

Task 4.2

Title

Global observatory of electricity resources

Projects (presented on the following pages)

Transformation of the Energy-related Severe Accident Database (ENSAD) into an interactive, web-based GIS application

Poster see task 4.1

P. Burgherr, W. Kim, M. Spada, A. Kalinina, S. Hirschberg

Bi-level electricity market model (BEM)

M. Densing, E. Panos

Optimization of photovoltaic potential and its integration in Switzerland using genetic algorithm and optimal power flow

J. Dujardin, A. Kahl, B. Kruyt, M. Lehning

World Energy Scenarios 2016

T. Kober, E. Panos

A preliminary Spatial Multi-Criteria Decision Analysis for Deep Geothermal Systems in Switzerland

Poster see task 4.1

M. Spada, P. Burgherr

Marginal electricity supply mixes and their integration in version 3.4 of the ecoinvent database: results and sensitivity to key parameters

L. Vandepaer, K. Treyer, C. Mutel, C. Bauer, B. Amor



SWISS COMPETENCE CENTER for ENERGY RESEARCH
SUPPLY of ELECTRICITY

PAUL SCHERRER INSTITUT



Bi-level electricity market model (BEM)

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In cooperation with the CTI

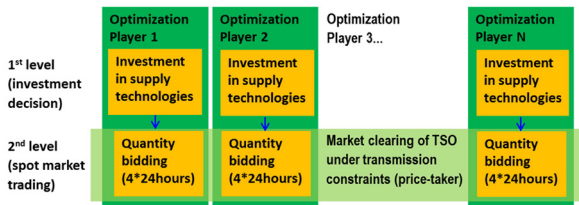


Motivation

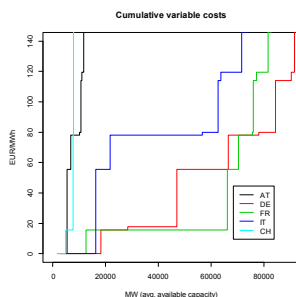
- Goal: Decision-support for policy makers (ES2050 and beyond): Improved understanding of investment, production and trading decisions of producers on the European electricity market, especially for Switzerland
- Focus: Electricity producer-side (not consumer-side)
- Oligopolistic market modelling is required, because producers (in corpore, or single utilities) influence prices: E.g.
 - Producers withhold production, or limit investment to drive prices up deliberately, or are forced by technical or regulator's outages
 - Market power may be exerted only in some sub-markets having scarcity effects (e.g. during peak-hours)
- **Research questions:**
 - How can we capture the **volatility of the electricity price** with a numerical model that is also suitable for academic purposes (without infeasible parametrization efforts, e.g. modelling each plant separately and each day's idiosyncratic market/demand situation)
 - Can we understand **profit-oriented investment** behaviour?
- Partners (Projects also with **BFE-SFOE** and with **VSE**):
 - Chair of Quantitative Business Administration, UZH
 - Energy Economics, Uni Basel (Data harmonization)

Method

- Multi-leader-follower game: Investment and subsequent production decision of several power producers
- Complements PSI's energy-system cost-optimization models
- Producers can influence prices by withholding investment or production capacity in certain load periods

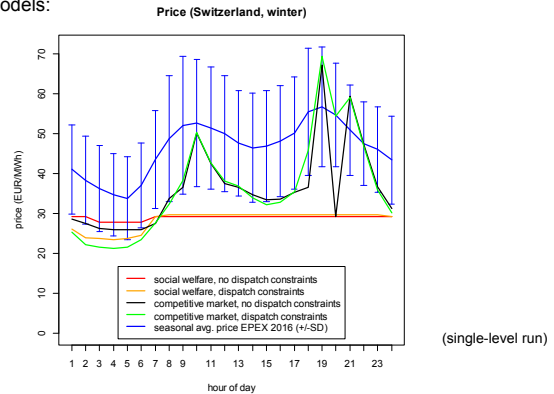


- Bi-level Nash-Cournot Game for electricity market
- General framework model with several operation sub-modes: (i) Investment-decision and production-decision on same level (ii) Single scenario (deterministic) (iii) Social welfare maximization (price-taker, marginal cost perspective)
- Transmission constraints between players: DC (linear) flow model
- Wholesale consumers represented via demand-price elasticity on spot market. Additional in-elastic perfect-competition market (OTC)
- Hourly trading over a day in four seasons of a future representative year: (24*4 = 96 trading hours = load periods)
- **Base configuration: Players are countries, i.e. each player has country specific generation portfolio**
- Input: CAPEX & OPEX per technology, seasonal availability, merit-order curves (=cumulative variable costs) per country, etc.

