



Automated Demand Response Strategies for Market Participation and Grid Management

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Outline

- General Aspects of Automated Demand Response
- Exemplary DR Strategies
 - Coordination & Dispatch of Thermostatically Controlled Loads (TCLs)
 - PV Self-Consumption Optimization
- Development of a Smart Distribution Grid Simulator
- Conclusions

General Aspects of Automated Demand Response

Our Definition of Automated DR:

- A directed influence on inherently flexible loads on the customers' premises triggered by externally or internally computed signals.

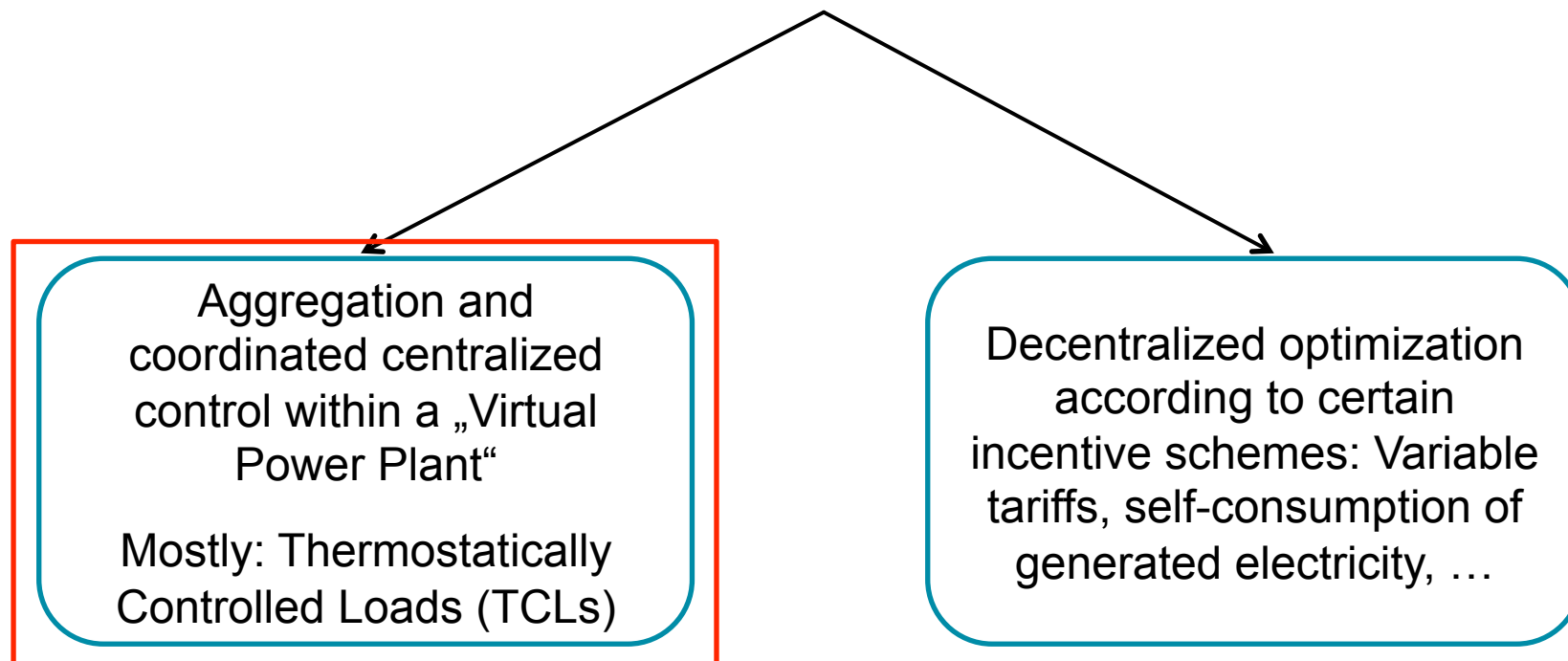
Possible Objectives:

- To shape the load curve in a desired way (e.g., peak shaving)
- To trade portions of energy on the market (day-ahead, intra-day)
- To balance prediction errors
- To deliver ancillary services from the demand side
- To protect the local grid infrastructure

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Coordination and Dispatch of TCLs

Project: Local Load Management (2008 – 2012)

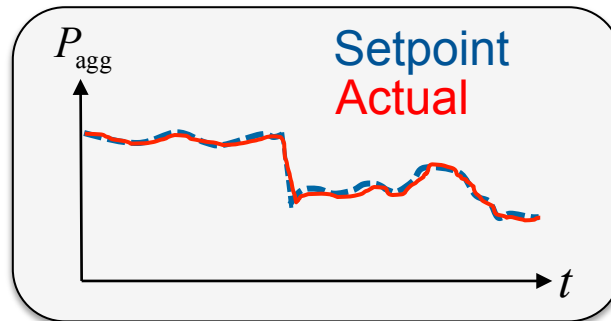
Partners: FHNW, Alpiq, Landis+Gyr

Researcher: Stephan Koch

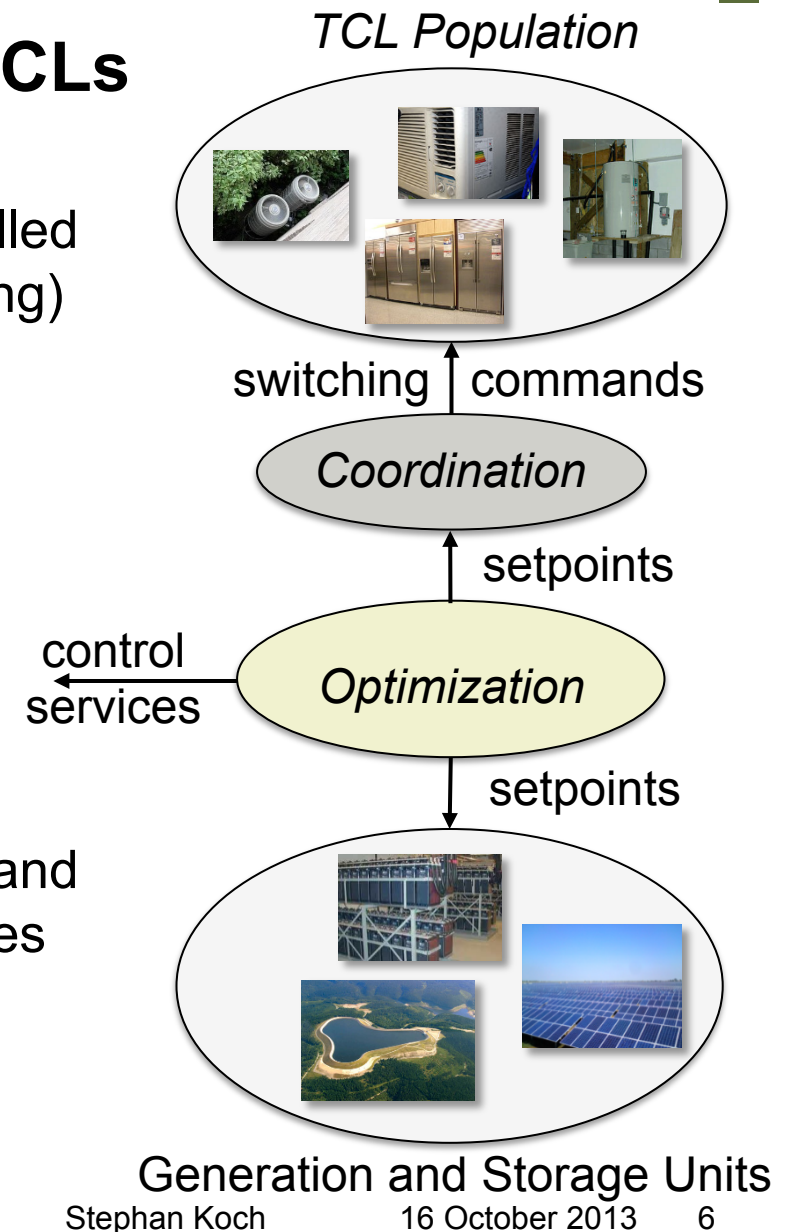
Funding: Swisselectric Research

Coordination and Dispatch of TCLs

- Make aggregated thermostatically controlled loads (TCLs) controllable (setpoint tracking)



- Integrate controllable loads with storage and generation units to provide control services

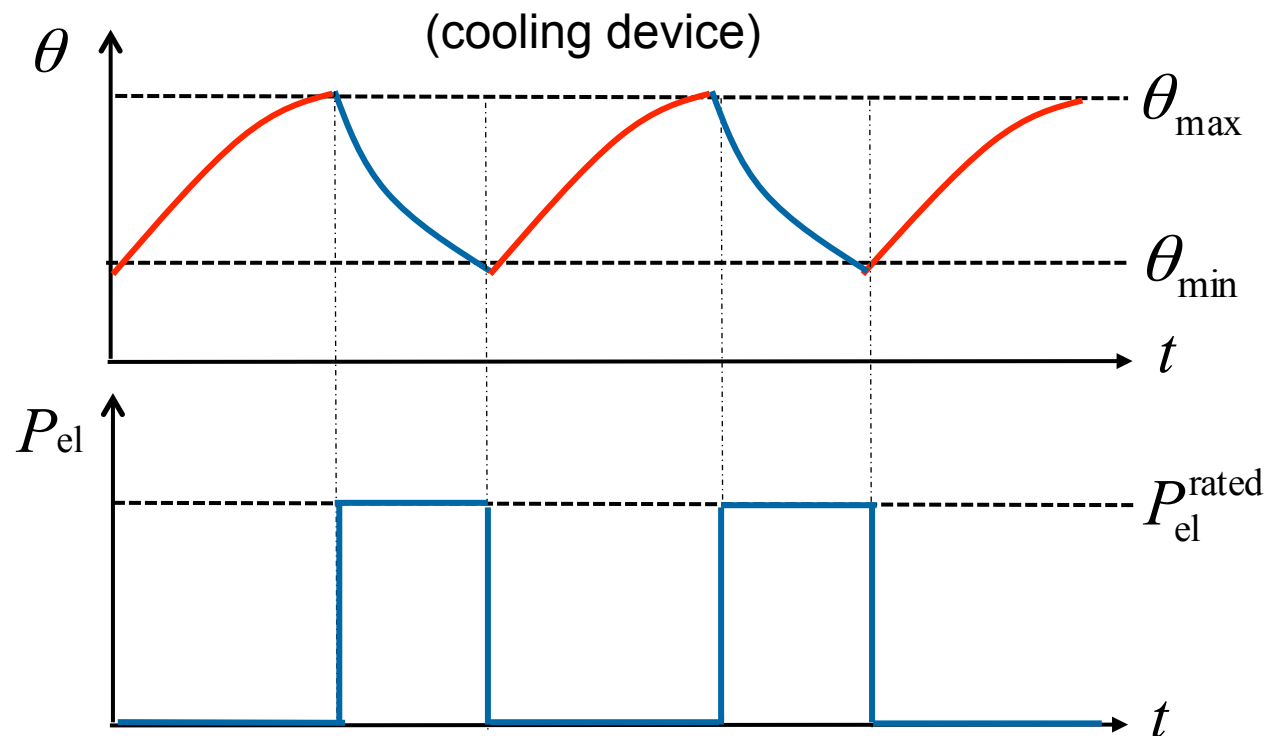


TCL Modeling Approach

- First-order differential equation for temperature dynamics of single TCL:

$$\theta_{i,t+1} = a_i \theta_{i,t} + (1 - a_i)(\theta_{a,i} - m_{i,t} \theta_{g,i}) + \omega_{i,t}$$

- Hysteretic thermostat controller produces characteristic cycling:

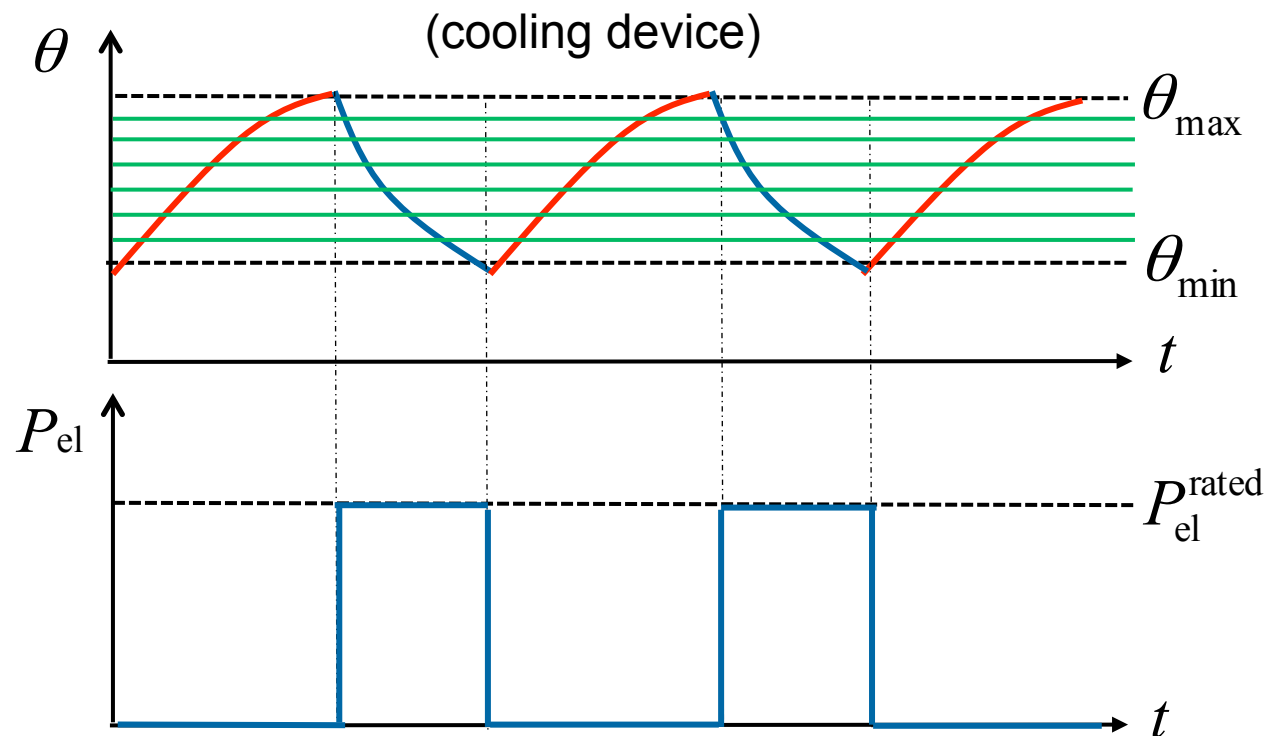


TCL Modeling Approach

- First-order differential equation for temperature dynamics of single TCL:

$$\theta_{i,t+1} = a_i \theta_{i,t} + (1 - a_i)(\theta_{a,i} - m_{i,t} \theta_{g,i}) + \omega_{i,t}$$

- Discretization of the temperature space into “bins”:

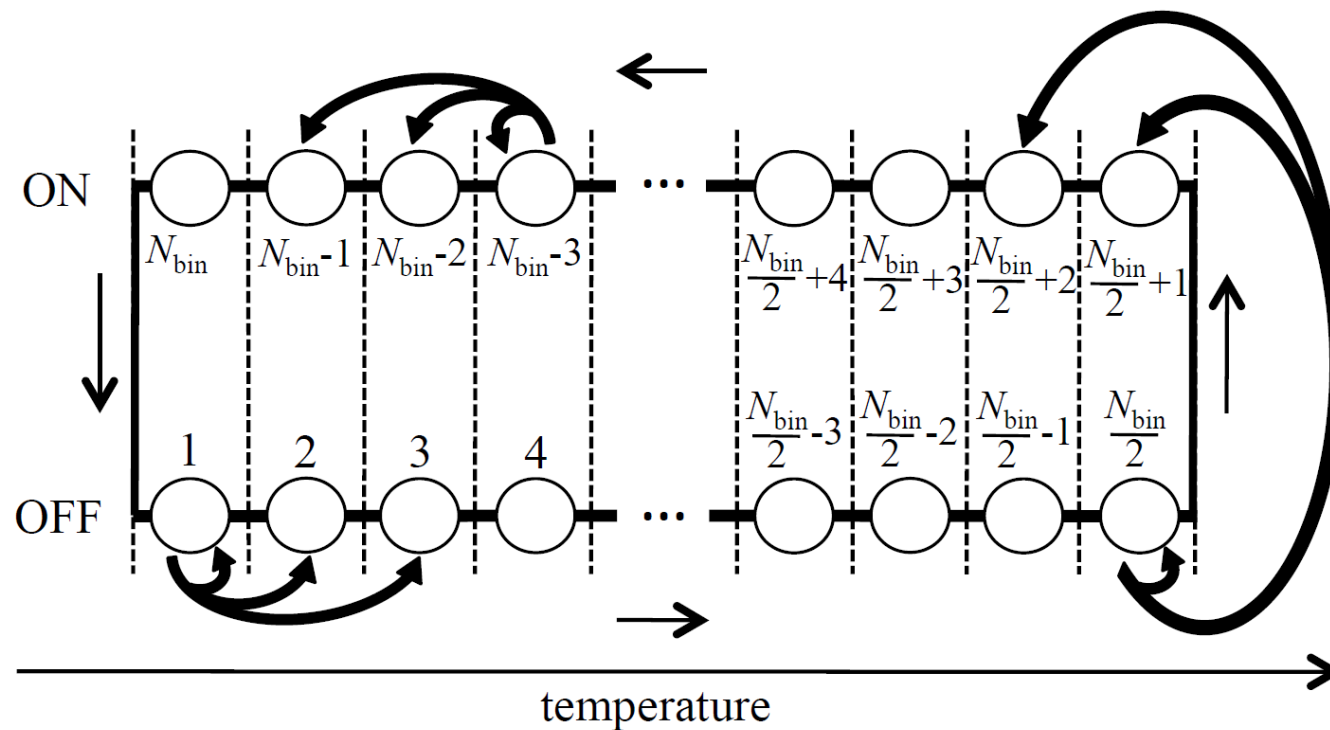


TCL Modeling Approach

- First-order differential equation for temperature dynamics of single TCL:

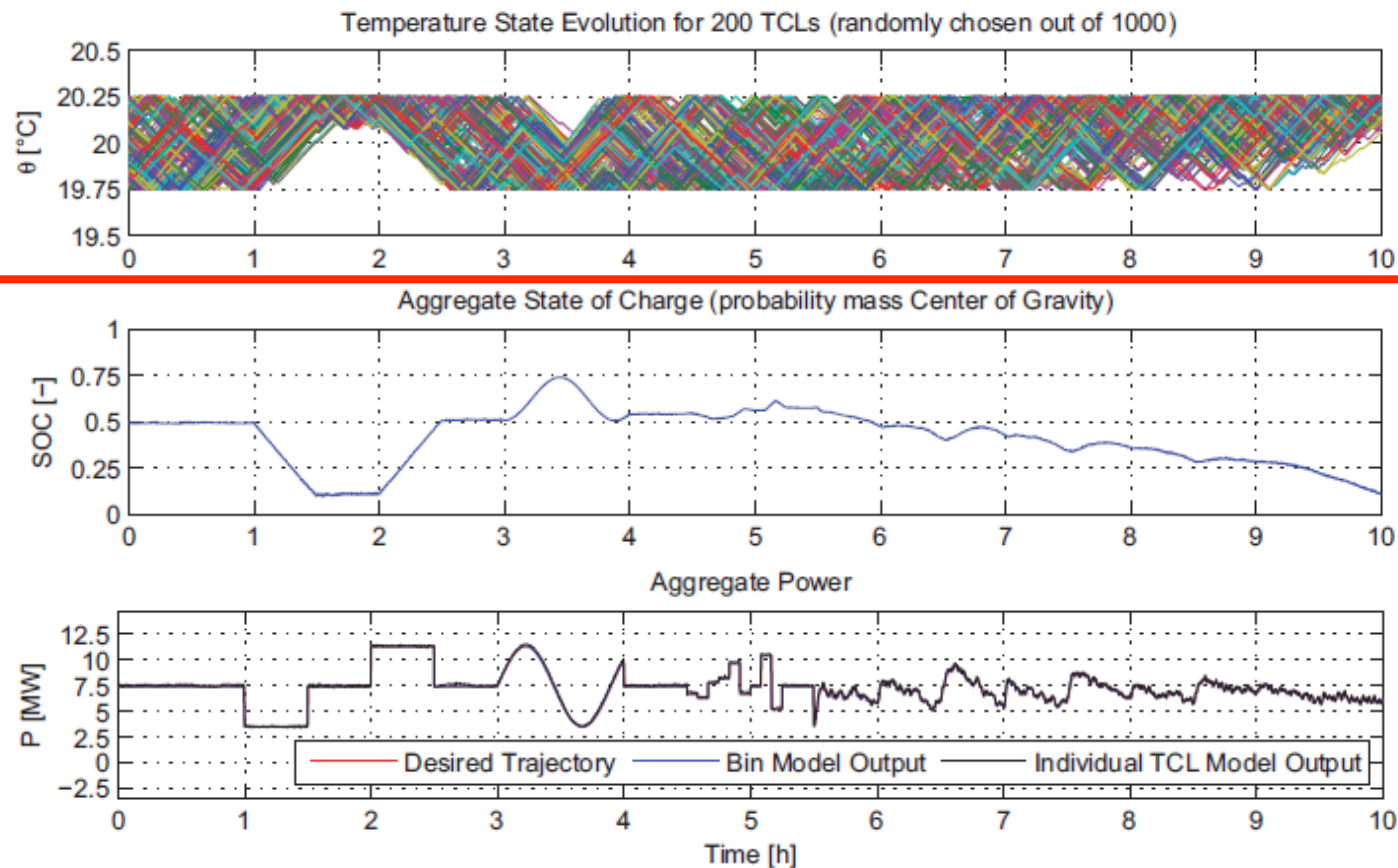
$$\theta_{i,t+1} = a_i \theta_{i,t} + (1 - a_i)(\theta_{a,i} - m_{i,t} \theta_{g,i}) + \omega_{i,t}$$

- Discretization of the temperature space into “bins”:



Control Strategy for TCL Setpoint Tracking

Simulation example: 1,000 air conditioning units



→ Storage characteristics

Integration into a Dispatch Framework

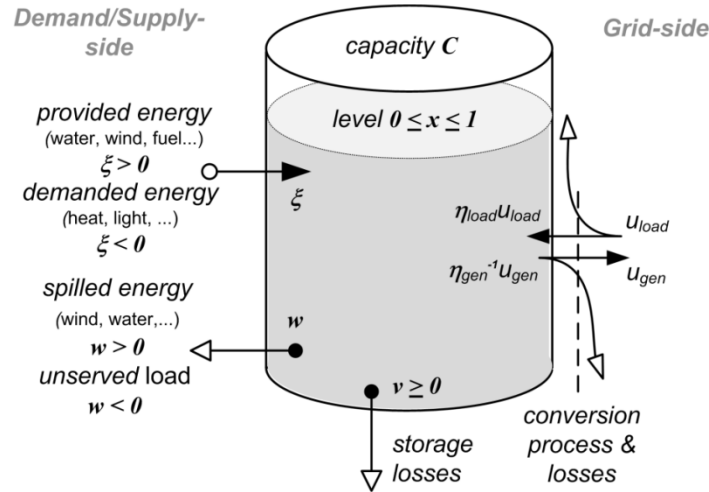
Coordinated Groups of TCLs can track a time-varying setpoint.

But where does the setpoint come from?

