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Integrated energy system modelling:

Linking the electricity generation and demand side through heat-pump demand response

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Electricity generation system operation and planning





Electricity generation system operation and planning

- LUSYM framework
 - Operational model
 - Mixed-integer programming for unit commitment
 - Unpredictability: forecast errors, reserve sizing/allocation/activation
 - New technologies: e.g., CCS, active grid elements, power-to-gas
 - Expansion planning models TIMES framework
 - Improve operational representation and technical detail
 - Uncertainty and market elements
 - Energy policy and market applications



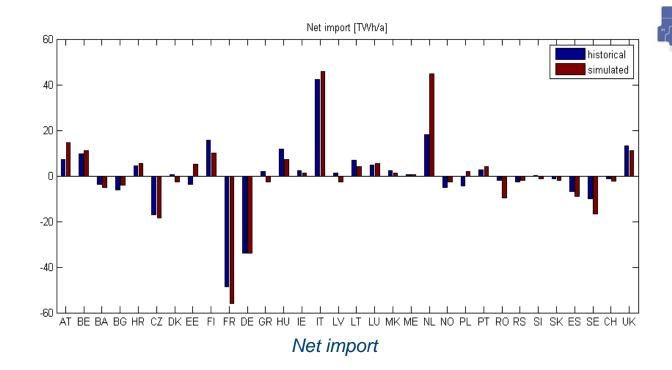
LUSYM: Set-up of the standard unit commitment model

- Mixed Integer Linear Programming
 - Solve on HPC
 - Large-scale model set up
 - Tight and compact formulation
 - "Flexible" model

• Computation time is important factor



LUSYM: Validation of the standard unit commitment model

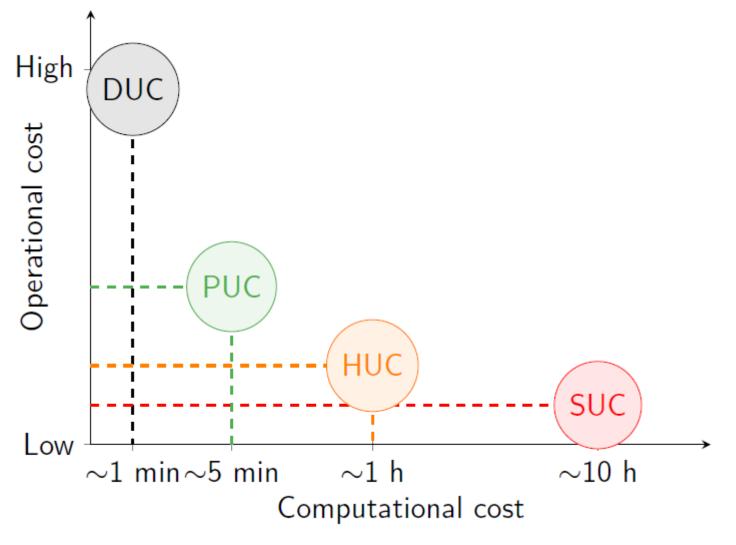


• Calibration on the European system (ENTSO-E area, 2013)



RG Nordi

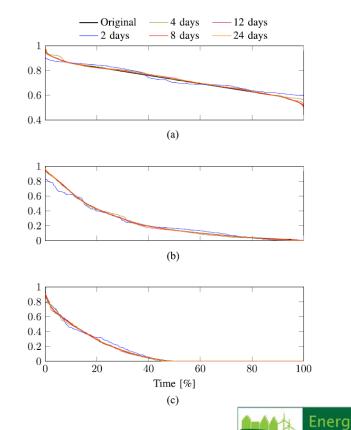
LUSYM: Uncertainty in the unit commitment model





