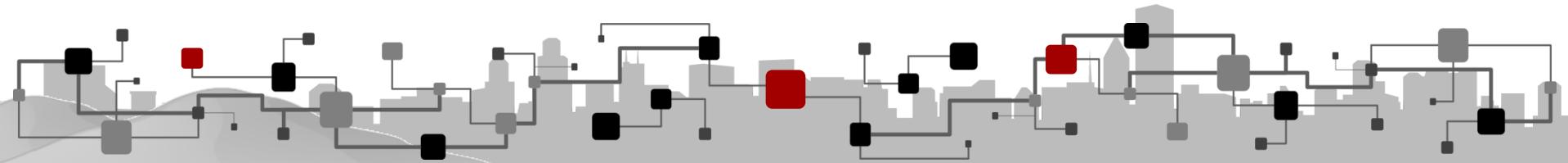


Der nächste intelligente Schritt im Gebäudebereich

29. Energie-Lunch: Künstliche Intelligenz im Gebäude, 22. April 2021

Philipp Heer

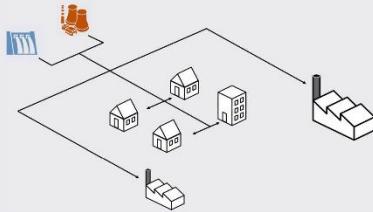
Deputy Head Urban Energy Systems Laboratory, Empa



Motivation

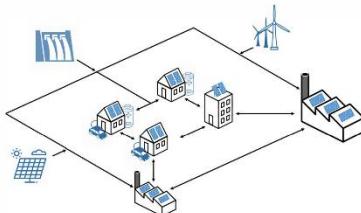
STATUS QUO

CENTRAL GENERATION



2050

DECENTRAL GENERATION



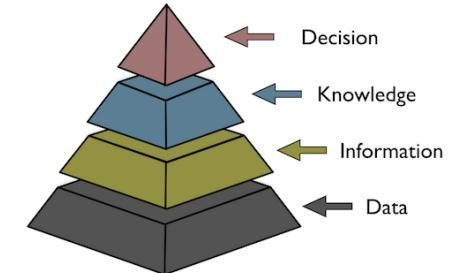
STATUS QUO

DOMINATING FOSSILE ENERGY CARRIERS

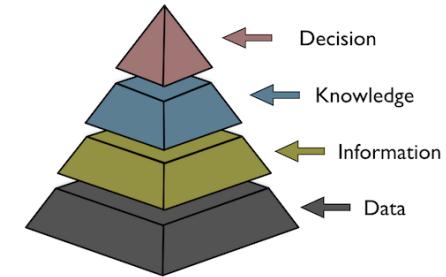
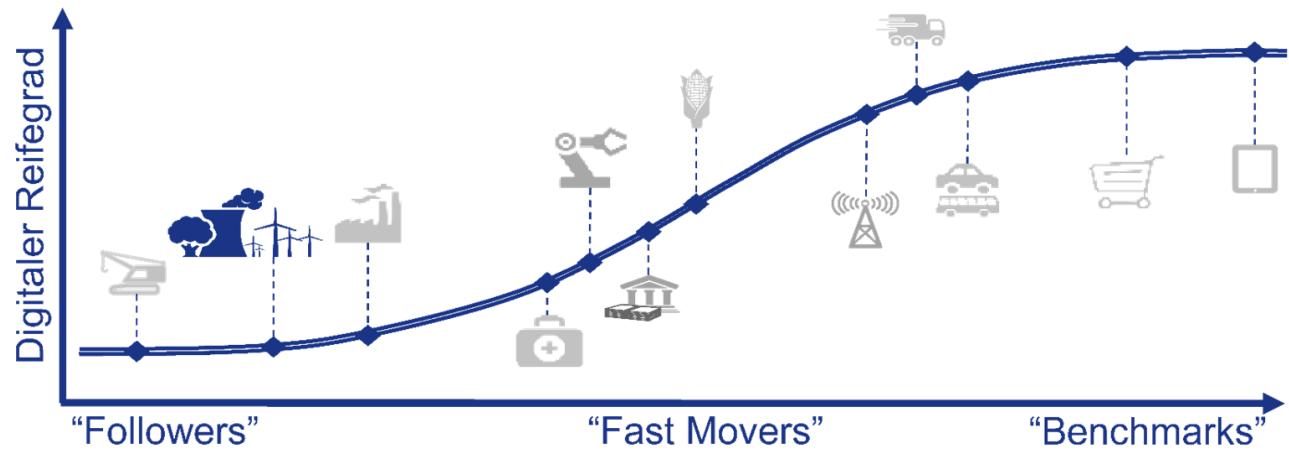