- Idea:
- Unify the modeling of load, storage, and generation units
 - Jointly optimize a flexible unit portfolio for a common goal

Integration into a Dispatch Framework

The Power Nodes
Modeling Framework



Storage capacity
×
state-of-charge

Power out-feed from grid

Power in-feed to grid

Shedding term

Internal losses

$$C\dot{x} = \eta_{load} u_{load} - \eta_{gen}^{-1} u_{gen} + \xi - w - v$$

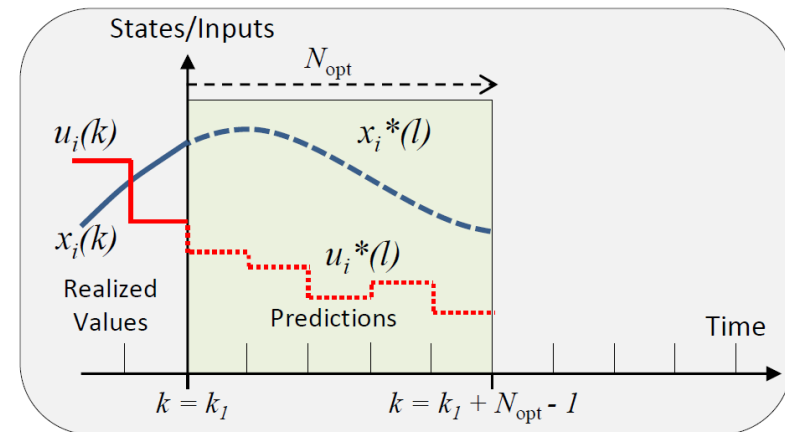
Efficiency factors

Provided / demanded power

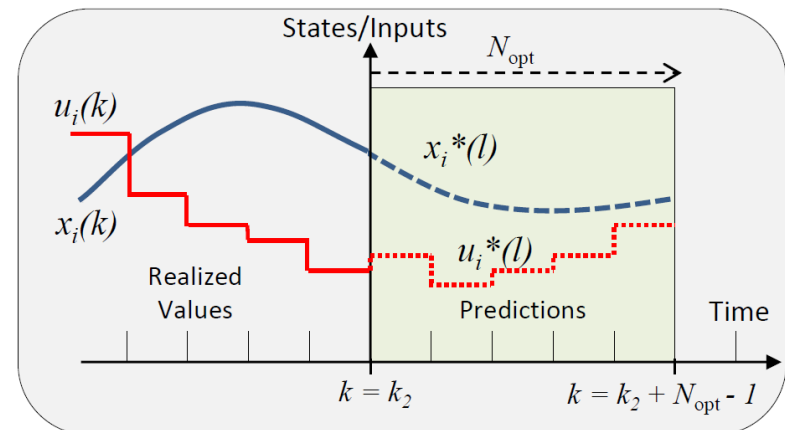
Integration into a Dispatch Framework

Dispatch with Model Predictive Control

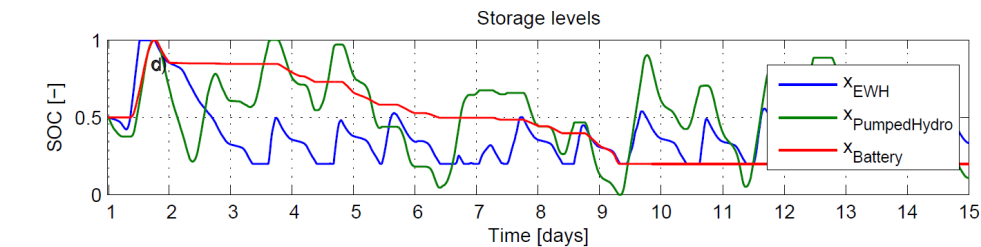
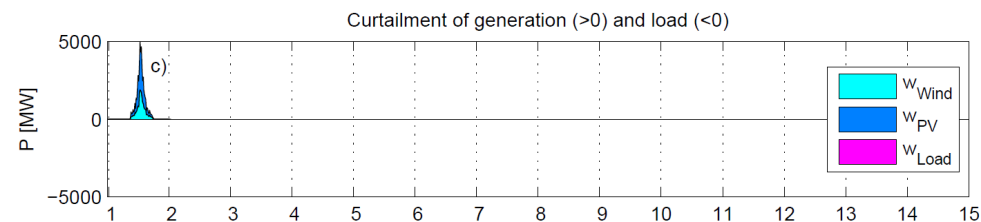
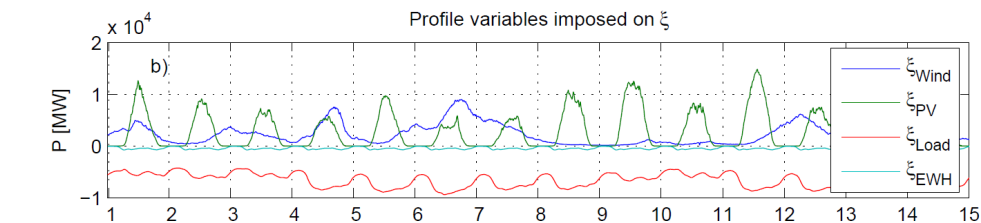
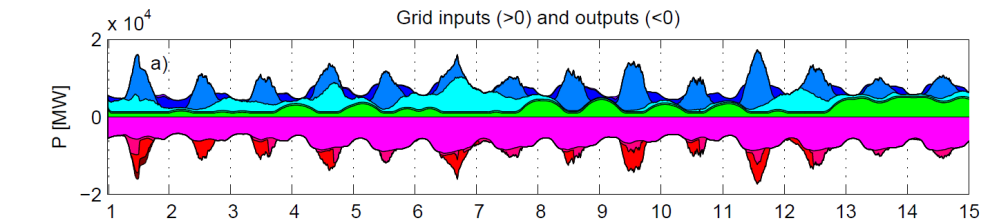
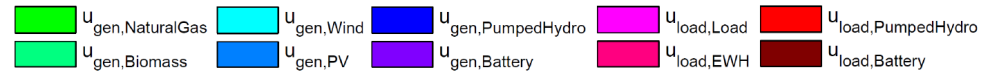
- Joint predictive optimization of a power node portfolio
- Cost function and constraint design allows to cover a variety of use cases:
 - Least-cost dispatch
 - Market-based VPP operation
 - Balancing of schedule deviations
 - Provision of frequency control reserves
 - Capacity firming of intermittent generation
 - Peak shaving
 - Residual load ramp-rate reduction



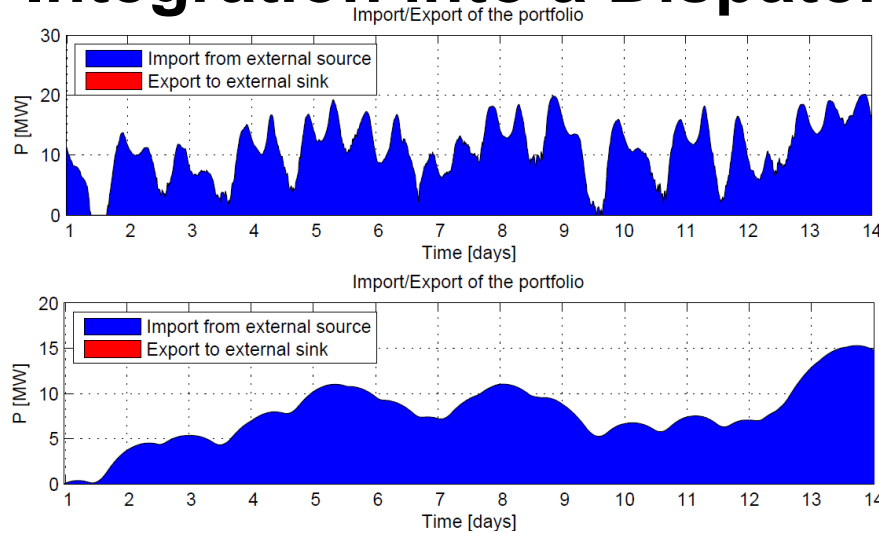
k_{opt} Optimization frequency: once every k_{opt} steps



Integration into a Dispatch Framework



Integration into a Dispatch Framework

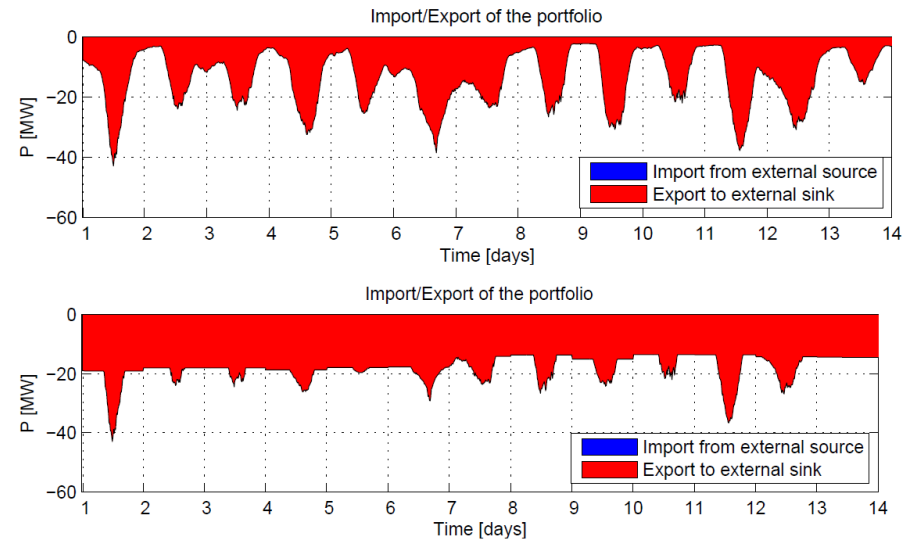


- In-feed of intermittent generation can attain low values
→ lack of reliably available capacity
- Increasing the minimum in-feed by dispatching flexible units accordingly:

$$J_k = \sum_{l=k}^{k+N_{opt}-1} J_{endo}^*(l) - \pi_{cap} \cdot \min_{l \in [k, k+N_{opt}-1]} u_{load}^{*slack}(l)$$

- Residual load can exhibit high ramps
→ high strain on conventional generation assets
- Smoothing via dispatch of flexible units:

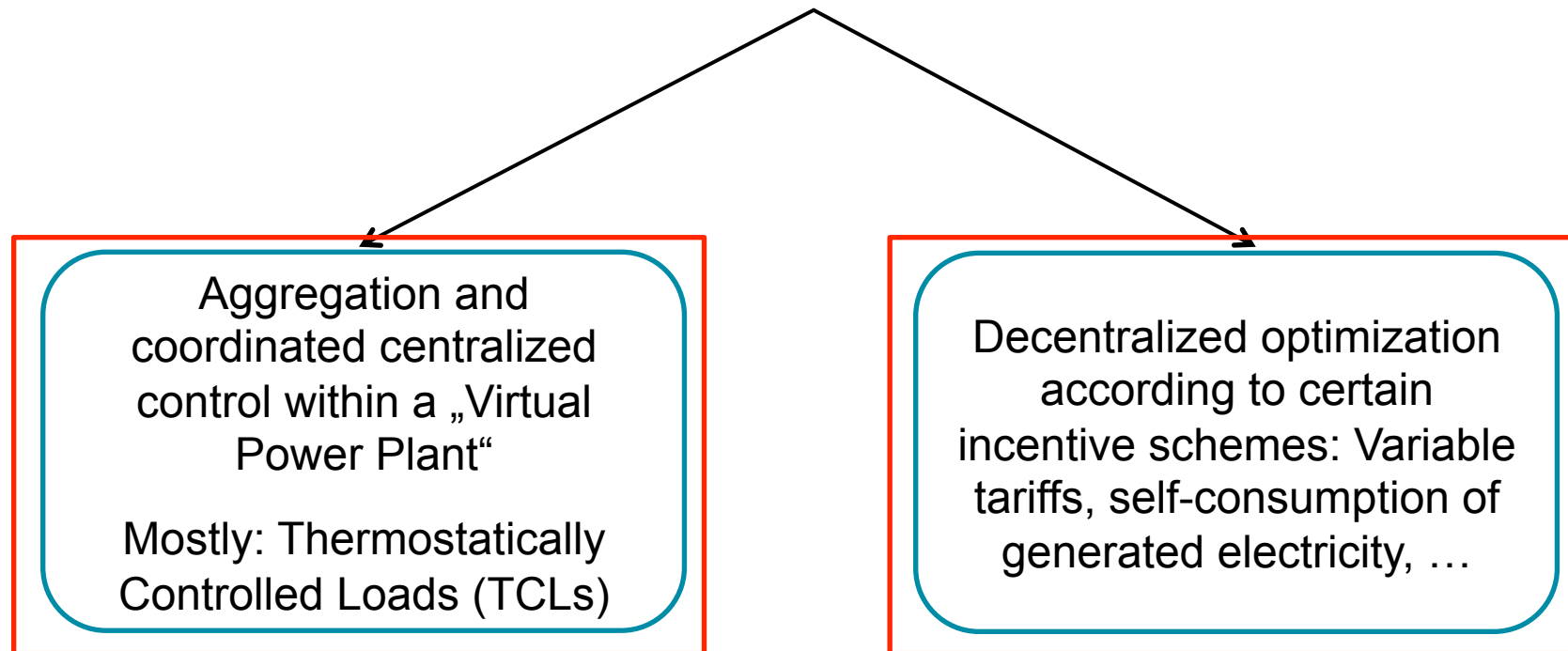
$$J_k = \sum_{l=k}^{k+N_{opt}-1} \pi_{ramp}^{slack} \frac{1}{t_s} (\delta u_{load}^{*slack}(l))^2 + \sum_{l=k}^{k+N_{opt}-1} J_{endo}^*(l)$$



