

Potential für alpine Energieanlagen im Schweizer Energiesystem



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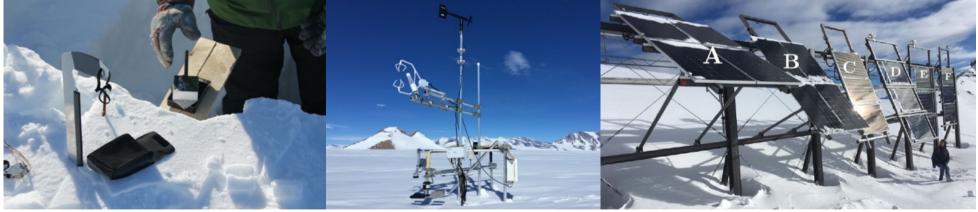




Progress to Report on

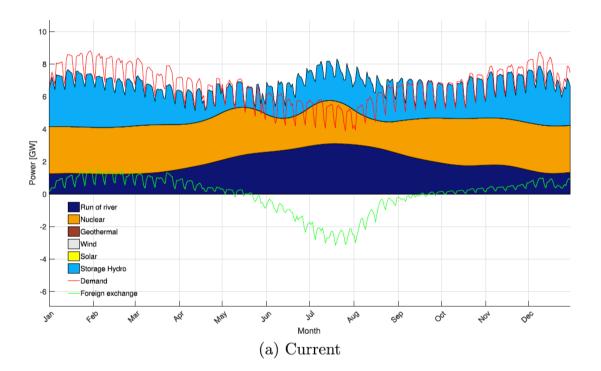
- Assessment of scenarios for the Swiss electricity system (OREES model)
- Looking at forward scattering snow at its effects on the surface
- Very high resolution wind modelling in complex terrain with ML
- Assessing the untapped potential of renewable energy in the Alps







Starting from National Scenarios for PV and Wind



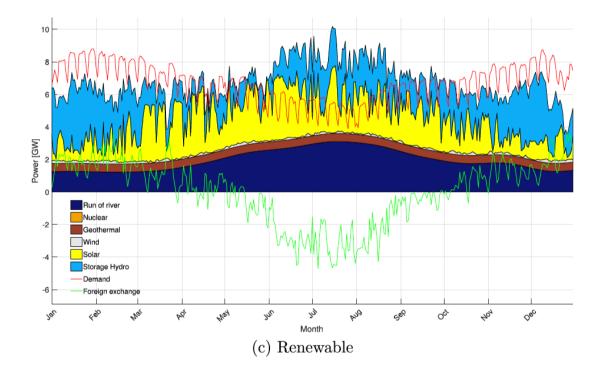


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Bartlett, et al. (2018), Charting the course: A possible route to a fully renewable Swiss power system, *Energy*.



Nuclear Power replaced by wind (little) and PV



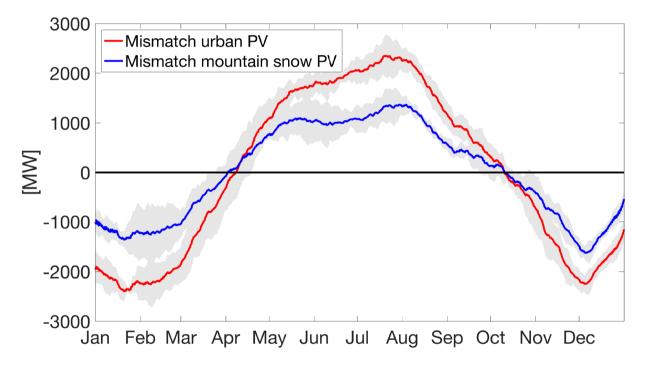


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How to reduce the seasonal mismatch?



→ Need to fully understand potential of mountain PV:

- Incoming Radiation
- Existing Infrastructure
- Acceptance
- Finance









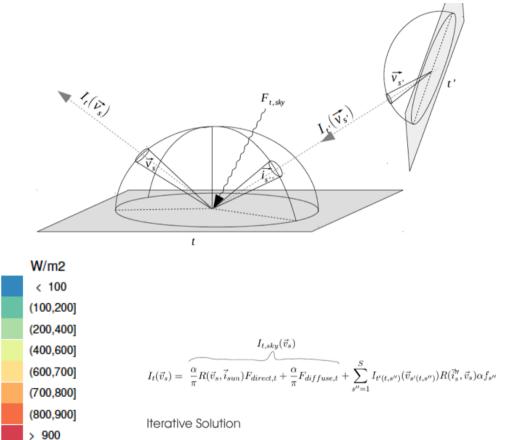
Consider the role of Snow in Radiative Transfer