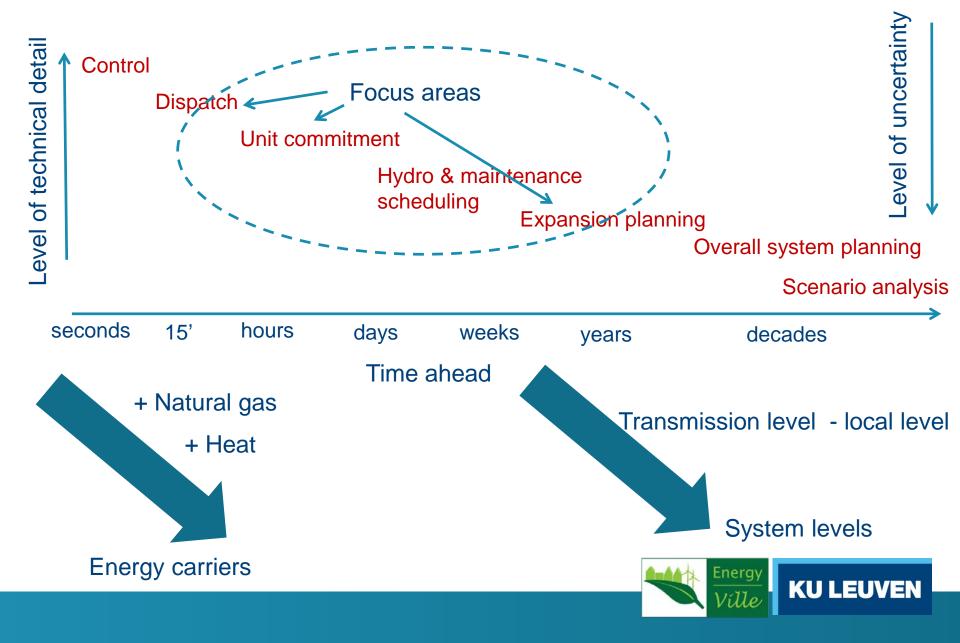
LUSYM: Expansion planning model

- Increasing the level of temporal detail
 - Optimization to select set of representative days
 - Duration curves
 - Load
 - RES series
 - Ramping
 - Correlations
 - ...

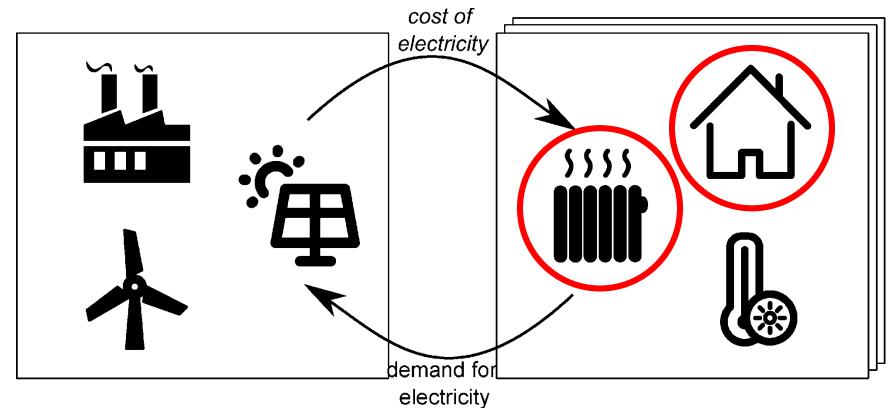


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Electricity generation system operation and planning