General Aspects of Automated Demand Response

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PV Self-Consumption Optimization

Project: SmartGrid-Polysun – Design Tool for Local Load Management
Funding: Federal Office of Energy, Swisselectric Research

Researcher:

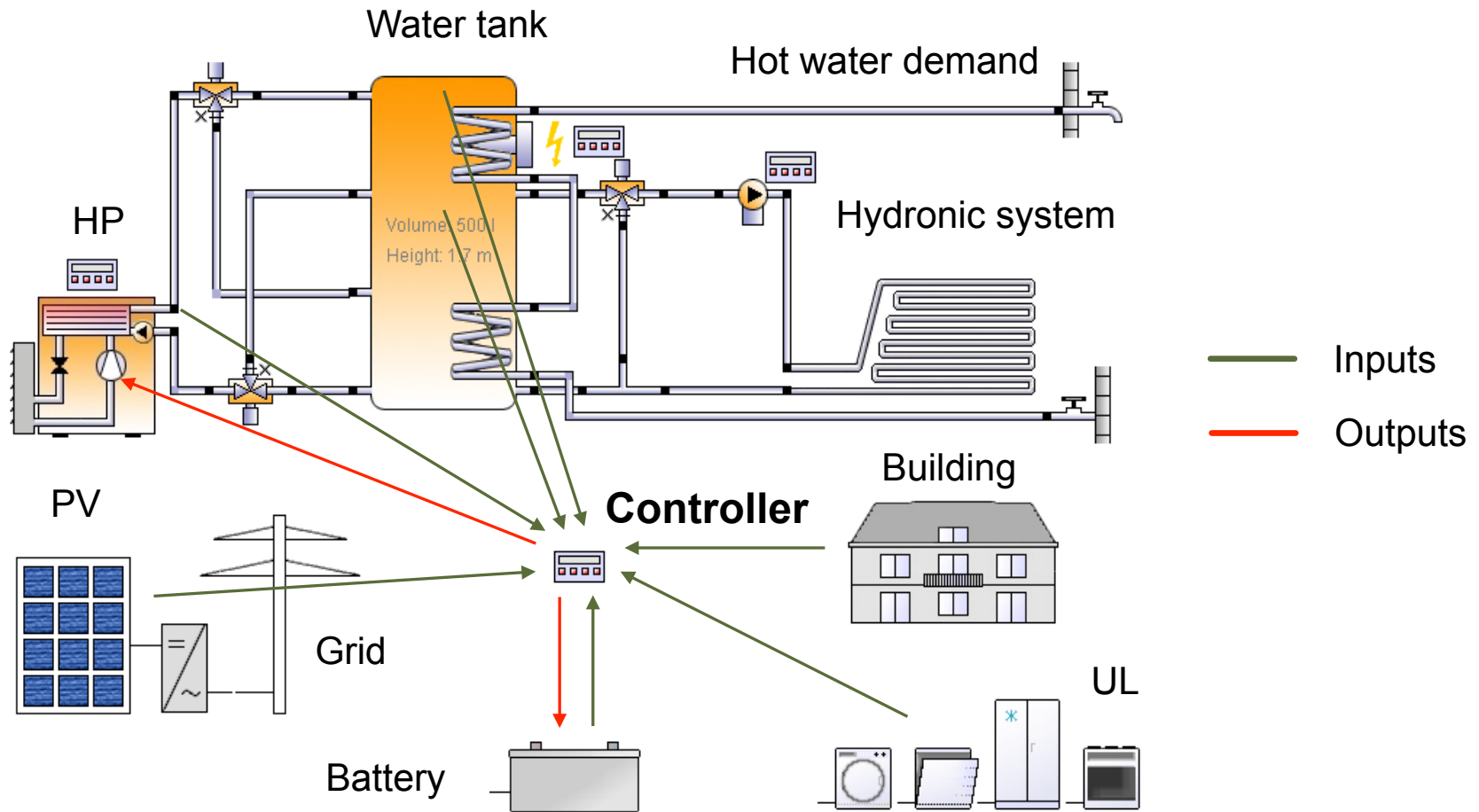


Evangelos Vrettos
PhD student
Power Systems Laboratory
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Project Partner:



Modeling of a Residential Building

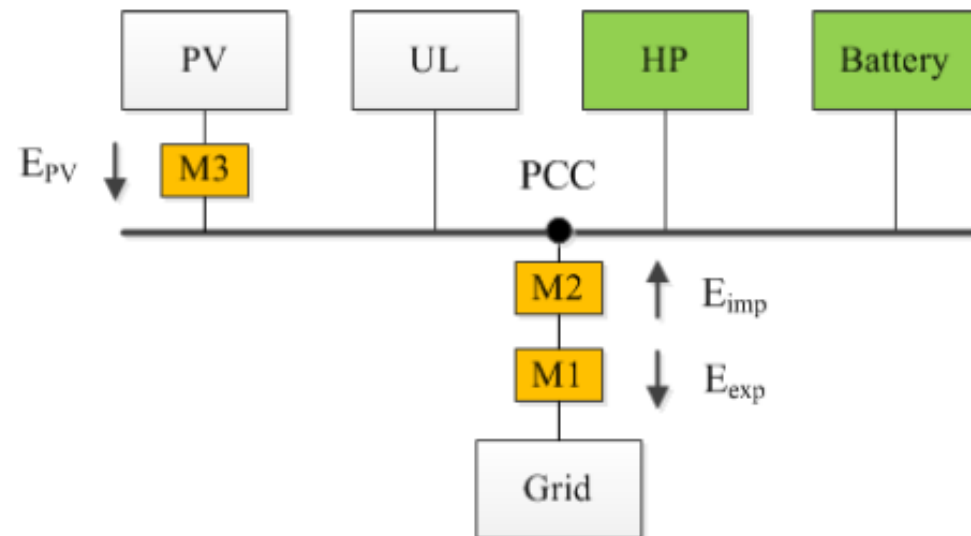


PV Self-Consumption Problem Setup

- Goals: (a) shift HP operation towards hours with high PV production
(b) charge battery when PV surpluses exist
(c) discharge battery to cover load in evening / at night
- Definitions: PV self-consumption ratio (ξ), total electricity cost (E_{cost})

$$\xi = \frac{E_{\text{PV} \rightarrow \text{bld}}}{E_{\text{PV}}} = \frac{E_{\text{PV}} - E_{\text{exp}}}{E_{\text{PV}}}$$

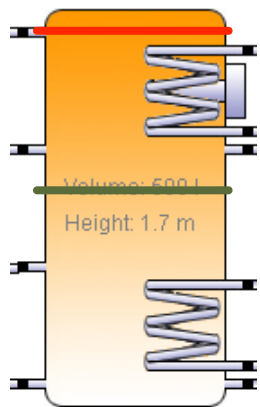
$$E_{\text{cost}} = C_{\text{FIT}} \cdot E_{\text{exp}} - C_{\text{el}} \cdot E_{\text{imp}}$$



PV Self-Consumption Optimization

Base case (A0)

- HP operates based on internal thermostat
- No battery


 $T_{s10,max}$
 $T_{s7,min}$

Algorithm (A1)

- Smart HP operation
- No battery

Algorithm (A2)

- HP operates based on internal thermostat
- Battery is present

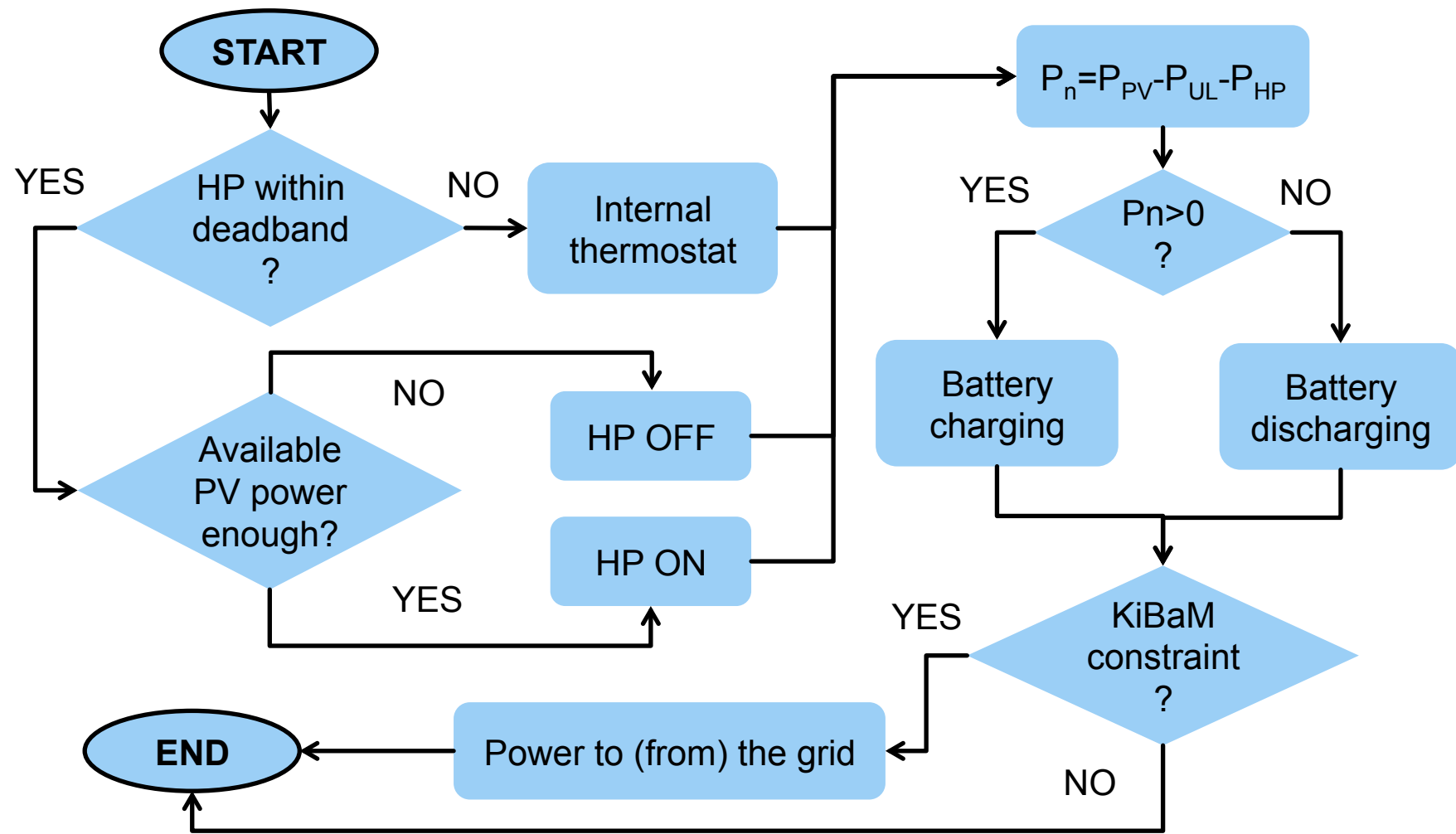
Algorithm (A3)