- Snow scatters light (high albedo) but does so anisotropically
- Not considered in any model of radiative transfer
- Combine snow modelling with near surface atmospheric modelling to assess influence on surface energy balance and yield of solar panels (PV) in snow environments
- Current models perform poorly

Domain total = 546.6 W/m2

Total Irradiance

• Use for potential assessment and operational forecasting



$$I_{t,k+1}(\vec{v}_s) = I_{t,sky}(\vec{v}_s) + \sum_{s''=1}^{S} I_{t'(t,s''),k}(\vec{v}_{s'(t,s'')}) R(\vec{i}_s'',\vec{v}_s) \alpha f_{s''}$$

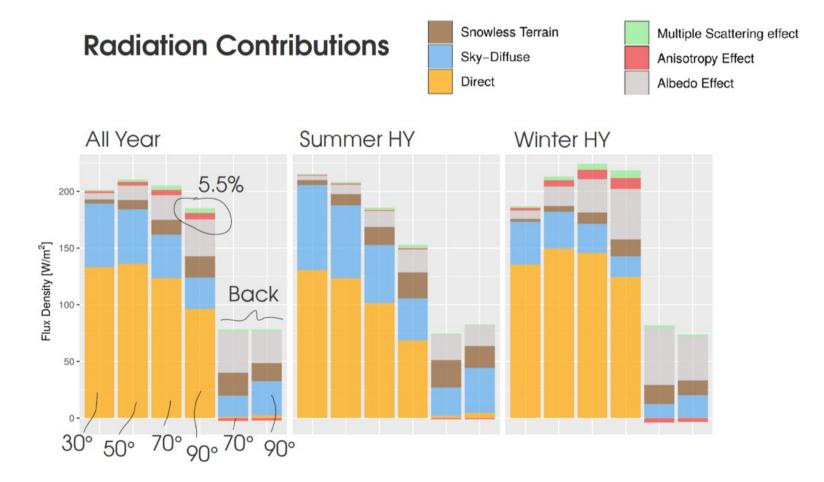
Von Rütte, Kahl and Lehning, JGR, 2021



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Consider the role of Snow in Radiative Transfer



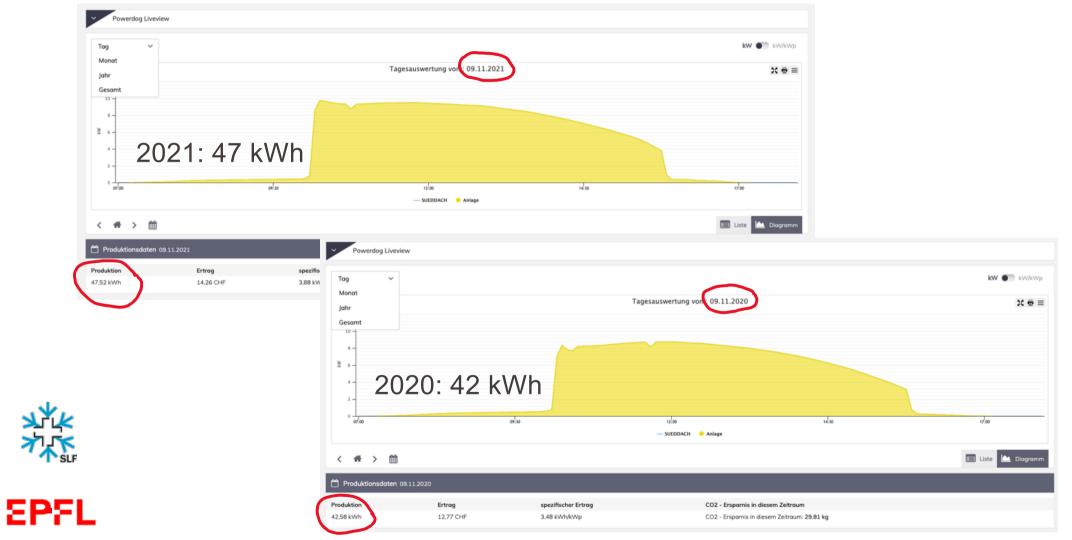


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Von Rütte, Kahl and Lehning, JGR, 2021