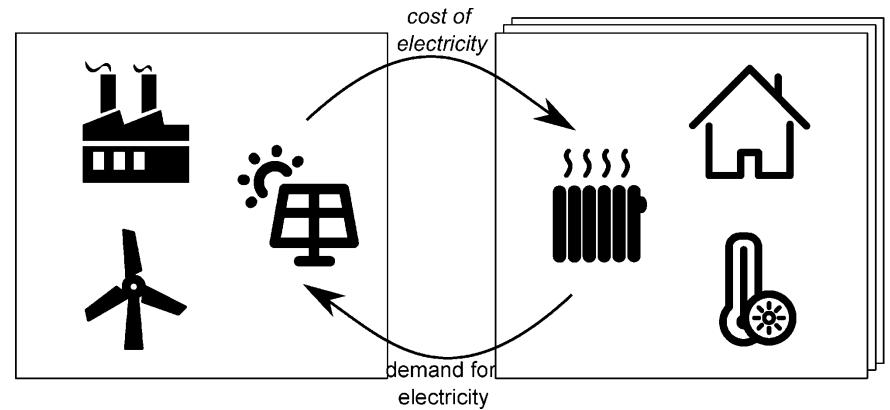


Scope & motivation



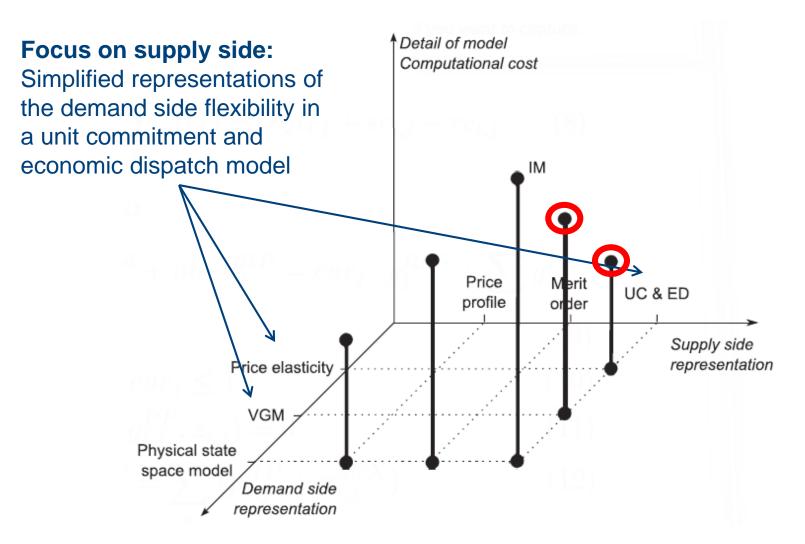
Conventional & stochastic RES-based electricity generation Thermal inertia allows decoupling the electrical demand and the thermal demand without loss of comfort



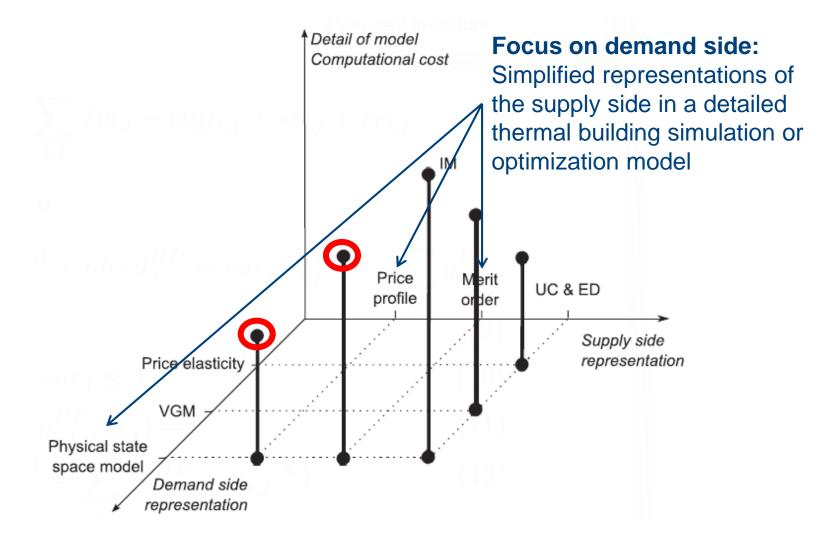


Complex interactions between demand and supply: how do you capture this in an operational model?