- Smart HP operation
- Battery is present
- **Priority to HP**

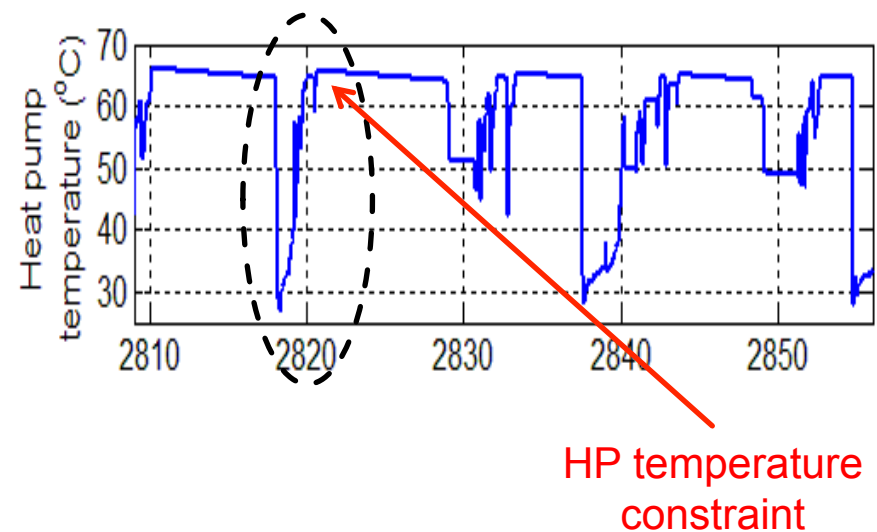
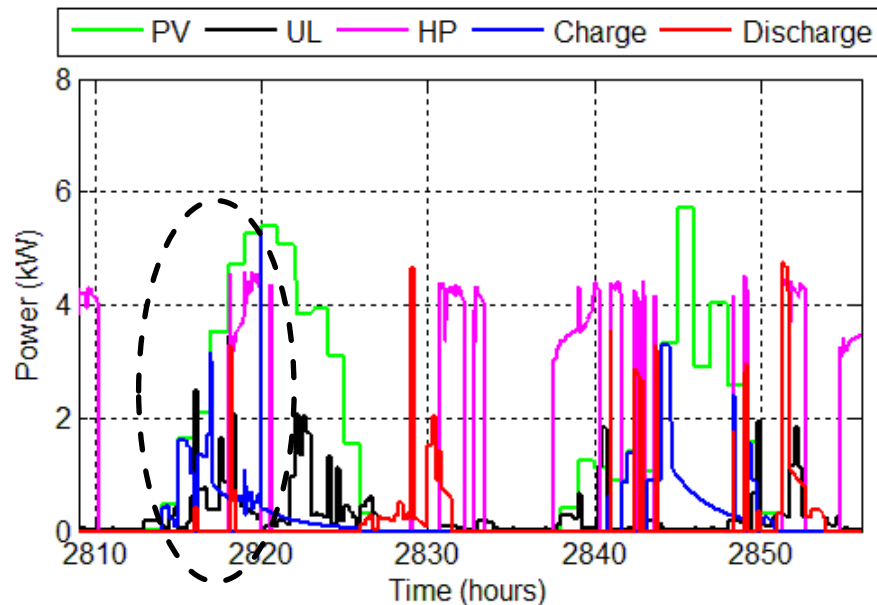
Algorithm (A4)

- Smart HP operation
- Battery is present
- **Priority to battery**

PV Self-Consumption Optimization: Algorithm A3



Simulation Results



- **Highest potential:** in spring when both PV energy and heat demand are significant
- **Charging priority:** when $P_{PV} > P_{HP} + P_{UL}$, the HP turns on
- **HP limited potential:** after a few hours the HP gets overheated and turns off
- **Barrier for HP:** temperature constraints of HP, rather than building thermal inertia

Simulation Results

Yearly Simulation

Algorithm	PV self-consumption (%)	Electricity bill (€)
Base case (A0)	19.71	1248.63
smart HP (A1)	21.14	1249.24
Battery (A2)	35.19	1169.32
Both (A3)	36.46	1162.98
Both (A4)	35.49	1163.36

- **HP only:** limited potential for PV self-consumption, cost can increase due to losses
- **Battery only:** (a) high potential for PV self-consumption, and savings of ~79 € per year
(b) with current electricity tariffs, pays off only if battery cost < 60 €/kWh
- **Both:** priority to HP achieves highest self-consumption and savings of ~85 € per year
- HP saves additionally only ~6 € per year, but at virtually zero investment cost

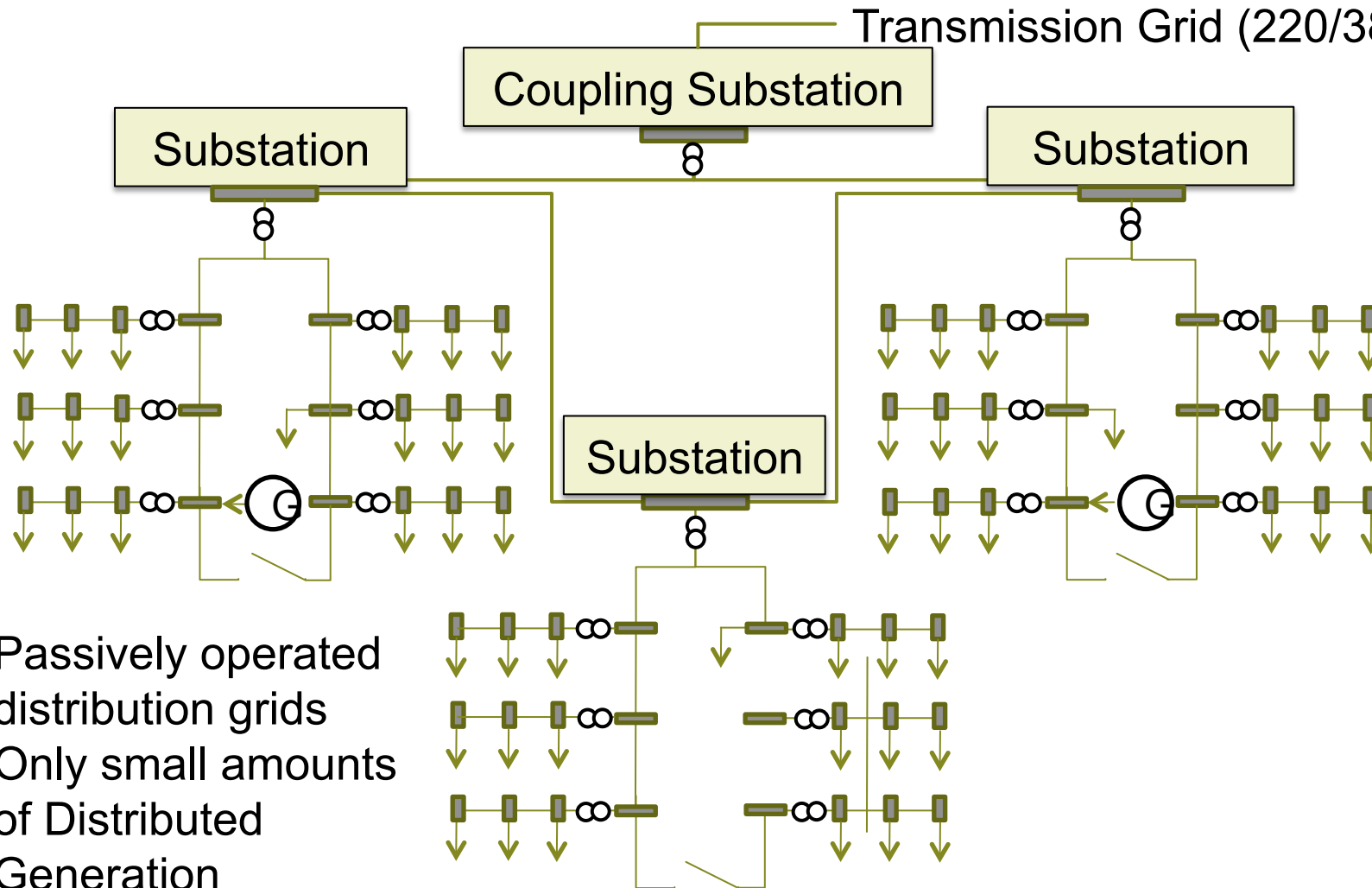
Development of a *Smart Distribution Grid* Simulator

Project: “Pioneer Fellowship“ at ETH Zurich (PostDoc Employment)

