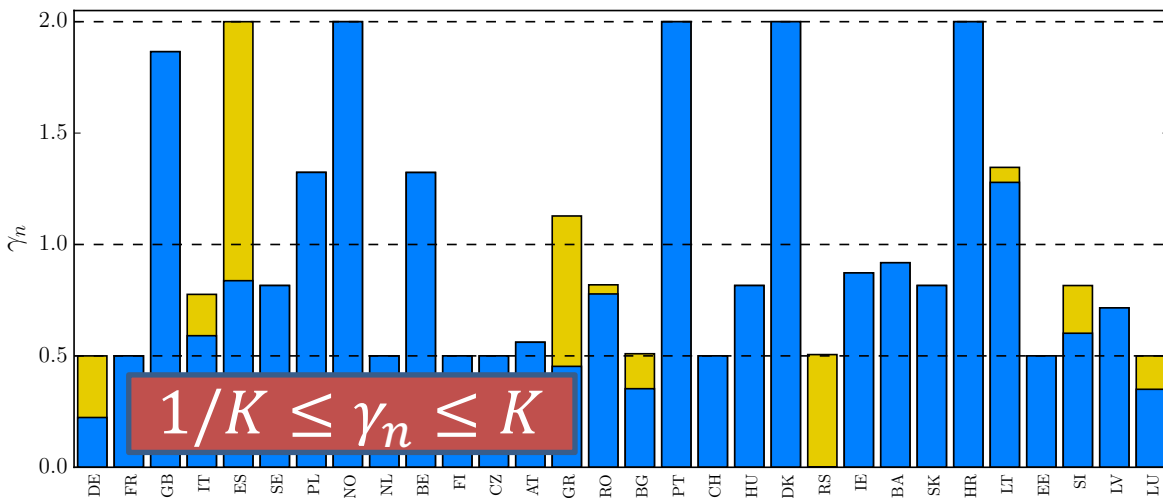
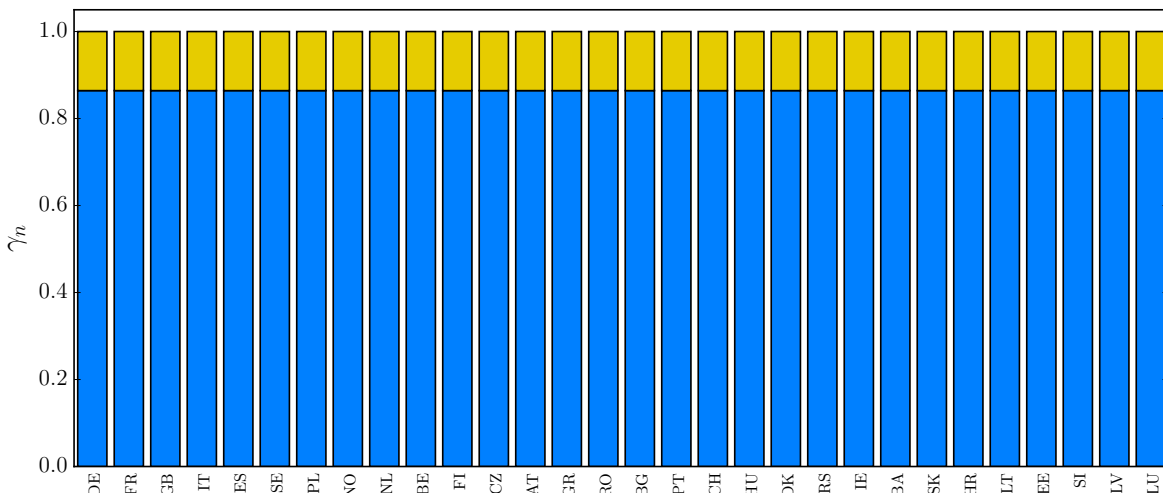
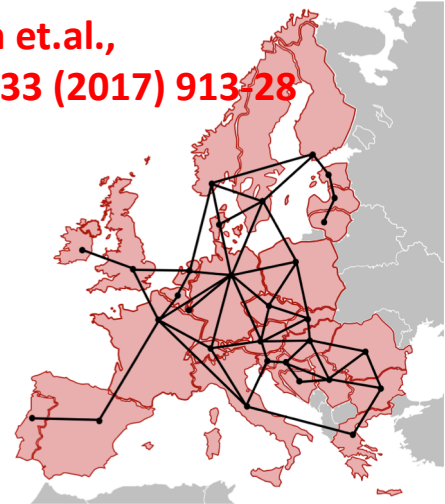
**annual consumption (2009)
= 3400 TWh**