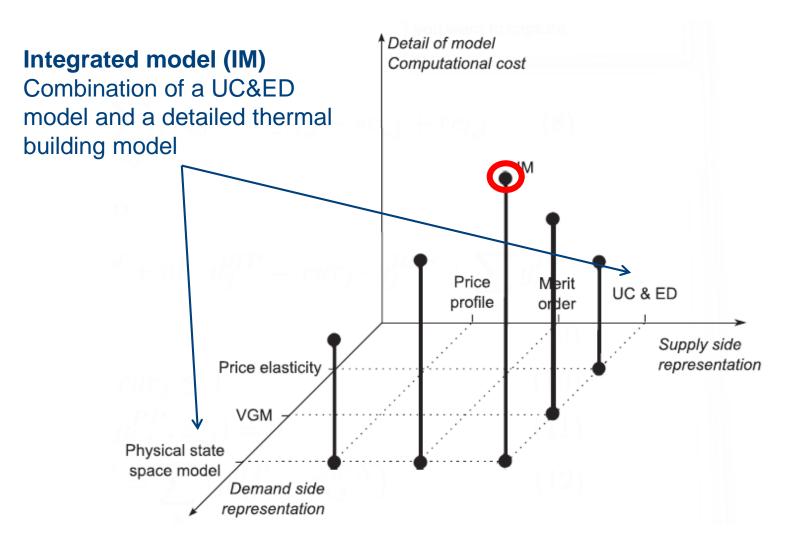








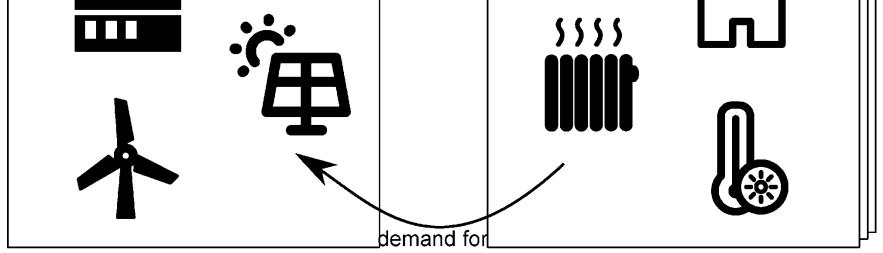






An integrated model

Joint optimization: minimize total operational cost cost of electricity state of state of



electricity

UC & ED model, considering set of power plants, RES-based generation and a fixed demand profile (MILP) **DR-adherent demand model:**

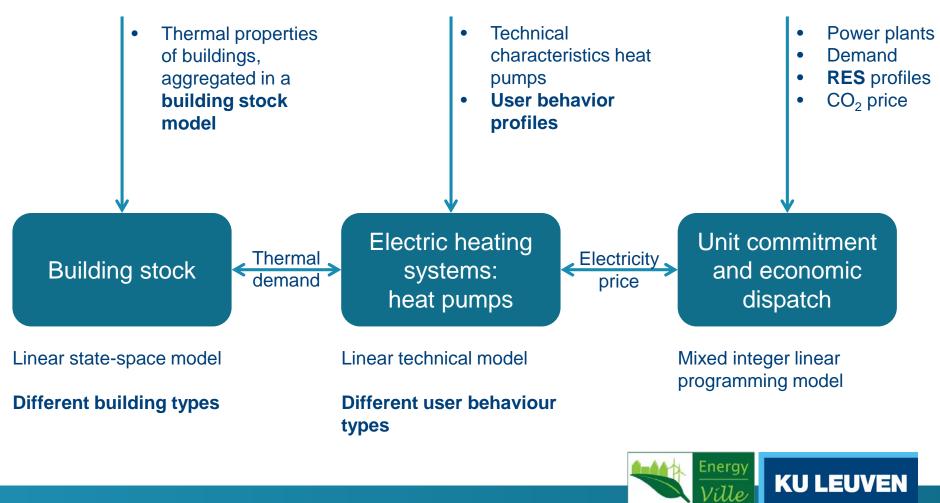
RC network (thermal dynamics building), linear heat pump model, user behavior & external gains (LP)