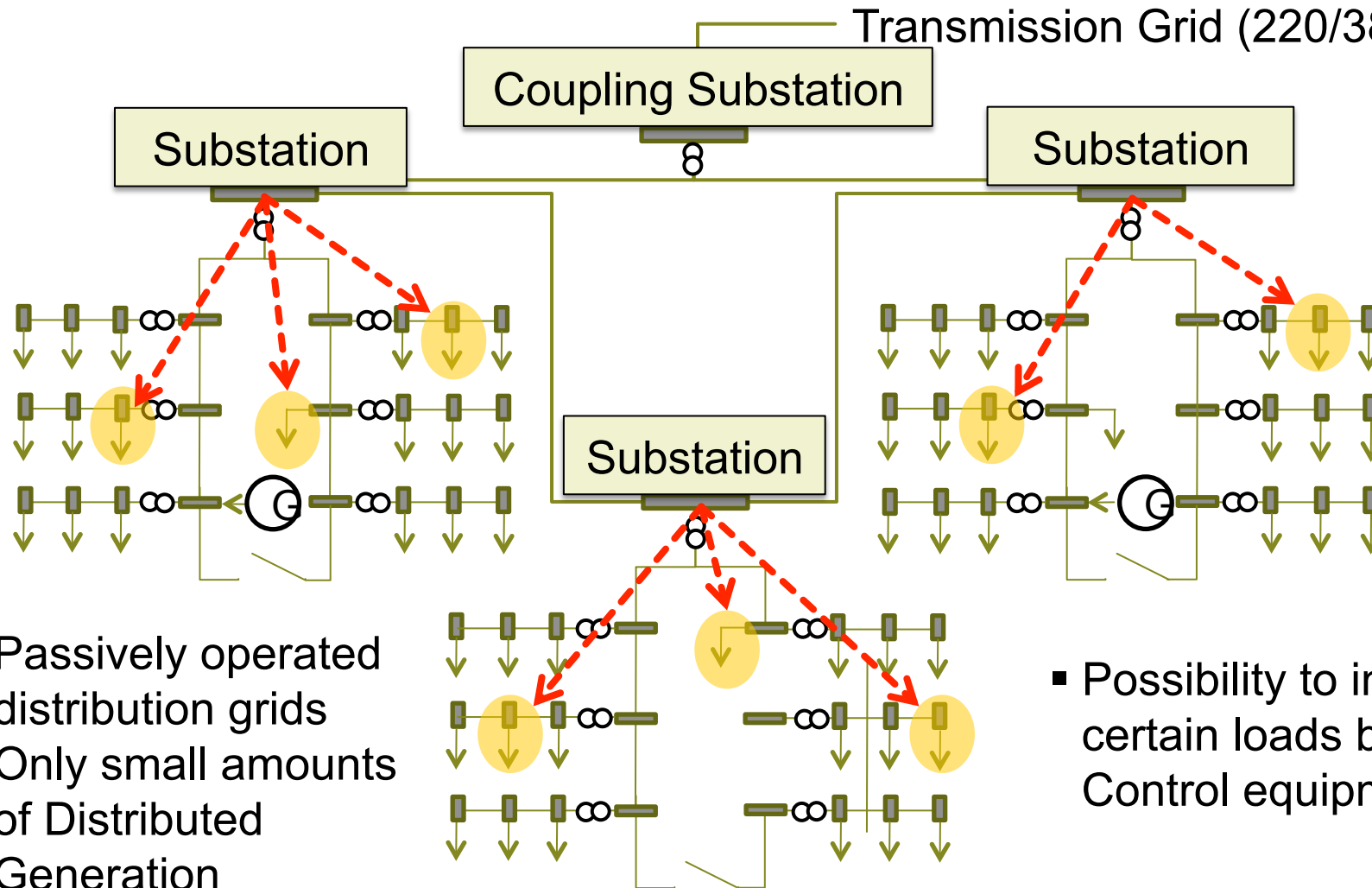
Researcher: Stephan Koch

Funding: ETH Zurich

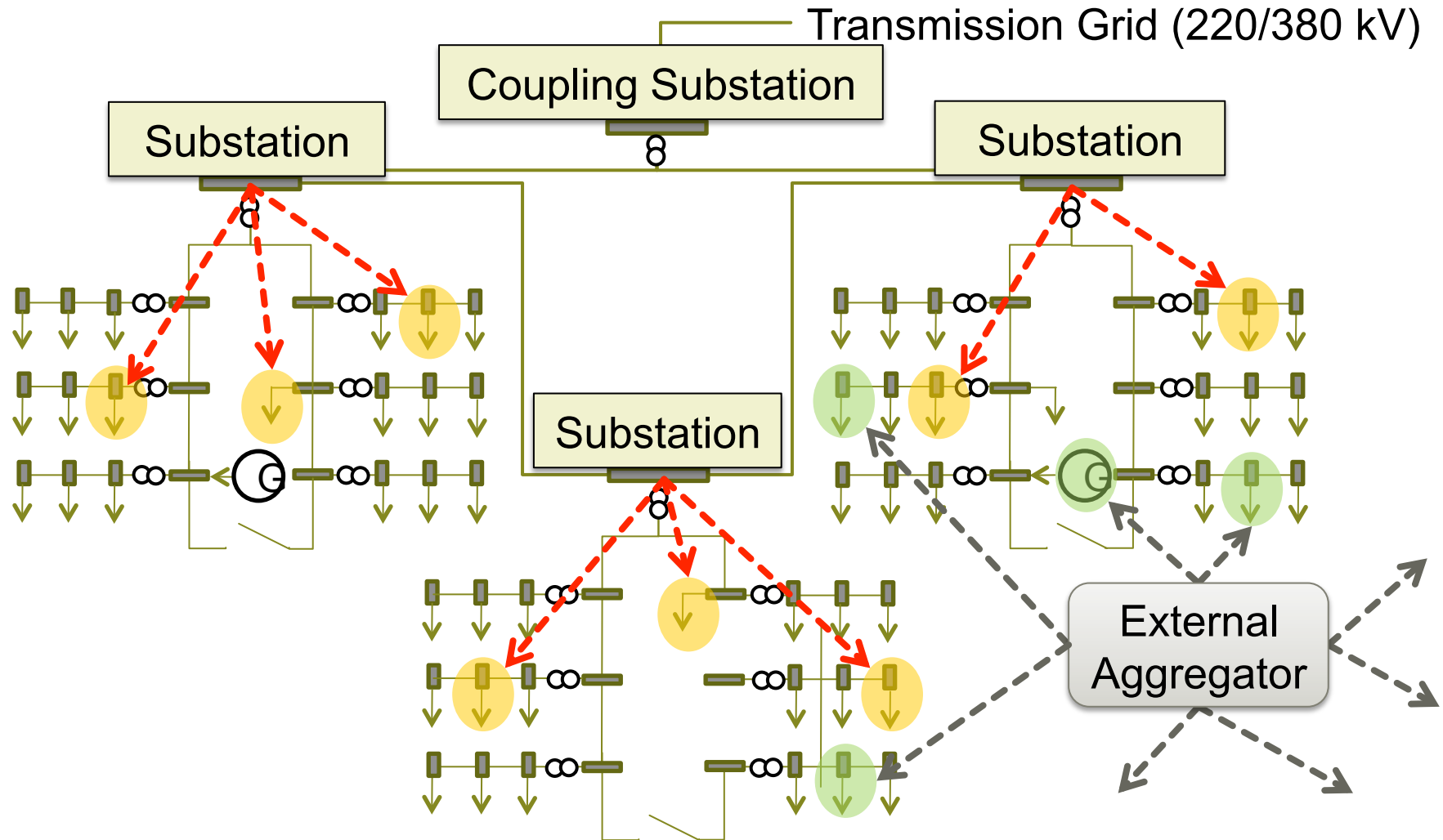
The Challenge: Distribution Grid Operation (present)



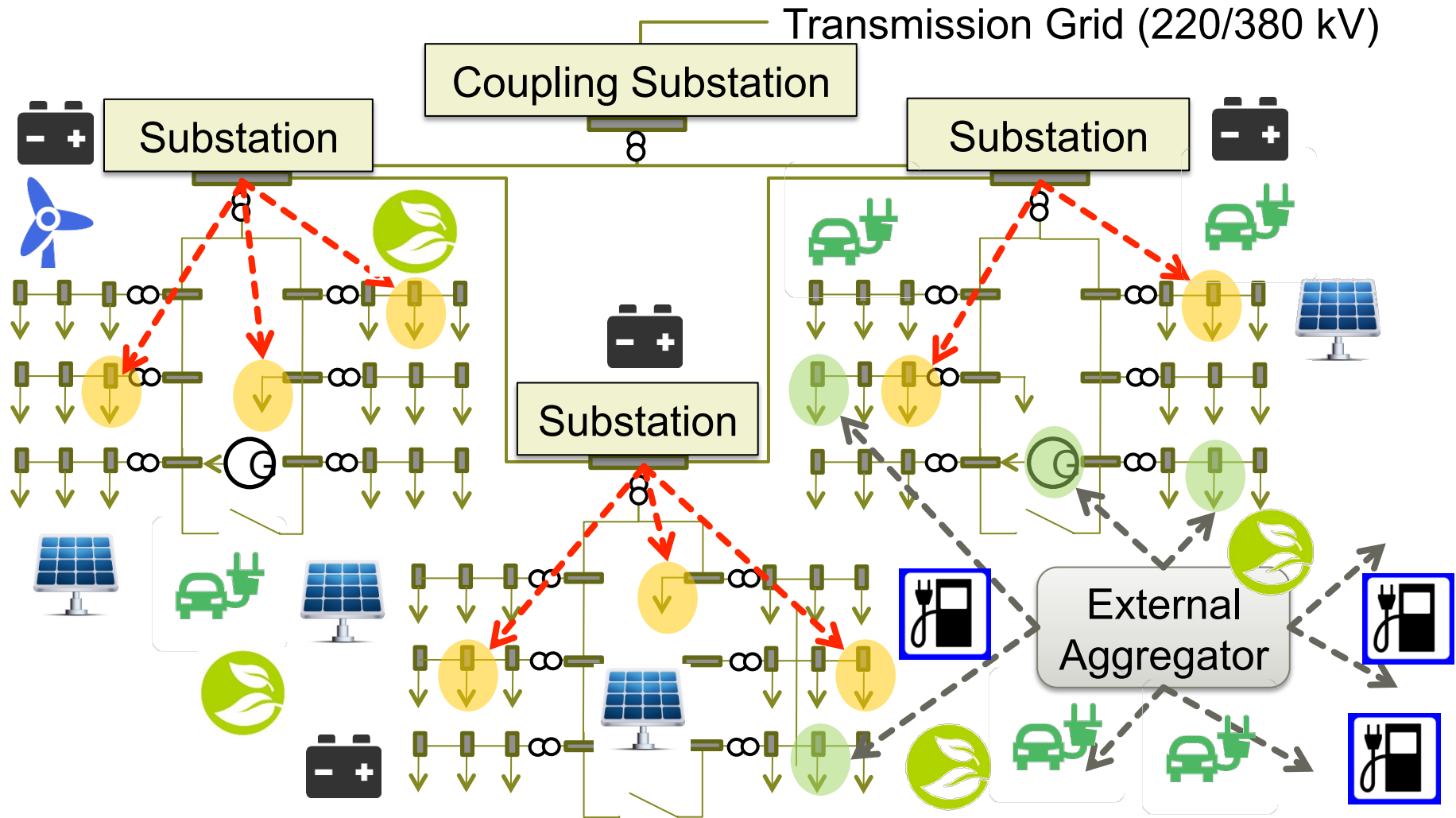
The Challenge: Distribution Grid Operation (present)



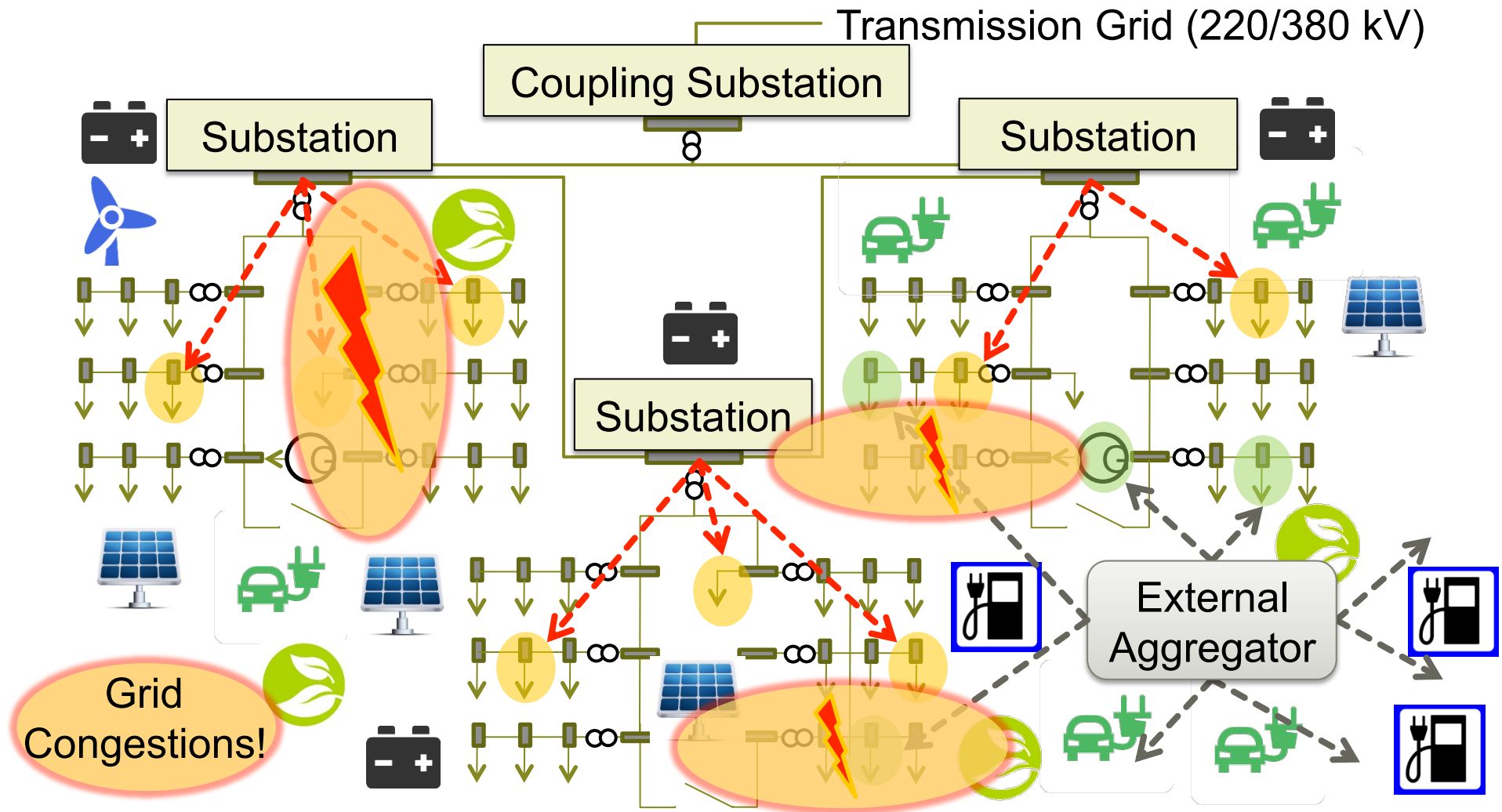
The Challenge: Distribution Grid Operation (future)



The Challenge: Distribution Grid Operation (future)



The Challenge: Distribution Grid Operation (future)



Distribution Grid Challenges

- Keeping all voltages within permissible limits in all operational situations
- Avoiding component overloadings
- Being aware of the current grid condition and taking correct countermeasures in case of disturbances



Investments in even more copper and steel?

Distribution Grid Challenges

- Keeping all voltages within permissible limits in all operational situations
- Avoiding component overloadings
- Being aware of the current grid condition and taking correct countermeasures in case of disturbances



The electricity grids need to become *smarter*.



This requires novel software tools.