**80% wind power generation
= 1000 GW installed capacity
= 200.000 x 5 MW turbines
= 5000 x 200 MW wind farms
≈ 130000 km²**

**20% solar PV power generation
= 370 GW installed capacity
≈ 2500 - 5000 km²**

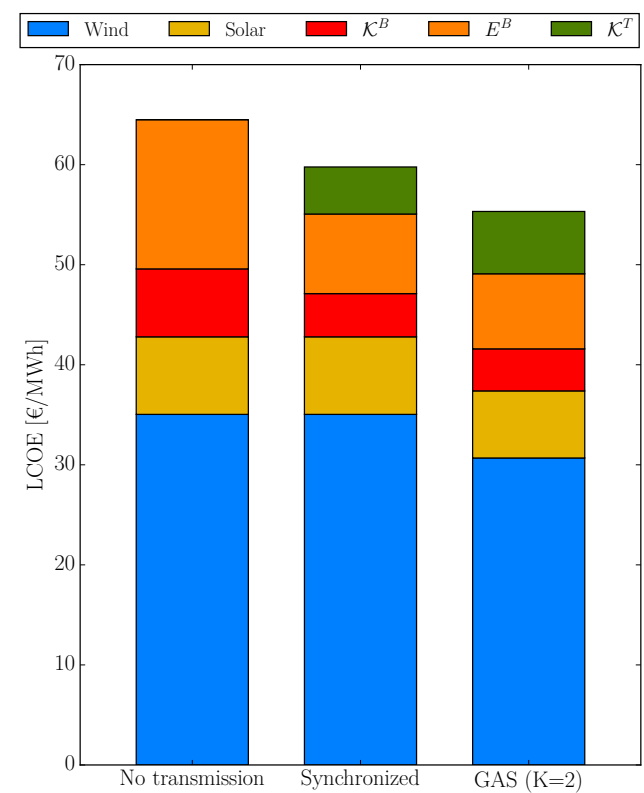
Breaking homogeneity: cost-optimal heterogeneity