2050

DOMINATING ELECTRICITY AS ENERGY CARRIER

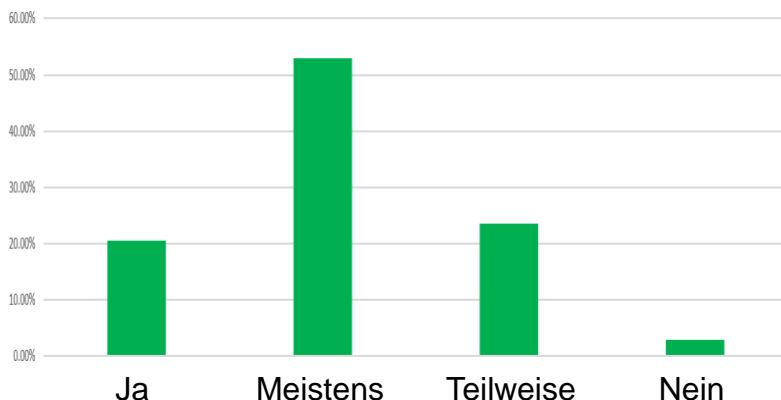


Digital Maturity Assessment

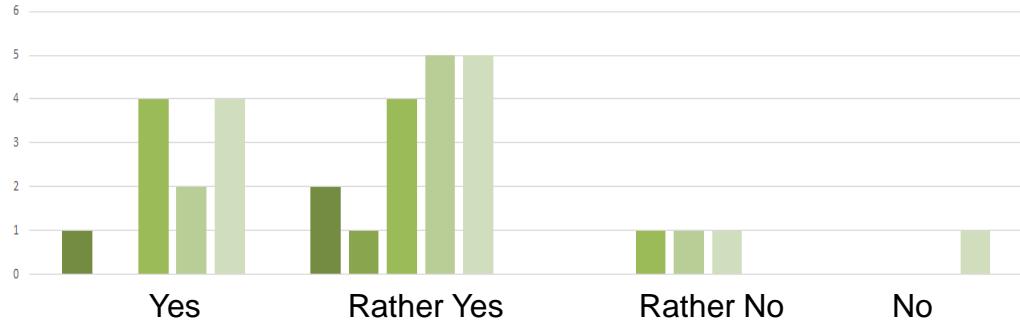


Umfrage bei Gebäudeeigentümern (~9'500 Gebäude)

Funktionieren ihre technischen Installationen gemäss ihren Erwartungen?



Denken sie ihre Messdaten könnten besser genutzt werden?



- Full interconnectivity. Predictive control. EE Class A+
- Interconnected fields. Premium room automation. EE Class A
- Partially interconnected fields. Basic room automation. EE Class B
- Individual fields. Basic room automation. EE Class C
- Individual fields. No Room automation. EE Class D

Was sind Smarte Gebäude?

Klassischer Ansatz: Gebäudeautomation bedient drei Kategorien:

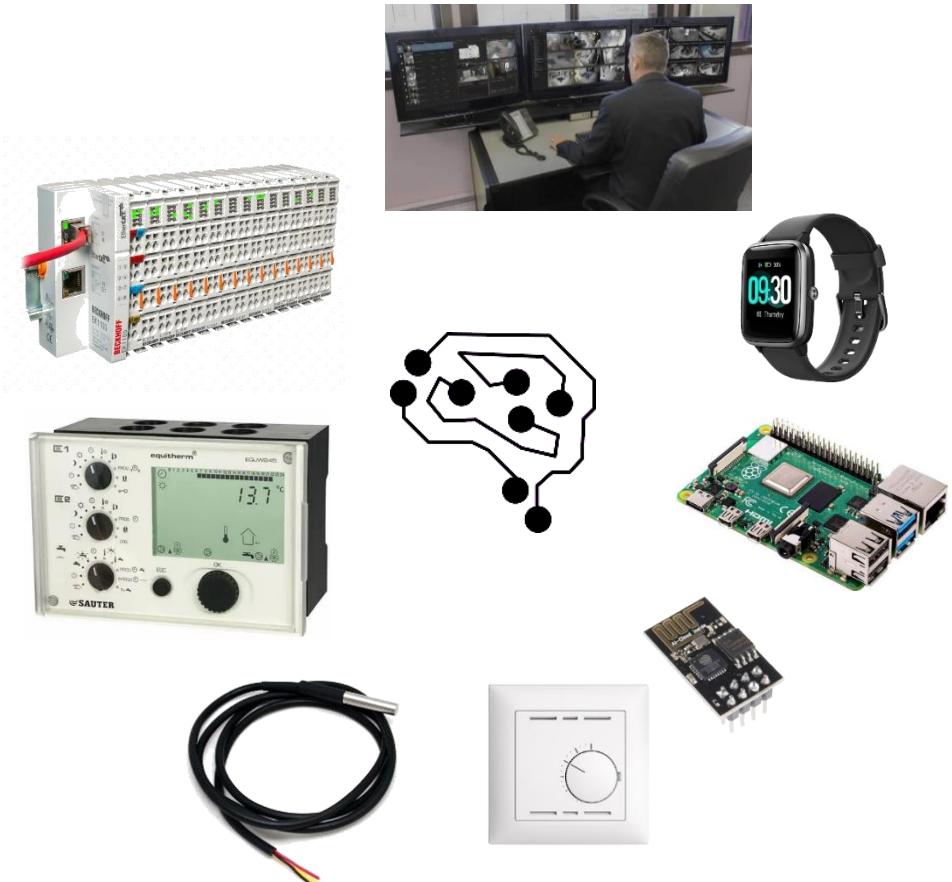
- Sicherheit
- Komfort
- (Energie) Effizienz

Smarter Ansatz: Gebäudeautomation ist eine **Plattform** die in einem grösseren Kontext, voll integriert, **Use cases** mehrerer Stakeholder bedient.

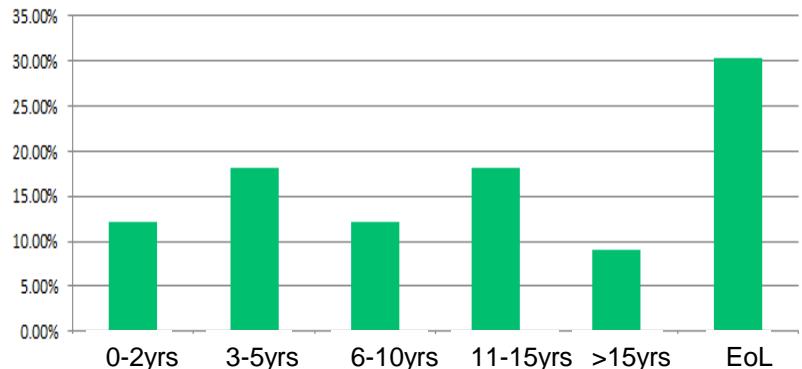
e.g. 49 Use cases in
«Navigating_SmartBuildings_Whitepaper»
<https://crem.locatee.com/use-case-navigator>



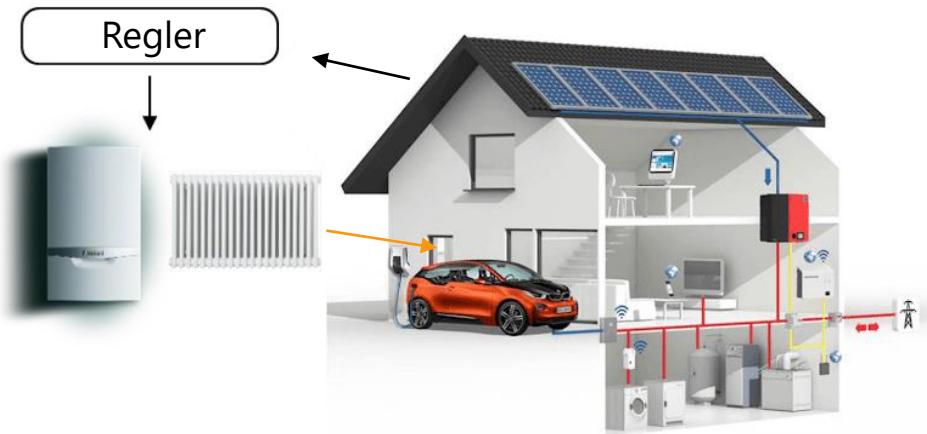
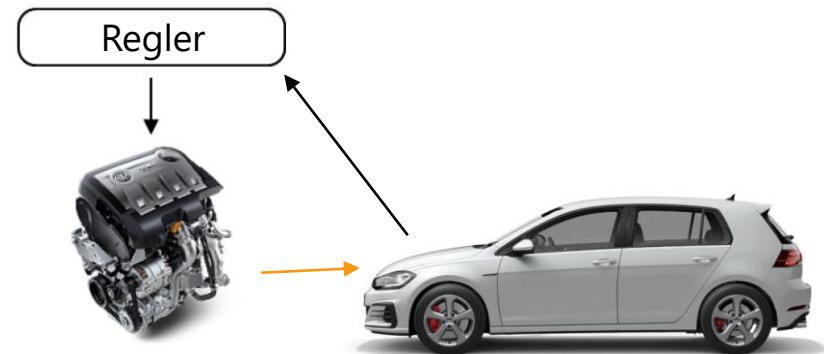
Gebäudeautomation im Wandel?