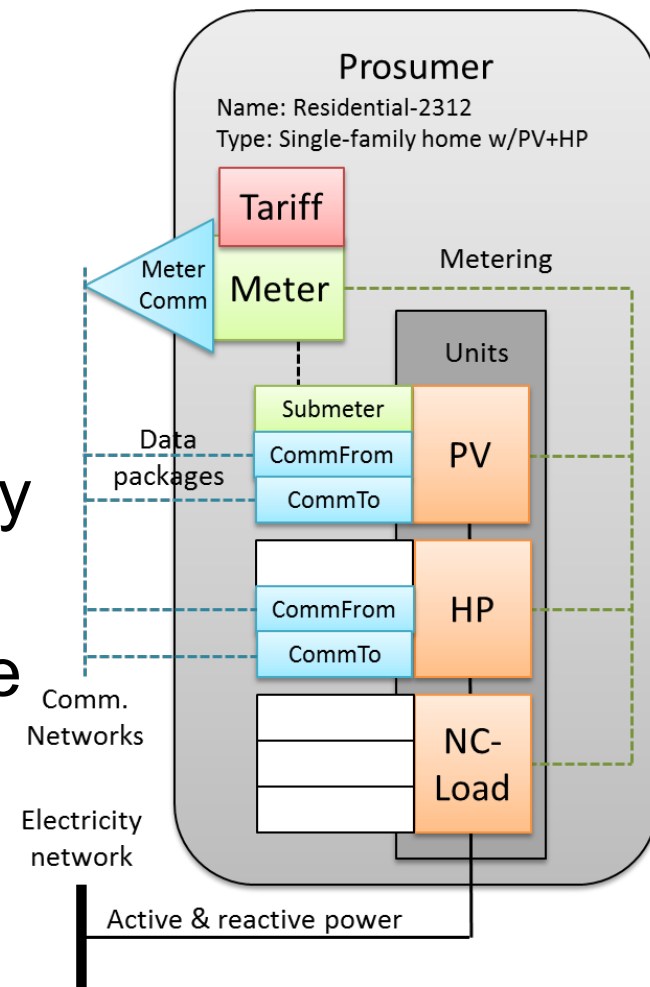
The Solution Approach

- Simulation of a *smart* distribution grid including the representation of the new physical reality in the grid
- Novel analysis and planning methods
- Innovative operation and control strategies

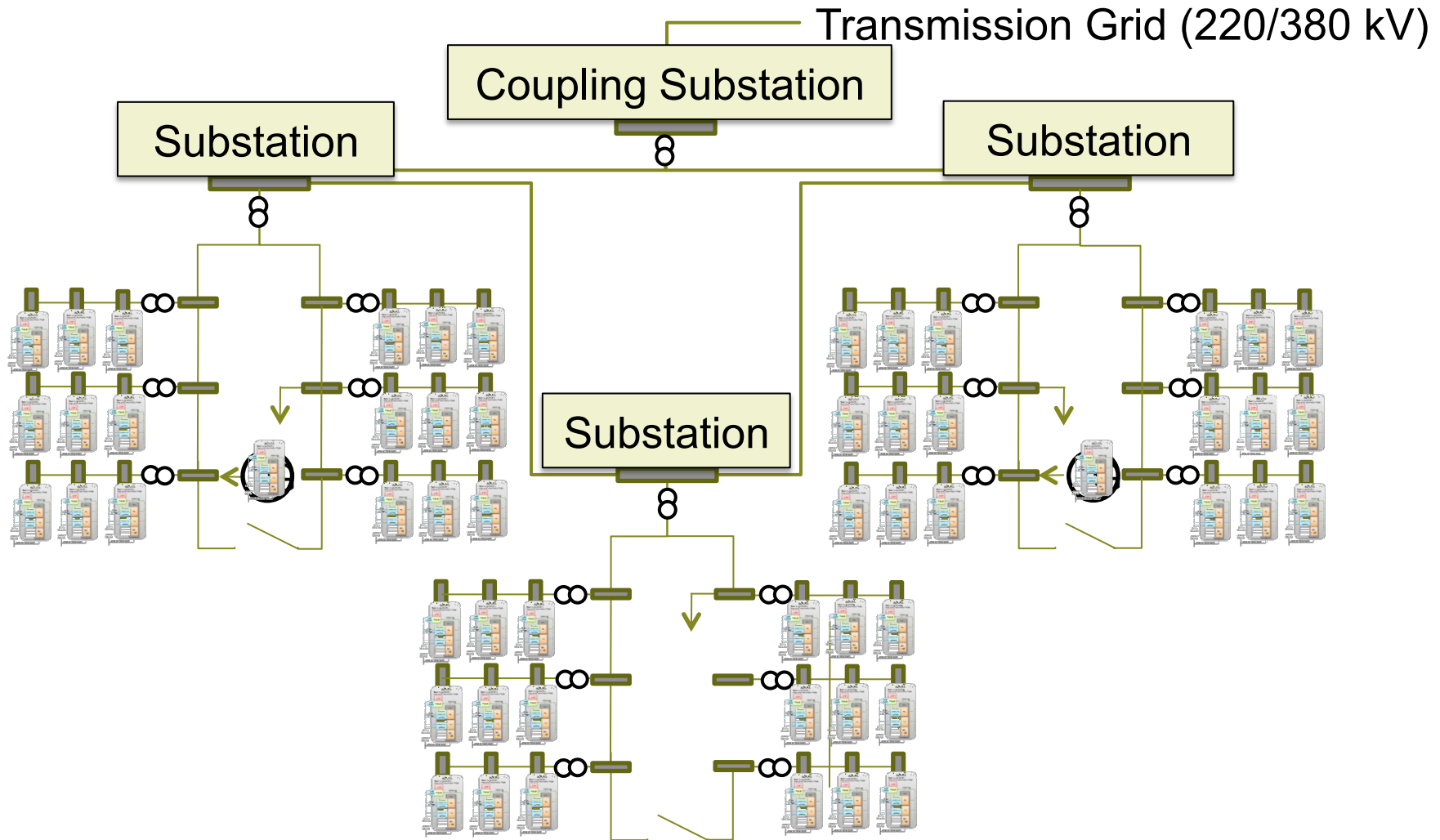
The Software DPG.sim – Basic Data Model

New Paradigm: Active Prosumers

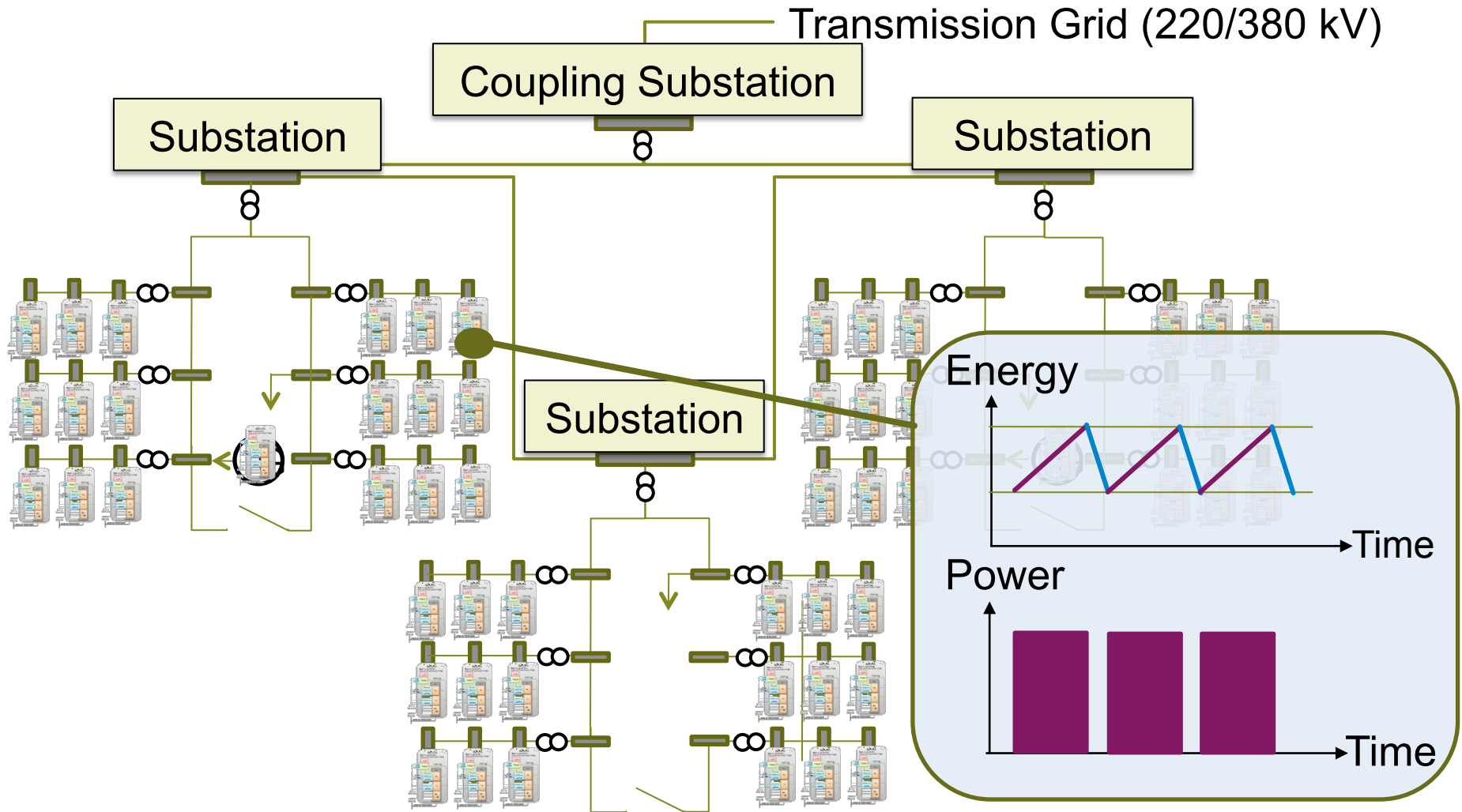
- Consumers can be producers at the same time → **Prosumers**
- Prosumers can locally optimize their interactions with the electricity grid
- External aggregators can bring the pooled capacity of decentralized units to the market



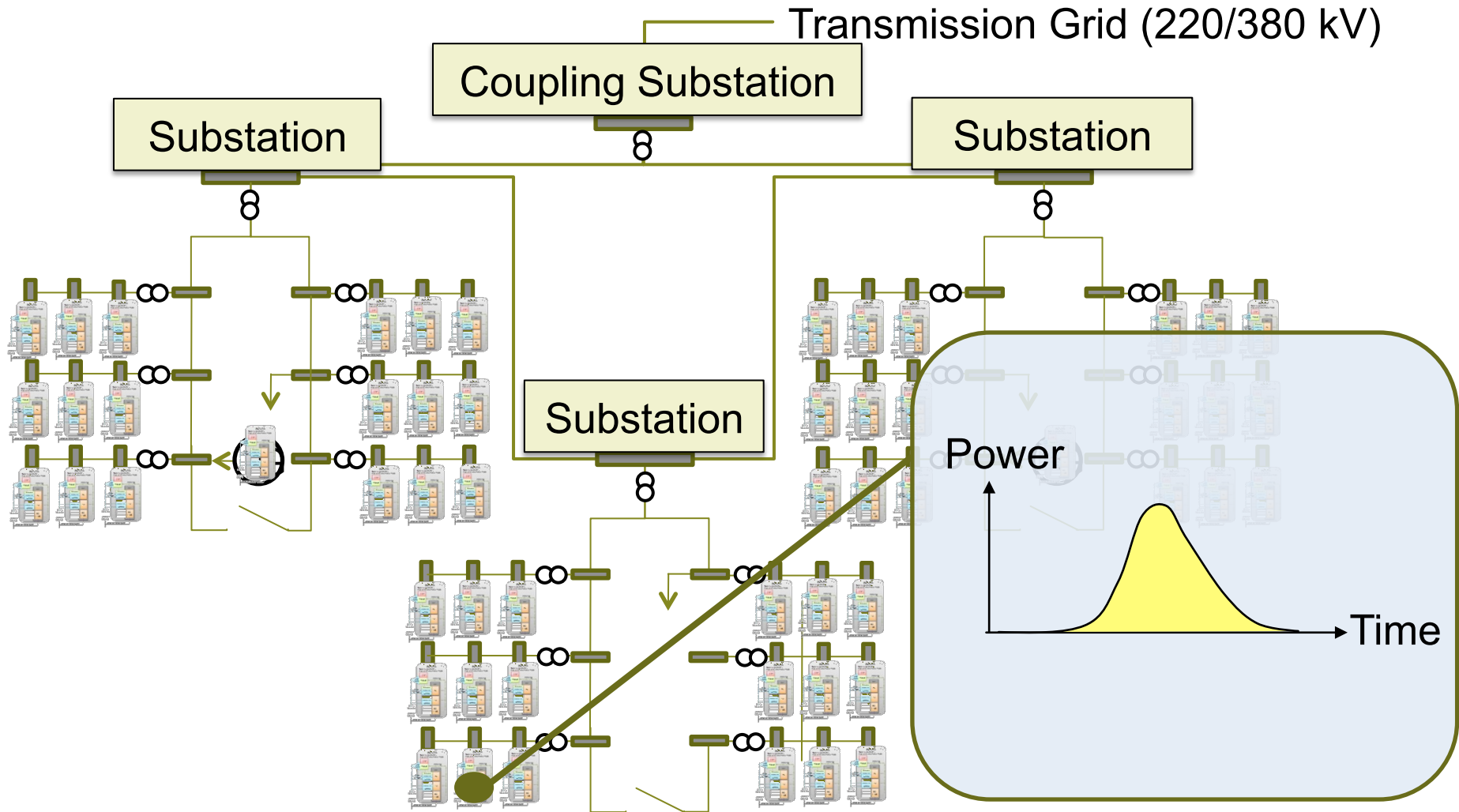
DPG.sim: Prosumer Dispersion on Grid Topology