E Eriksen et al.,
Energy 133 (2017) 913–28



$$\langle G_n^R \rangle = \gamma_n \langle L_n \rangle$$

$$\langle G_n^W \rangle = \alpha_n \langle G_n^R \rangle$$



Back-on-the-envelop estimate

OPT-HOM-noT(K=1): **64.5** €/MWh

OPT-HOM(K=1): **56.6** €/MWh

OPT-HET(K=2): **53.8** €/MWh

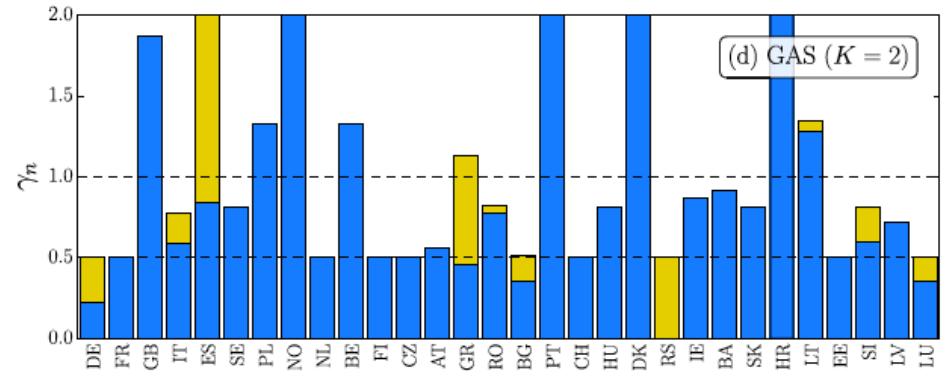
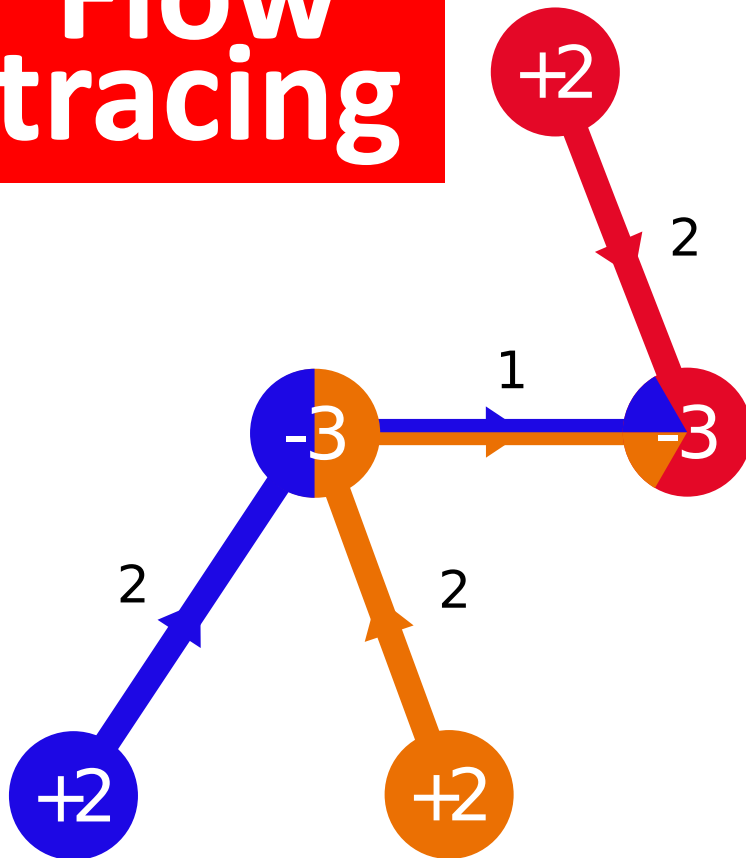
EU cost reduction / y

= 3500 TWh/y x **10** €/MWh

= **35×10^9** €/y

Who pays for the heterogeneity?