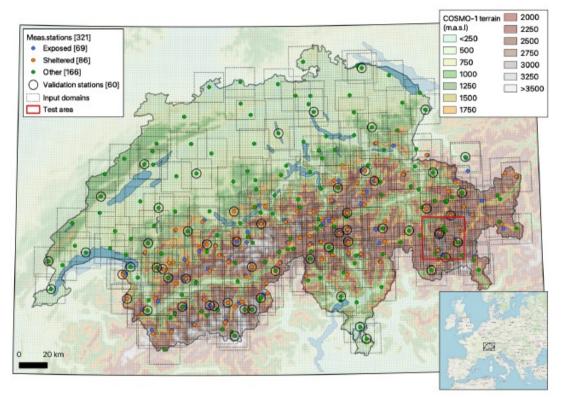


Does it matter in practice? – Even on a Rooftop





Make HR Wind Fields Computationally Affordable with ML





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Dujardin and Lehning, QJRMS, 2022

- Try machine learning (ML) and specifically convolutional neural networks with training on weather station data (MeteoCH and IMIS)
- **Topographic Parameters** determined in an environment around the grid point of interest and characterizing wind exposure → make the wind turn as it hits topography
- Find a suitable **Loss Function**, which not only minimizes the error but also preserves the distributions of modelled winds as expressed by e.g. Weibull parameters

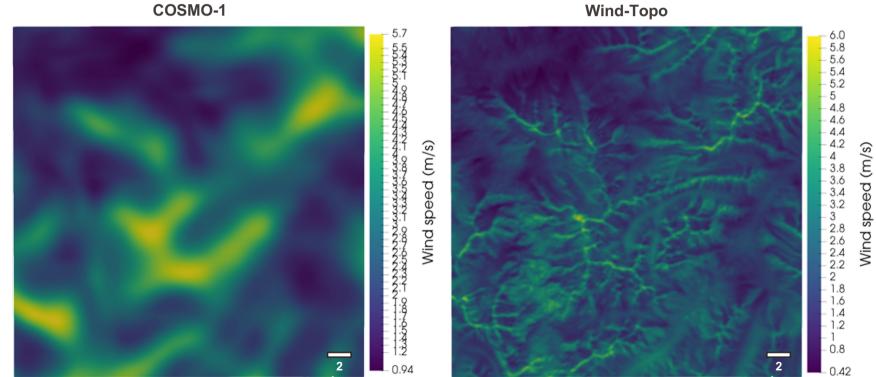




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Make HR Wind Fields Computationally Affordable with ML

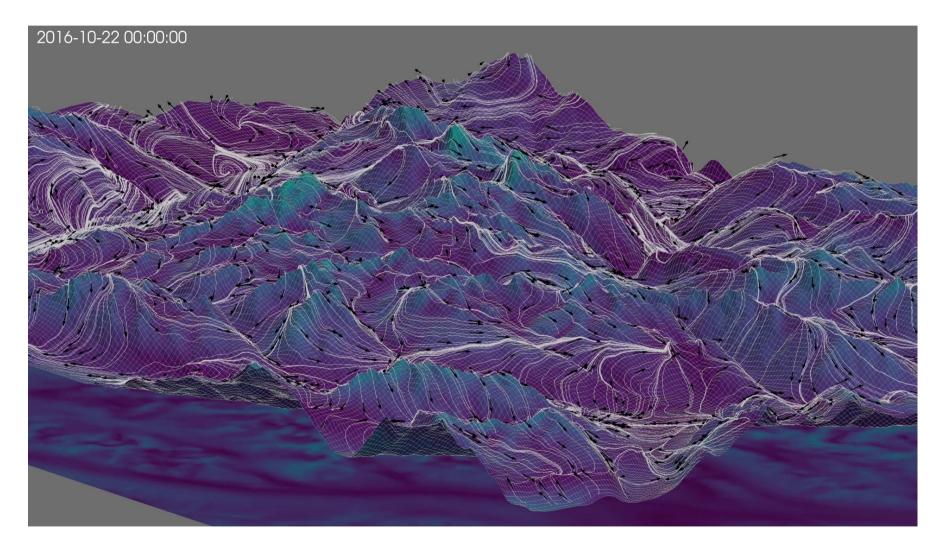
Average wind speed (1 year, hourly)