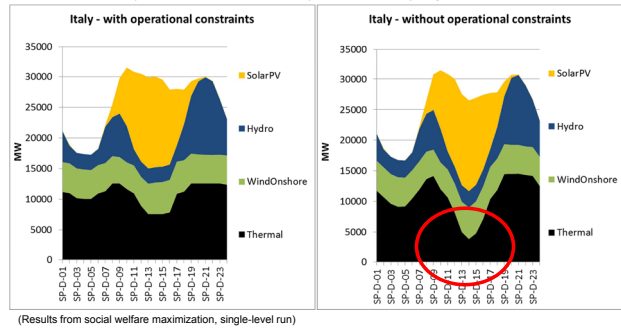


Results (Preliminary)

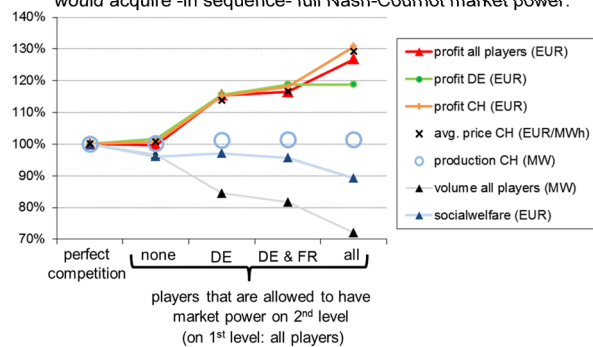
- Models with competitive market representation can explain price volatility better than (aggregated) social welfare maximization models:



- Representation of dispatch constraints on thermal generation is needed. Without such constraints, flexibility is overestimated, e.g. combined-cycle plants start freely without paying for start-up costs:



- Test of Bi-level game: What if the supply portfolio of the countries would acquire -in sequence- full Nash-Cournot market power:



Outlook

- Status of project: Model operational, first results are obtained
- Stochasticity, geographical expansion (EU), several investment steps

References

- Densing, M., Panos, E., Schmedders, K. (2017). Stochastic bi-level electricity market modeling, *2nd Workshop of SET-Nav WP10 Modelling Forum*, ETHZ
- Densing, M., Panos, E., Schmedders, K. (2015). Decision making in electricity markets: Bi-level games and stochastic programming, *ESC Workshop*, ETHZ
https://www.psi.ch/eem/ConferencesTabelle/BilevelAndSP_MartinDensing_TALK.pdf
- Densing, M., Panos, E., Schmedders, K. (2015). Bilevel oligopolistic electricity market models: The case of Switzerland and surrounding countries, *OR2015*, Vienna

Optimization of photovoltaic potential and its integration in Switzerland using genetic algorithm and optimal power flow

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How to harvest solar energy effectively?

Given the natural resources?

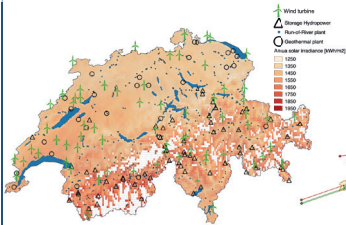


Fig.1. Annual solar irradiance and other electricity generation sources

Given the transmission grid?

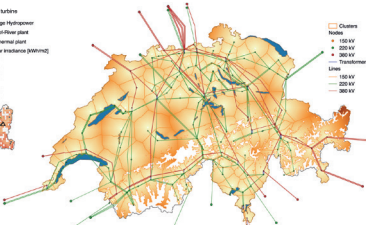


Fig.2. Transmission grid (as planned for 2025)

Hybrid deterministic / stochastic approach

1. Local photovoltaic (PV) configuration for **maximum yield**: Differs from the classical setup (south at 39° tilt)

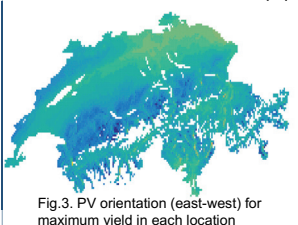


Fig.3. PV orientation (east-west) for maximum yield in each location

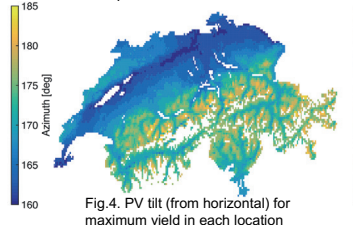


Fig.4. PV tilt (from horizontal) for maximum yield in each location

→ **Local settings** of PV panels are set, independently of the global configuration (locations within the country)

2. **Local specificities** of PV production:

- Annual yield
- Reduction of annual required import, due to higher winter production [1]
- Stability of annual production
- Stability of winter production