Flow tracing

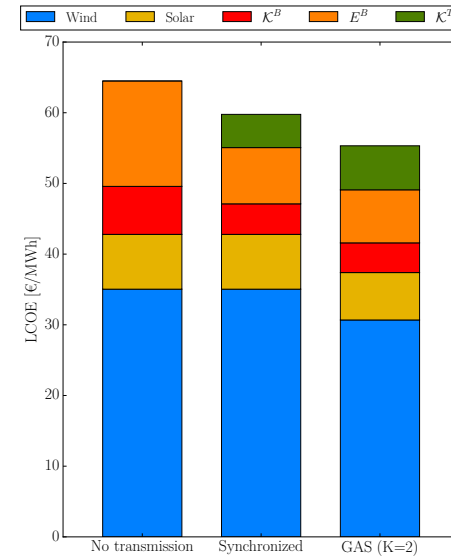
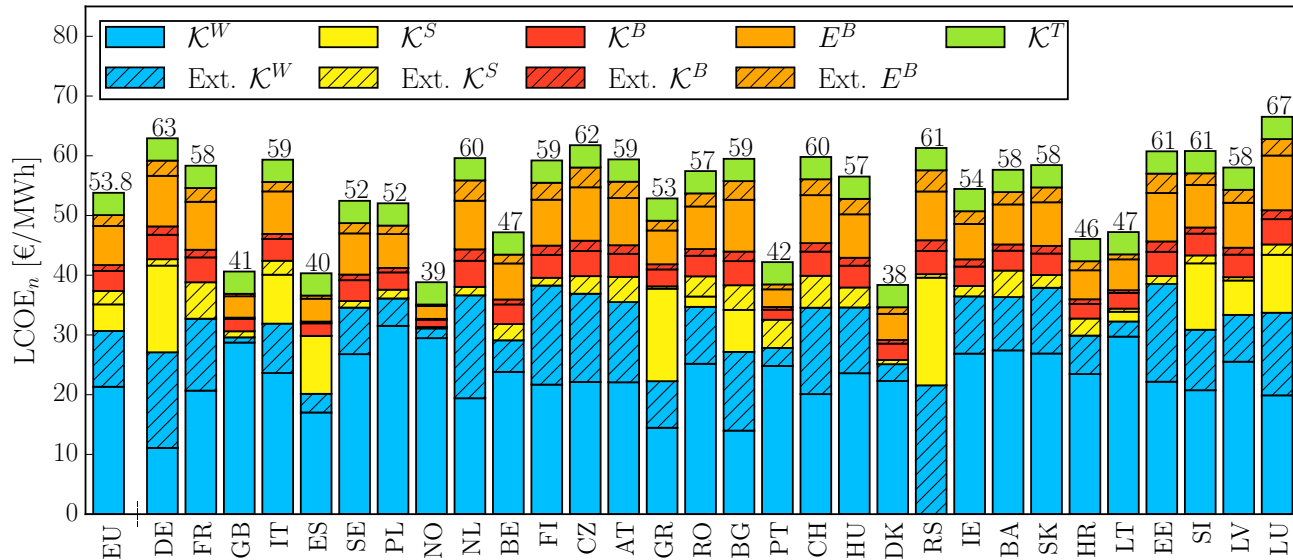
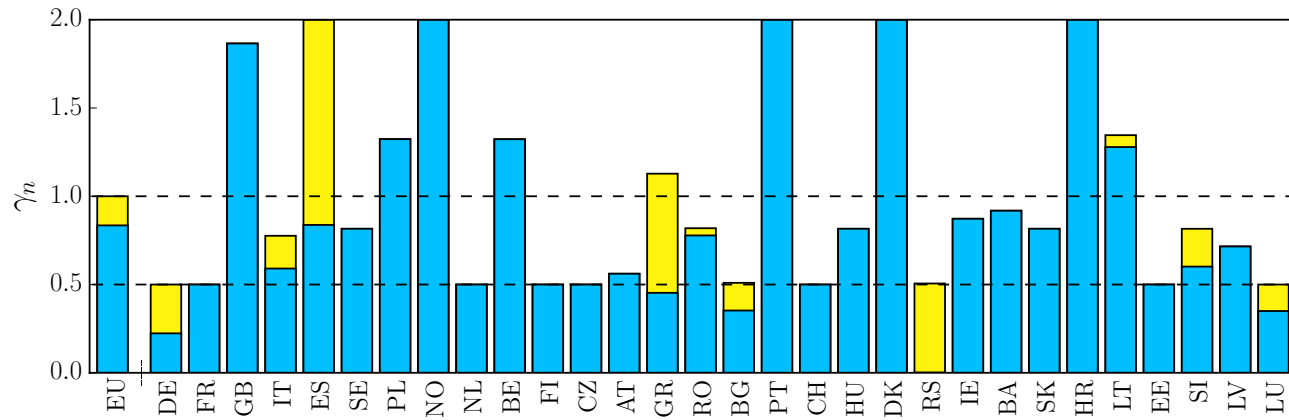


$$K_n^\mu = K_{nn}^\mu + \sum_{m \neq n} K_{nm}^\mu$$

$$\tilde{K}_n^\mu = K_{nn}^\mu + \sum_{m \neq n} K_{mn}^\mu$$

B Tranberg et.al., NJP 17 (2015) 105002
 M Schäfer et.al., Eur.Phys.J.B 90 (2017) 144
 B Tranberg et.al., Energy 150 (2018) 122-33