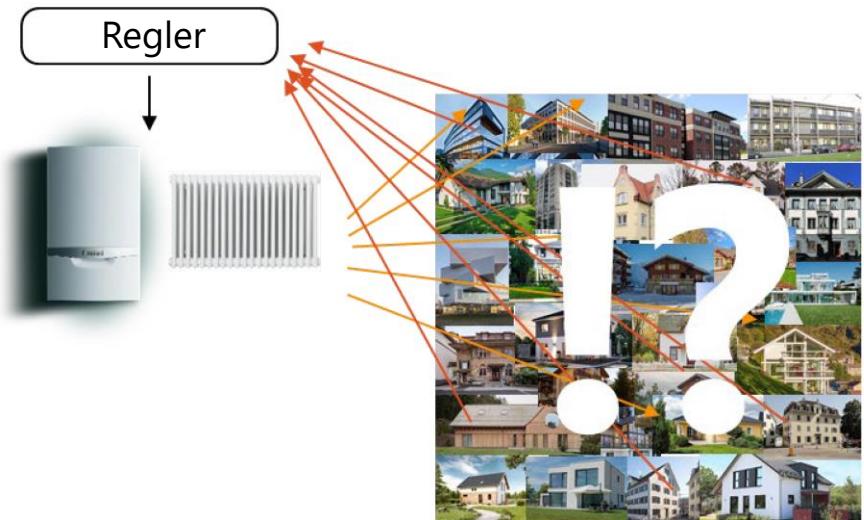
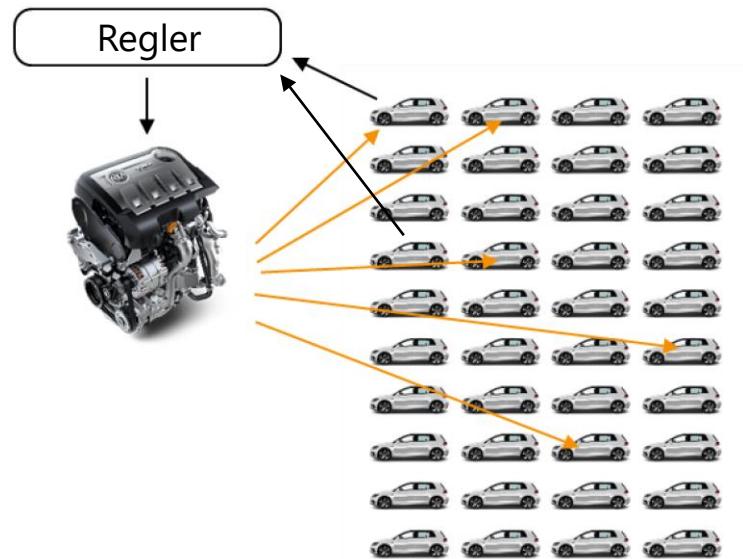
Wie oft passen sie ihre GA an?