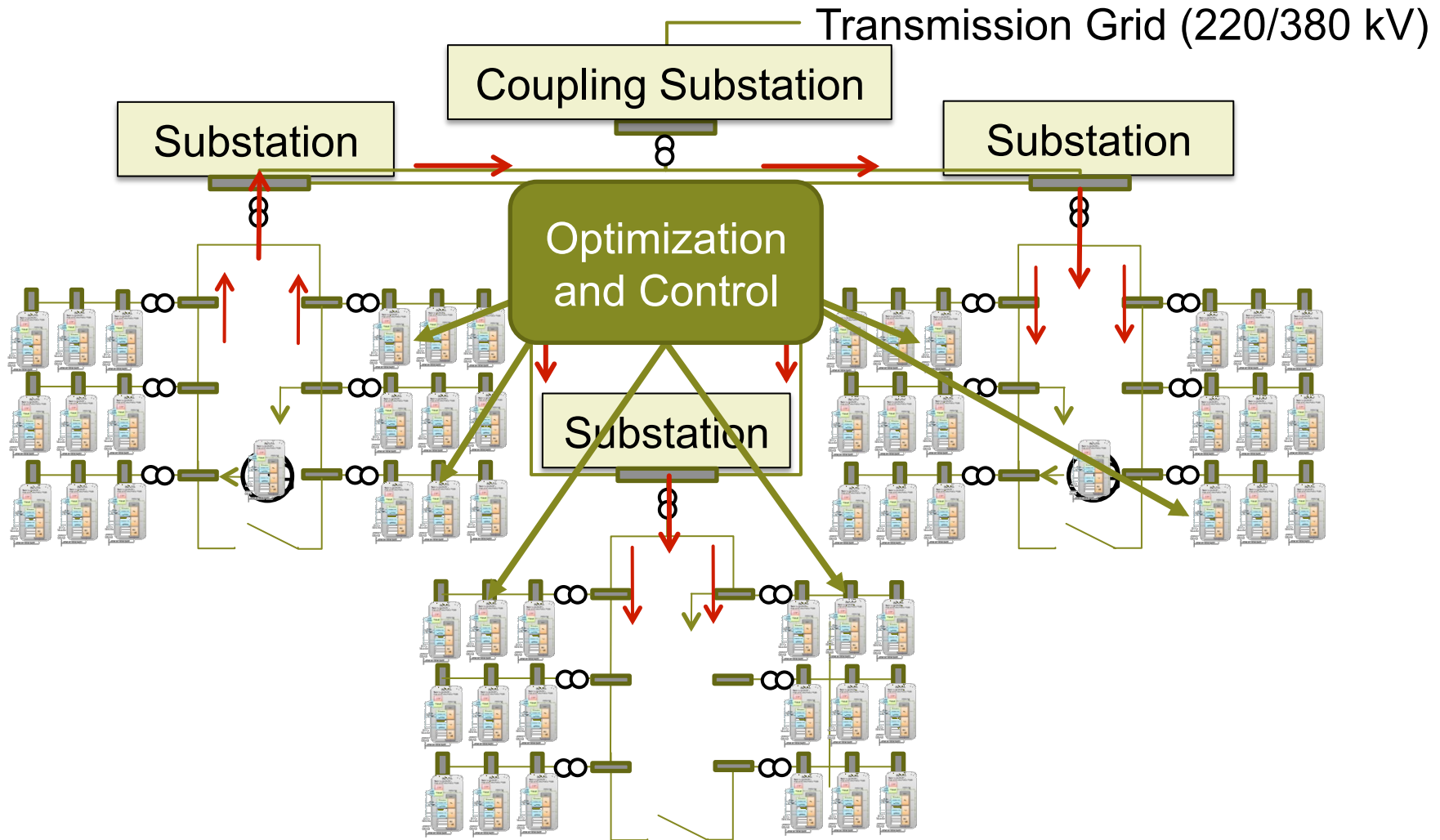
DPG.sim: Simulation of Individual Units



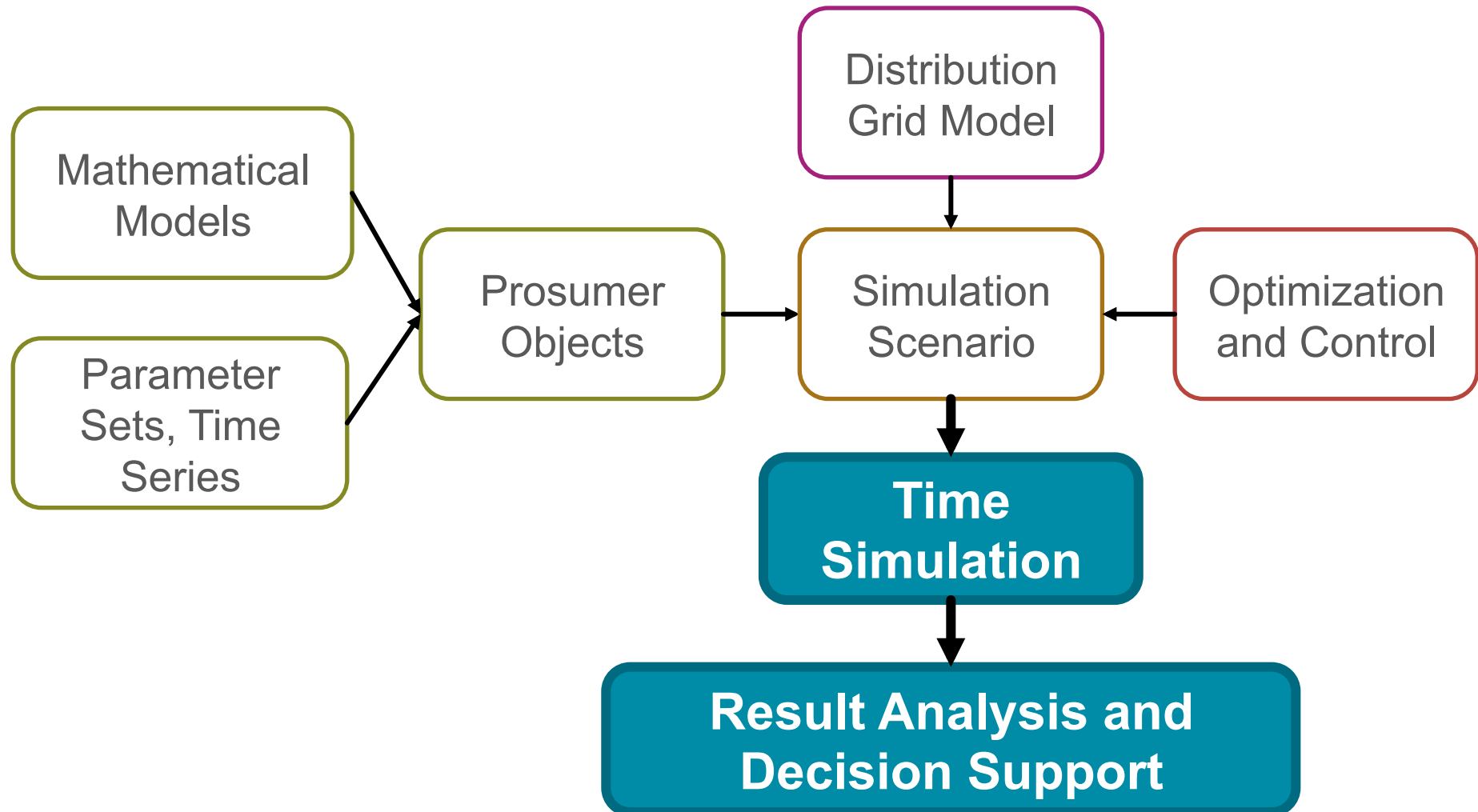
DPG.sim: Simulation of Individual Units



DPG.sim: Optimization and Control



DPG.sim: Structure of the Simulation Tool

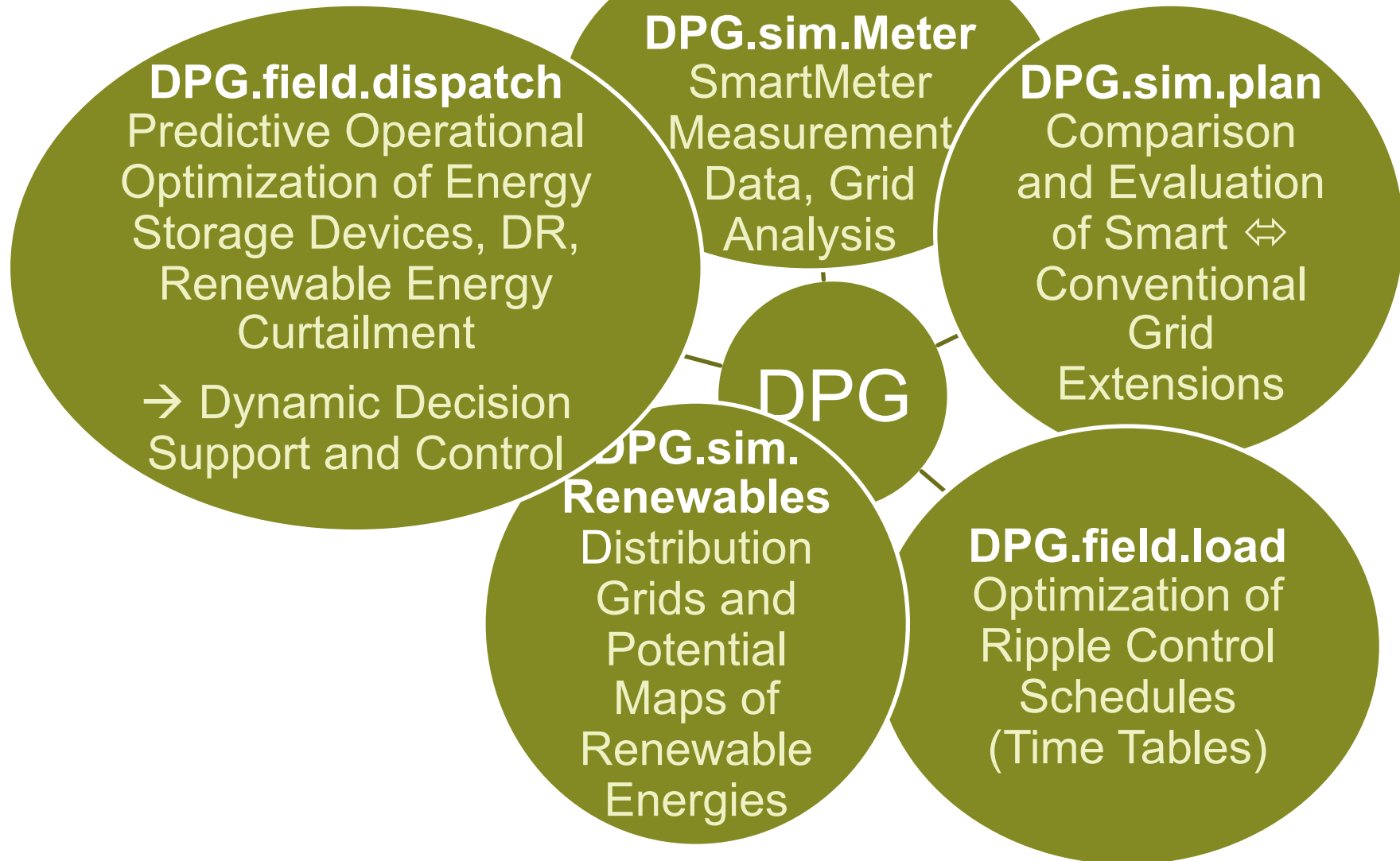


DPG.sim: Exemplary Screenshots

The image displays two screenshots of the DPG.sim software interface. The left screenshot shows the 'Edit Unit' configuration screen for a 'Generic Photovoltaic Unit'. It includes sections for General (Name: Generic PV unit, Unit category: generation), Unit Parameters (Symbol, Name, Unit table), Model Equations (Model Dynamics, Input Matrix B, Feedthrough Matrix D), Internal Controller (Controlled output, Manipulated input, Control Algorithm), and External interface (Electrical characteristic, Measureable from outside, Controllable from outside, Type of influence). The right screenshot shows the 'Simulation Status' screen, featuring a network diagram, simulation parameters (Simulation Status, Speed (key) 50, Current simulation time, Start time, End time), and four time-series plots: Total Load, Total Generation, Residual Load, and a plot for Base Admittance, Current Voltage, Voltage Angle, and Net Injection P/Q.

functional prototype

DPG Toolboxes and Extensions



Conclusions

Aggregating DR approaches represent an important factor for system flexibility in a power system with high renewable energy shares.

Local optimization and control on the customers' premises can increase self-consumption of generated power and alter the aggregate customers' behavior.

Both aggregating and local control approaches have an impact on distribution grids that needs to be properly managed.

Thank you for your attention!
Questions?