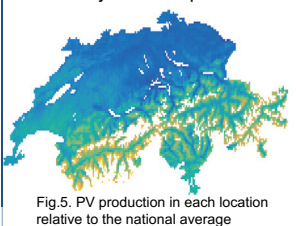


Fig.5. PV production in each location relative to the national average

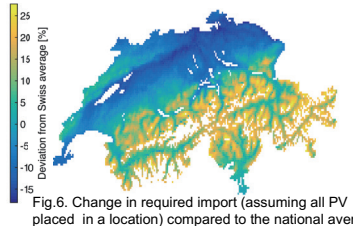


Fig.6. Change in required import (assuming all PV placed in a location) compared to the national average

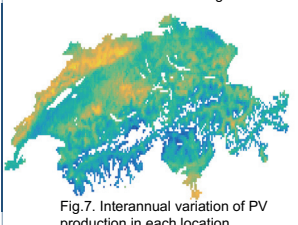


Fig.7. Interannual variation of PV production in each location

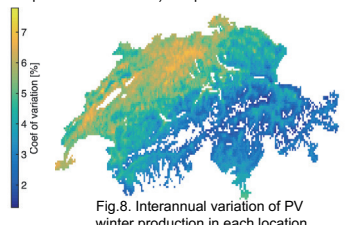


Fig.8. Interannual variation of PV winter production in each location

→ **Multi-objective function** for the genetic algorithm
→ Selection of the best locations within the clusters

3. **Maximum PV coverage** in each pixel based on CORINE land surface cover type.

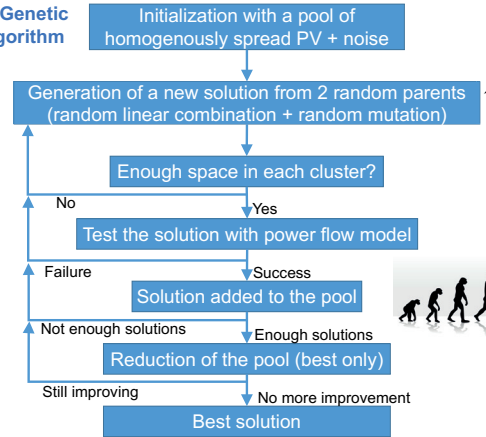
Land type	Urban	Industry	Pasture	Agriculture	Open
Max coverage	10%	10%	5%	5%	5%

→ Amounts to 670 km² (~10 times the required PV area)
→ Leaves freedom to the genetic algorithm

Constrained to altitudes below 2500m

→ **Constraints** on the solutions found by the genetic algorithm (upper limit of PV installed in each cluster)
→ Selection of the available locations within the clusters

4. Genetic algorithm



Results

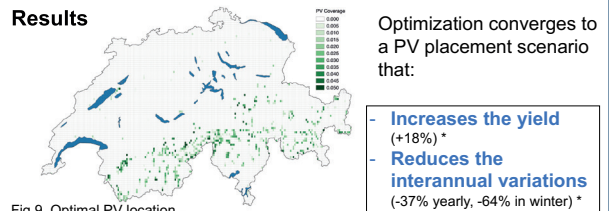


Fig.9. Optimal PV location

Optimization converges to a PV placement scenario that:

- **Increases the yield** (+18%)*
- **Reduces the interannual variations** (-37% yearly, -64% in winter)*

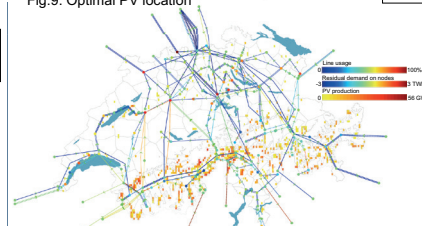


Fig.10. Max line usage

- **Never exceeds the line capacity**

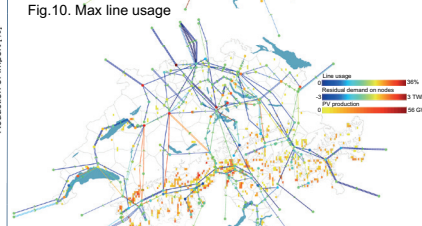


Fig.11. Mean line usage

- **Reduces the required import** (-17%)*

* Compared to a PV placement scenario proportional to population density

Perspectives

- Improvements on the definition of potential PV area by using more GIS products (access from road, complexity of the terrain).
- Increase resolution of topographic shading for better irradiance computation in **complex terrain**.
- Apply optimization strategies to **wind energy** as well.