Benefit of cooperation



$$K_n^\mu = K_{nn}^\mu + \sum_{m \neq n} K_{nm}^\mu$$

$$\tilde{K}_n^\mu = K_{nn}^\mu + \sum_{m \neq n} K_{mn}^\mu$$

$\forall n: LCOE_n^{hom, noT} > LCOE_n^{hom, T} > LCOE_n^{het, T}$

**30 EU friends
you have to be!**



“Gier” nach mehr Physik:

power-flow renormalization

M Schäfer et.al.,
EPL 119 (2017) 38004

storage phase transition

T Jensen + M Greiner,
EPJ ST 223 (2014) 2475-81

principal spatio-temporal patterns

M Raunbak et.al., Energies 10 (2017) 2934
F Hofmann et.al., EPL 124 (2018) 18005

mesoscale turbulence + climate change

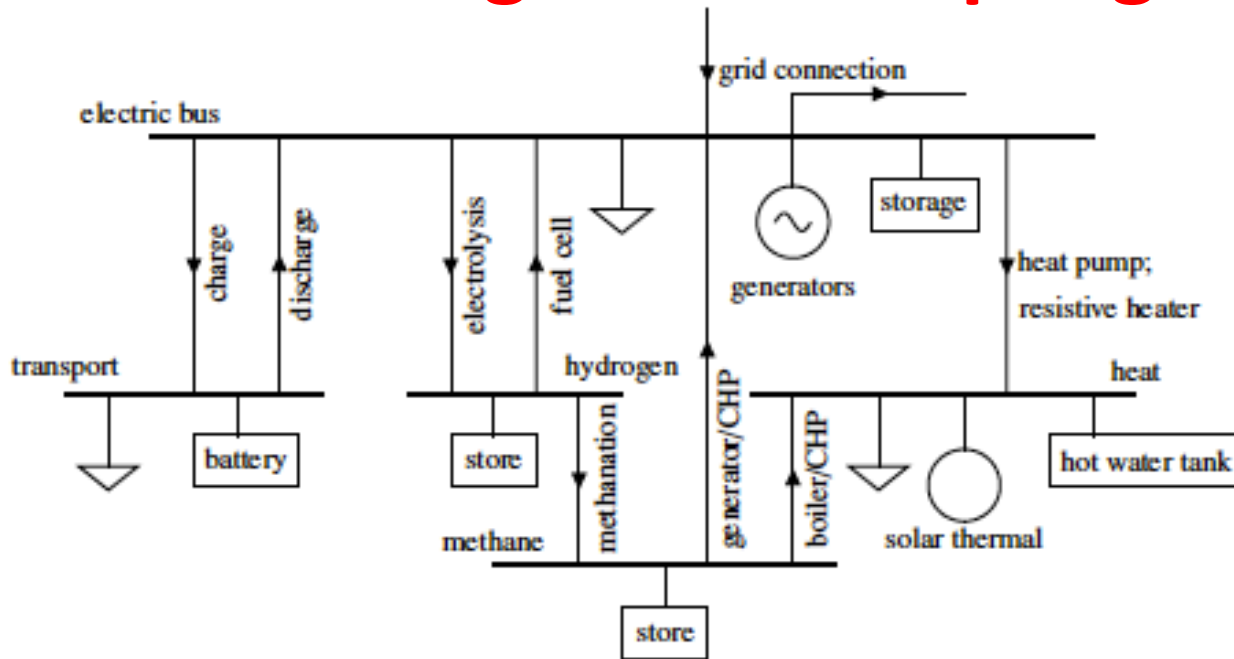
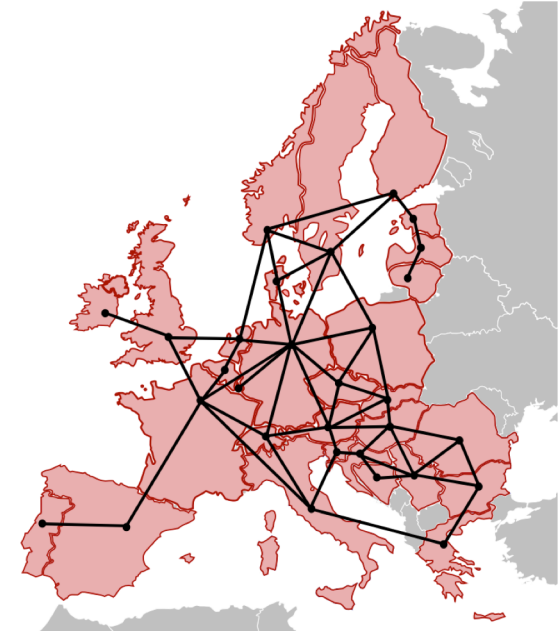
M Schlott, A Kies et.al.,
Applied Energy 230 (2018) 1645-59