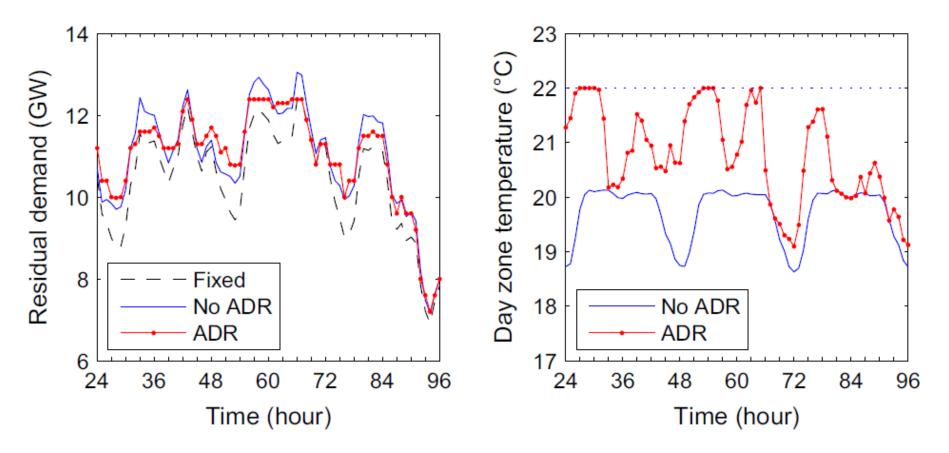
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An integrated model

Joint optimization: minimize total operational cost



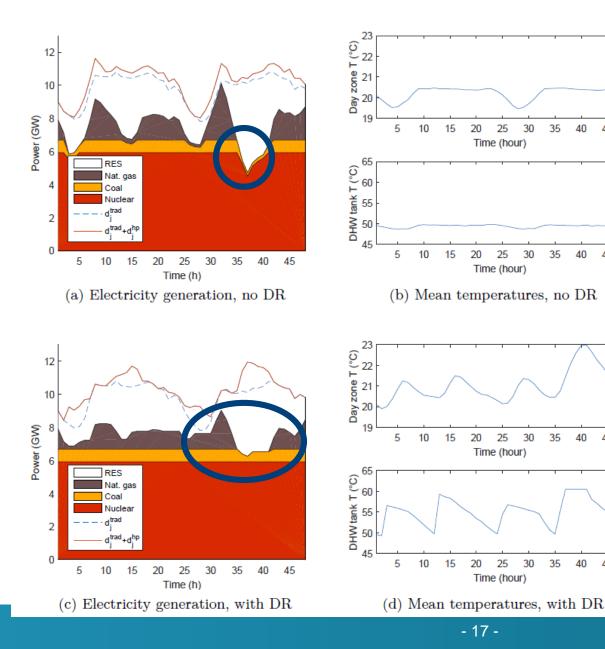
An integrated model: a first example



- Power system inspired on possible future setting of BE power system;
- 250,000 heat pumps;
- 52 user behavior profiles.



An integrated model: a second example



Case study:

35 40 45

35

35

35 40 45

40 45

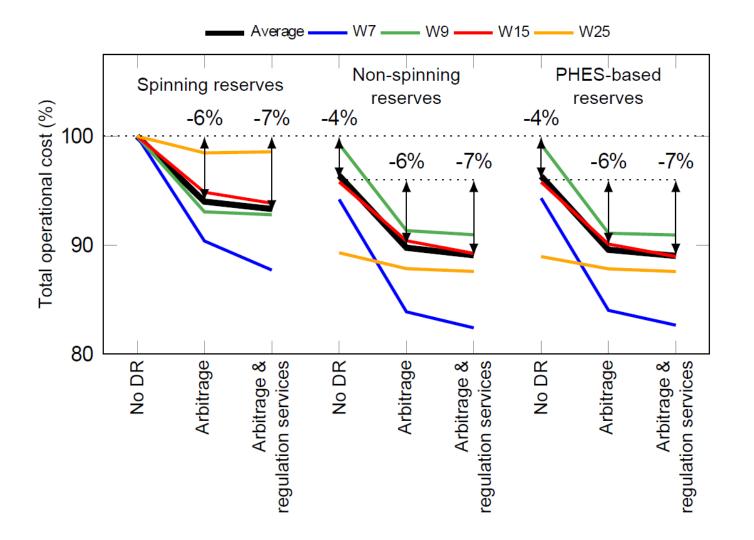
40 45

- Power system inspired on possible future setting of BE power system;
- 250,000 heat pumps;
- **Building properties** represented via an 'average' building (detached dwelling);
- 52 user behavior profiles.

Energy

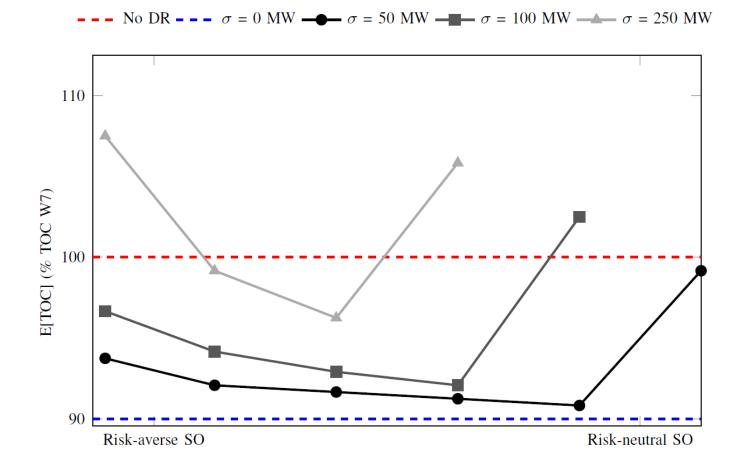
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Value of DR-based arbitrage and regulation services





Impact of limited DR-controllability



From: Bruninx, K., Dvorkin, Y., Delarue, E., D'haeseleer, W., Kirschen, D. Valuing Demand Response Controllability via Chance Constrained Programming. IEEE Transactions on Sustainable Energy, 2017, in press.



Conclusion

Integrated modelling framework

- Operational demand and supply side model formulated using MILP
- More accurate representation w.r.t. other methods
 - Merit order model provides valuable results at much lower computational cost
- Myriad of applications possible

2 Demand response with heat pumps

- Could hold significant environmental and economical advantages: operational cost savings, (additional) peak demand reduction, cost-effective regulation services
- Current modeling provides upper bound
- Controllability needs being accounted for

3 Future work

- Impact on heating system design and life time
- Heterogeneity of DR-loads, user behavior, building types
- Conflicting objectives building owner system operator
- Long term system adequacy



Further reading

[1] D. Patteeuw et al., *Integrated modeling of active demand response with electric heating systems coupled to thermal energy storage systems*, Applied Energy 151, pp. 306-319, 2015.

[2] D. Patteeuw et al., *CO2-abatement cost of residential heat pumps with Active Demand Response: demand-and supply-side effects*, Applied Energy 156, pp. 490-501, 2015.

[3] A. Arteconi et al., *Active demand response with electric heating systems: impact of market penetration*, Applied Energy 177, 636-648, 2016.

[4] K. Bruninx, E. Delarue (co-supervisor) and W. D'haeseleer (supervisor), *Improved modeling of unit commitment decisions under uncertainty*, PhD thesis, KU Leuven, May 2016.

[5] D. Patteeuw and L. Helsen (supervisor), *Demand response for residential heat pumps in interaction with the electricity generation system*, PhD thesis, KU Leuven, September 2016.

[6] K. Bruninx, et al., Valuing Demand Response Controllability via Chance Constrained Programming. Under Review at IEEE Transactions on Sustainable Energy, 2017.