Regler in verschiedenen Branchen



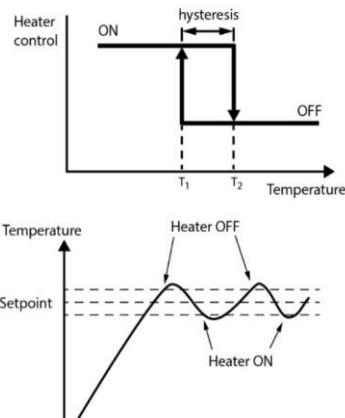
Regler in verschiedenen Branchen



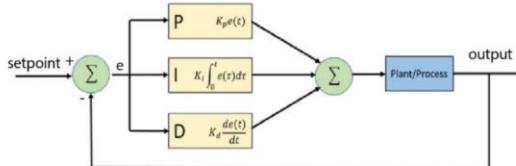
Adaptive Regler, lernende Regler



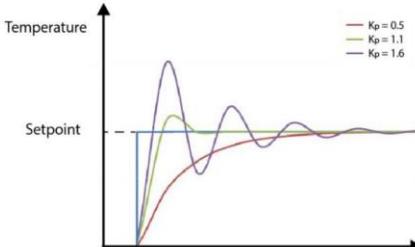
Regelbasiert



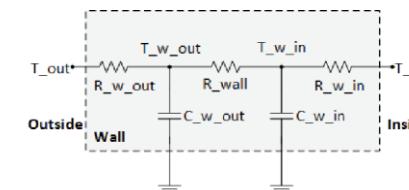
Hysterese



PID



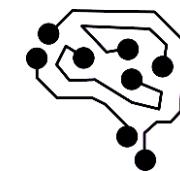
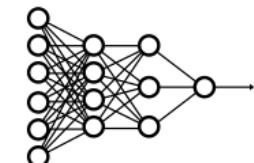
Modell basiert



$$\begin{aligned} & \text{minimize}_{x, u} \sum_{j=0}^{N-1} J(x_{k+j+1}, u_{k+j}) \\ & \text{subject to } x_{k+j+1} = f(x_{k+j}, u_{k+j}, d_{k+j}) \\ & \quad (x_{k+j+1}, u_{k+j}) \in (\mathcal{X}, \mathcal{U}) \\ & \quad \forall j \in [0, \dots, N-1] \end{aligned}$$