flexibility classes

D Schlachtberter et.al.,
Energy Conversion Management 125 (2016) 336-46

II. Advanced Modeling: electricity → “smart” energy network

Decarbonization of the European energy system with strong sector couplings



T Brown et al.,
Energy 160 (2018) 720-39

K Zhu, M Victoria et al.,
Applied Energy 236 (2019) 622-34

T Brown et al.,
Energies 12 (2019) 1032

simplified cross-sector network model

capture / extract general system dynamics + meaningful insights + inspirational results

PyPSA-Eur-Sec-30

Joint capacity + dispatch optimization:

$$\min_{\substack{G_{n,s}, E_{n,s}, \\ F_{\ell}, g_{n,s,t}}} \left[\begin{array}{l} \text{generation costs} \\ \sum_{n,s} c_{n,s} \cdot G_{n,s} \end{array} + \begin{array}{l} \text{storage costs} \\ \sum_{n,s} \hat{c}_{n,s} \cdot E_{n,s} \end{array} + \begin{array}{l} \text{transmission costs} \\ \sum_{\ell} c_{\ell} \cdot F_{\ell} \end{array} + \begin{array}{l} \text{variable costs} \\ \text{(excluding CO}_2 \text{ tax)} \\ \sum_{n,s,t} o_{n,s,t} \cdot g_{n,s,t} \end{array} \right]$$

Subject to constraints :

$$\sum_s g_{n,s,t} + \sum_{\ell} \alpha_{n,\ell,t} \cdot f_{\ell,t} = d_{n,t} \quad \Leftrightarrow \quad \lambda_{n,t} \quad \forall n,t \quad \text{Supply hourly inelastic demand}$$