References

[1] National energy balance model from: J Dujardin et al. Interplay between photovoltaic, wind energy and storage hydropower in a fully renewable Switzerland. Energy, vol135, p513-525, 2017

Data

- PV production time series based on satellite-derived irradiance (MeteoSwiss)
- Wind production time series based on wind speed measurements (MeteoSwiss)
- Demand time series from Swissgrid (publicly available on their website)
- Run-of-the-river monthly production and reservoirs' inflow from the Swiss Federal Office of Energy (SFOE) and PREVAH model (WSL)
- Storage / pumped hydropower characteristics from WASTA database (SFOE)

Acknowledgments

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Marginal electricity supply mixes and their integration in version 3.4 of the ecoinvent database : results and sensitivity to key parameters

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Introduction & objectives

- Marginal electricity supply mixes provided in previous versions of the ecoinvent database were based on historical data and limited by database features. This did not allow to capture accurately the consequences of additional demand for electricity given the complex nature of power markets.
- Objectives:
 - Provide long-term and consistent marginal electricity supply mixes based on energy scenarios to take into account future market trends and constraints.
 - Perform several sensitivity analyses to understand the influence of the key parameters and methodological choices on the mix composition and their corresponding environmental impacts.

Methodology

- The calculation method used to determine the marginal electricity supply mixes originates from (Schmidt et al. 2011; Muñoz 2015) :

$$Share_{i,TH} = 100 \cdot \frac{P_{i,TH} - P_{i,ref}}{\sum_n (P_{i,TH} - P_{i,ref})}$$

Where:

i: electricity producing technology

TH: the year chosen as time horizon

ref: the year chosen as a reference for the time of the decision

P: the quantity of electricity generated at time "TH" or "Ref" by technology *i*

n: includes all unconstrained electricity producing technologies with an increased production at TH with respect to ref

Share i: the percentage that supplier *i* contribute to the marginal mix

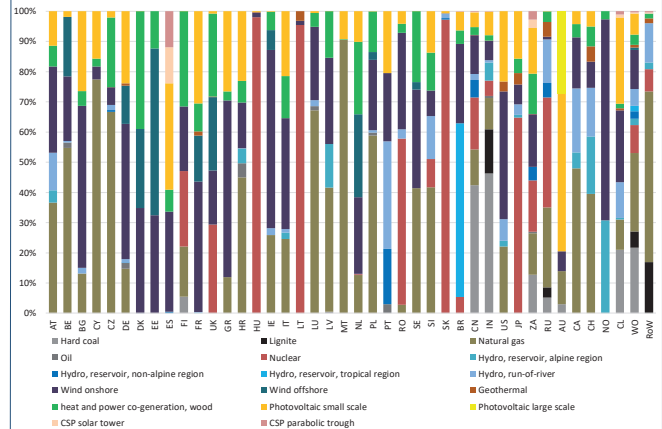
- The reference year is 2015 and time horizon is 2030.
- Public energy projections realized by national or supra-national official agencies are used as a source of data (e.g. European Commission, International Energy Agency, Energy Information Administration).
- Additional processes are created for missing activities.

Geographical coverage



- The long-term marginal electricity supply mixes of 40 countries are updated and integrated in version 3.4 of the ecoinvent database. These markets correspond to ~76.5 % of the global electricity production in 2015 (~76.9% in 2030).

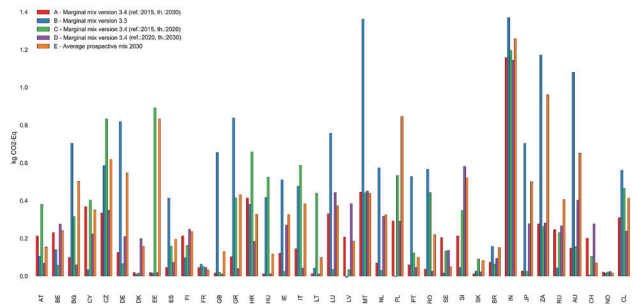
Marginal mixes composition



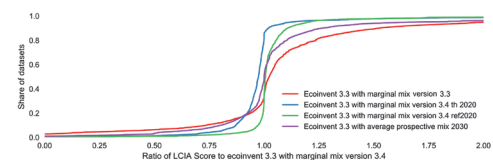
- The marginal mixes are composed on average by 29% fossil fuel power, 14% nuclear and 58% renewables

Meta-sensitivity analyses

- Different approaches to define the marginal mixes are tested : original v 3.3 approach, time horizon = 2020, reference year = 2020, average mix 2030.
- This compares the global warming potential (IPCC 2013, GWP 100a) of every activity of ecoinvent v.3.3 updated with v.3.4 marginal mixes to versions of ecoinvent v.3.3 generated with the different approaches.



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References

Muñoz I (2015) Example – Marginal Electricity in Denmark. In: consequential-lca.org. <http://consequential-lca.org/clca/marginal-suppliers/the-special-case-of-electricity/example-marginal-electricity-in-denmark/>. Accessed 15 Nov 2016

Schmidt JH, Thrane M, Merciai S, Dalgaard R (2011) Inventory of country specific electricity in LCA - consequential and attributional scenarios. Aalborg