COSMO-1



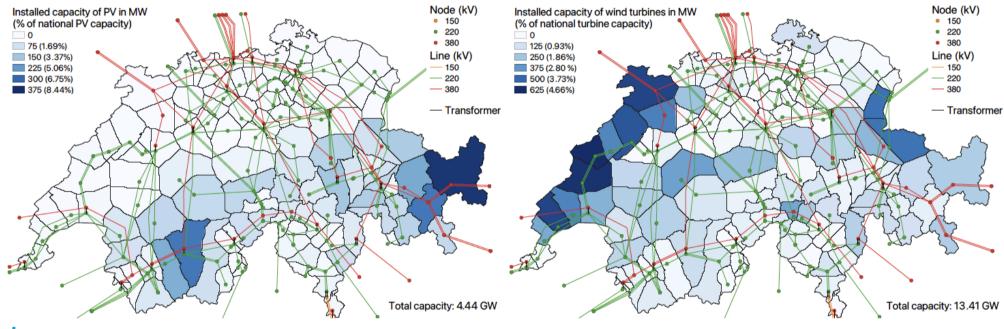
Make HR Wind Fields Computationally Affordable with ML







What is the optimal placement of PV and Wind? (for self-sufficiency)

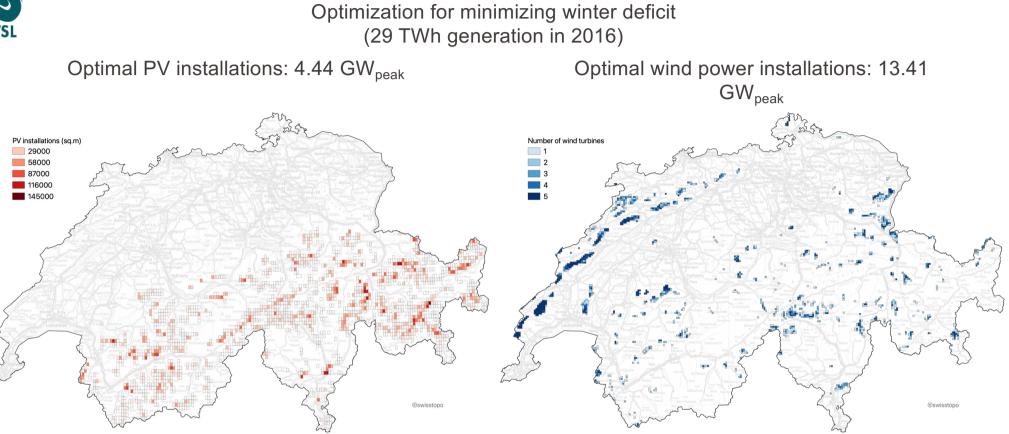




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Dujardin, J., A. Kahl, and M. Lehning (2021), Synergistic optimization of renewable energy installations through evolution strategy, *Environ Res Lett*, *16*(6), doi: ARTN 064016, 10.1088/1748-9326/abfc75.







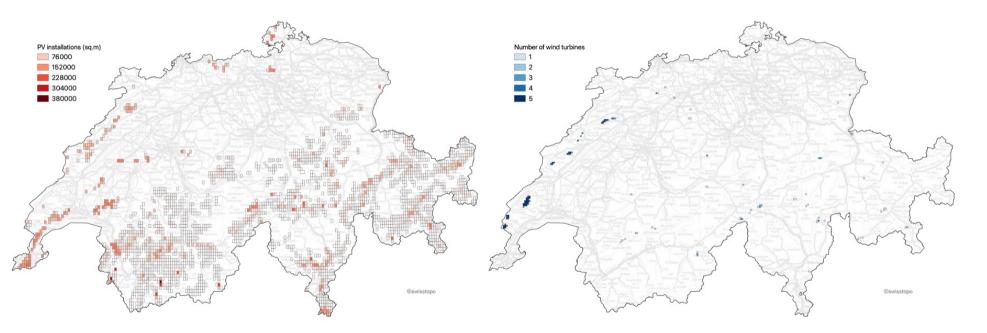




Optimization for maximizing revenues (25 TWh generation in 2019)

Optimal PV installations: 13.64 GW_{peak}

Optimal wind power installations: $1.18 \text{ GW}_{\text{peak}}$





Fresh from the Press: Jérôme Dujardin





Actionable Conclusion: Facilitate High-Mountain Installations of both PV and Wind

Recipe of Success:

100 km² of PV Panels (50% in the mountains) – 10 GW

1000 Wind Turbines – 3 GW

→ 10 TWh of Additional Storage





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From: https://twitter.com/ParmelinG/status/1528745252500164609