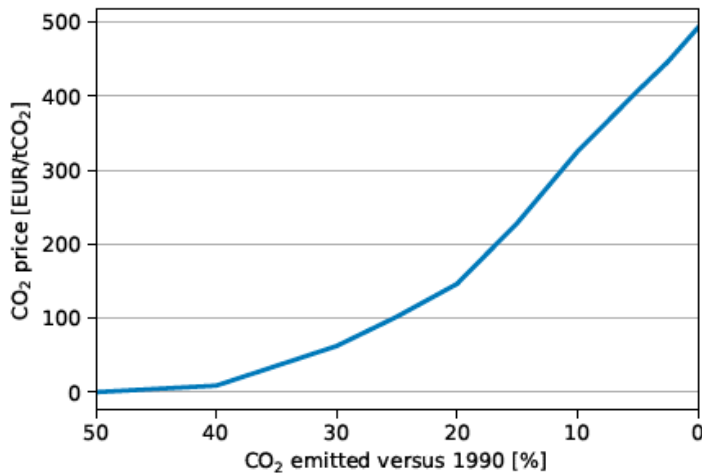
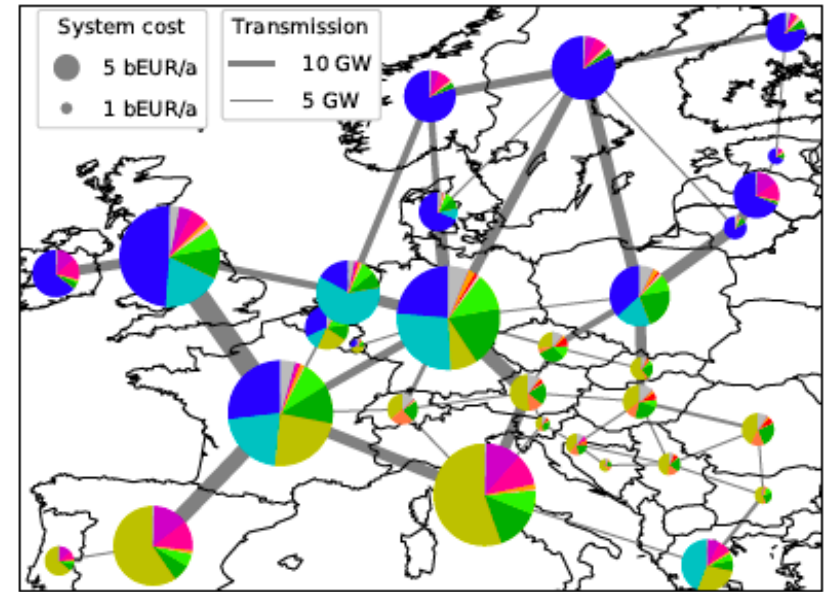
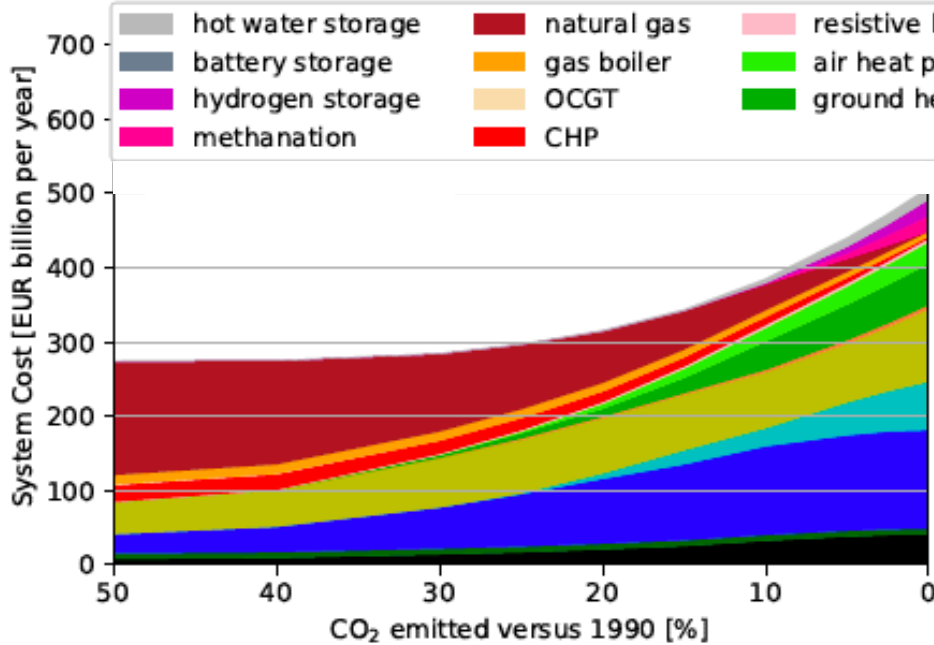
$$\underline{f}_{\ell,t} \cdot F_{\ell} \leq f_{\ell,t} \leq \bar{f}_{\ell,t} \cdot F_{\ell} \quad \forall \ell,t \quad \text{Maximum power flowing through the links}$$

$$\sum_{n,s,t} \varepsilon_s \frac{g_{n,s,t}}{\eta_{n,s}} + \sum_{n,s} \varepsilon_s (e_{n,s,t=0} - e_{n,s,t=T}) \leq \text{CAP}_{\text{CO}_2} \quad \Leftrightarrow \quad \mu_{\text{CO}_2} \quad \text{CO}_2 \text{ emission constraint}$$

$$\min (\text{Costs} - \lambda \cdot \text{Constraint})$$

electricity + heating + transportation

Costs with Optimal Transmission



Aalborg U
+ Aarhus U

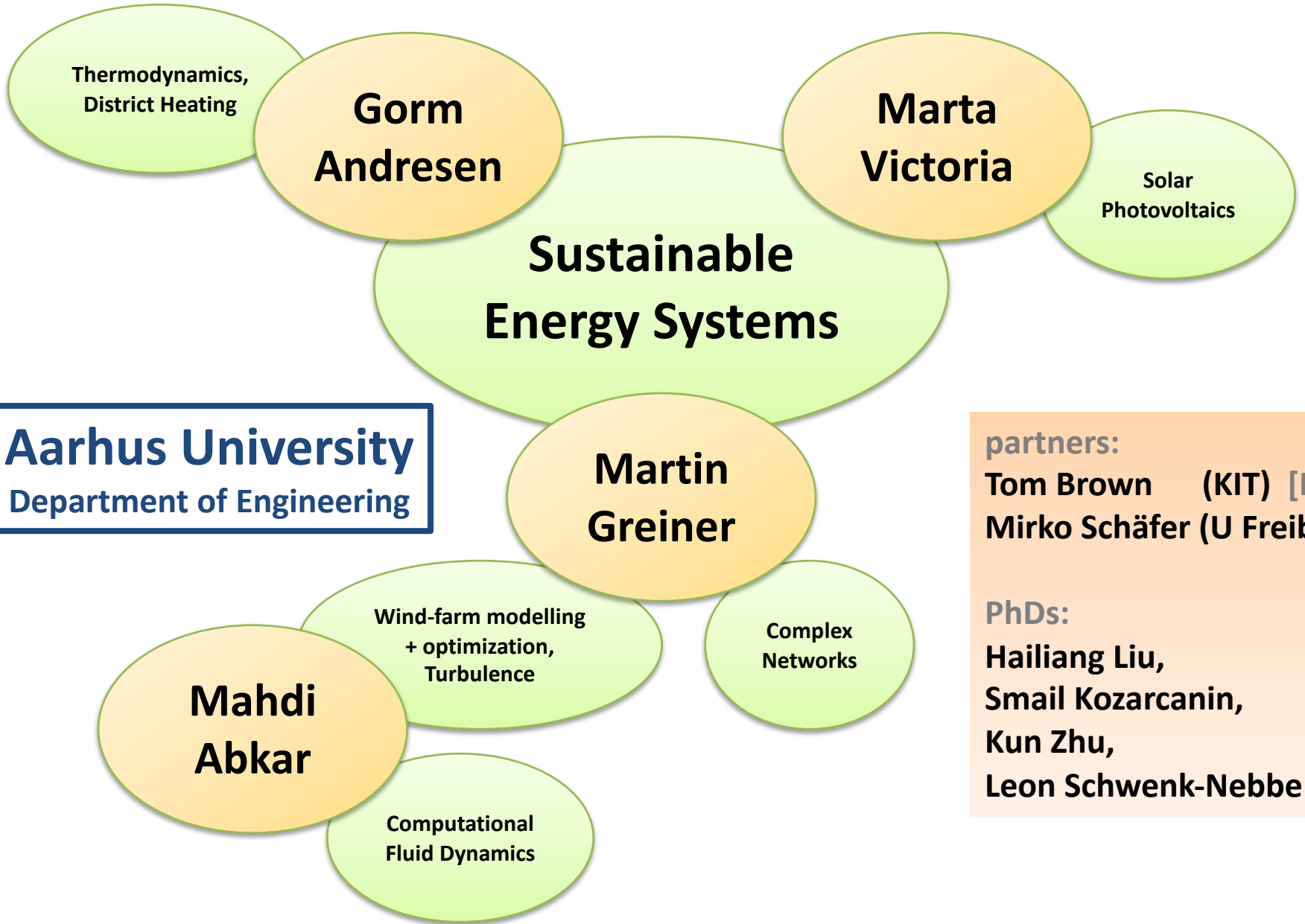
“engineering” challenges:



- **tech+econ transition pathways 2020 → 2050**
- impact of climate change
- large → small scale modelling
- **tech+econ+soc+pol modeling/control/consulting**



(Applied Theoretical) Physics of complex Socio-Economic Systems:
„modelling challenges to boldly go where no one has gone before“



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