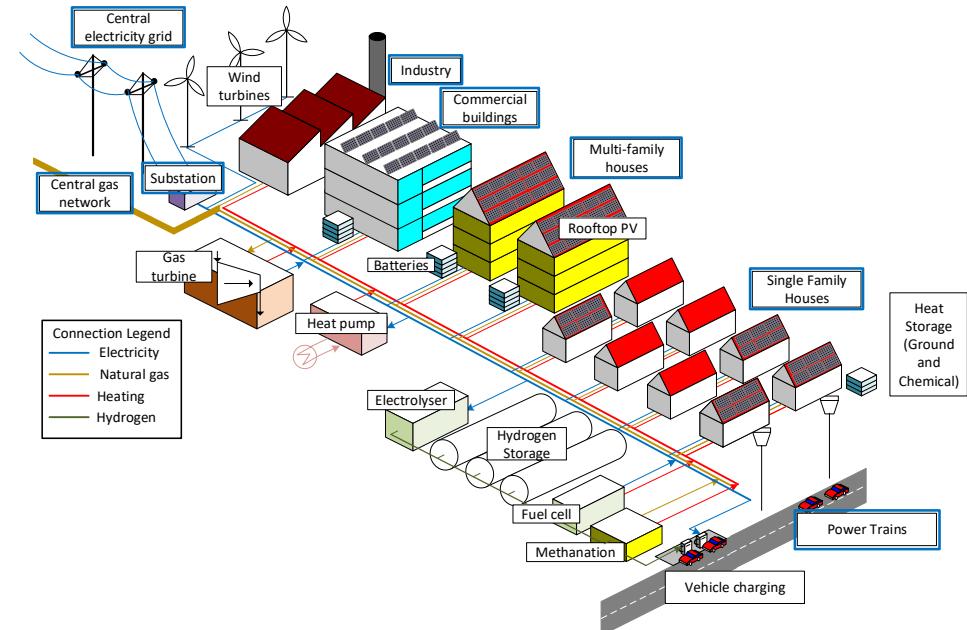
MPC

Lernende Regler



DPC

Der ehub demonstrator der Empa



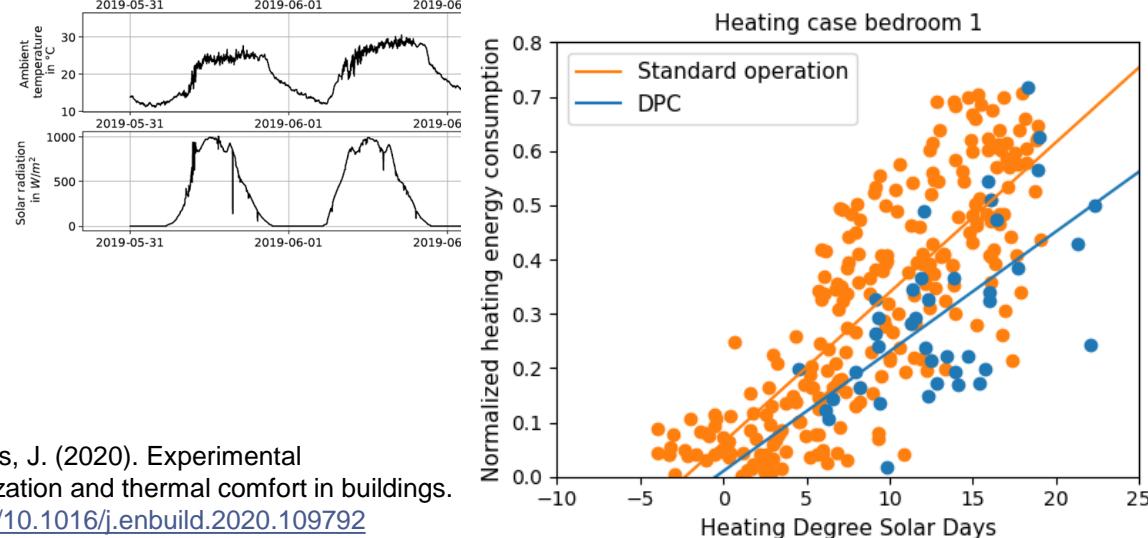
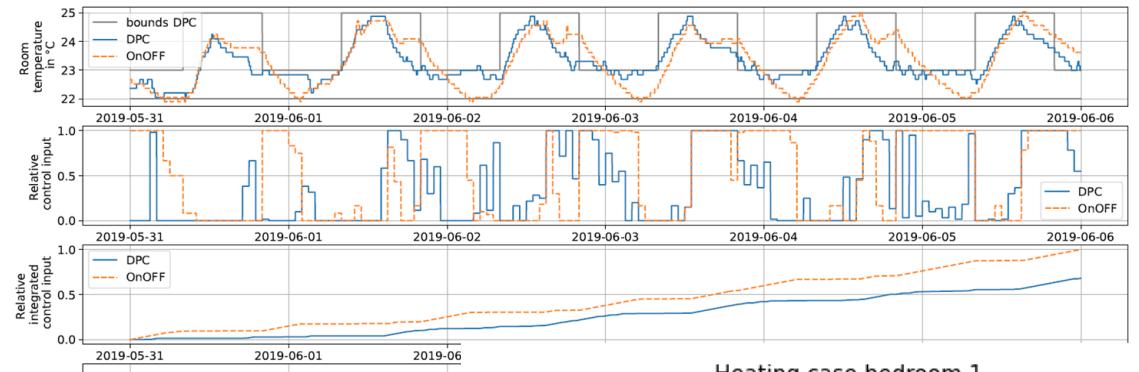
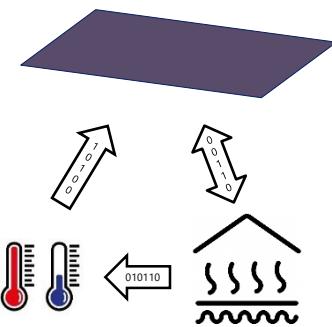
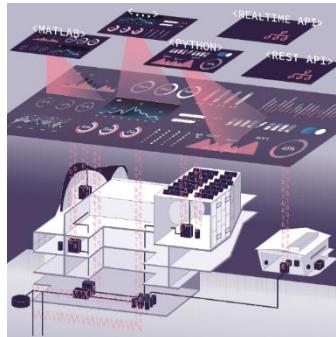
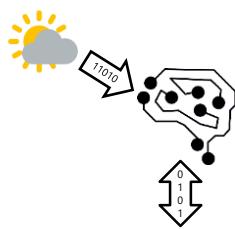
- 6 Heat pumps
- 3 Thermal buffers
- 1 Ice storage unit
- ⋮
- 2 Batteries
- 7 PV and thermal collectors
- 1 EV charging station
- ⋮
- 4 Thermal networks
- 4 Electrical networks

500+ Actors
1100+ Sensors
8000+ Datapoints

multi energy system

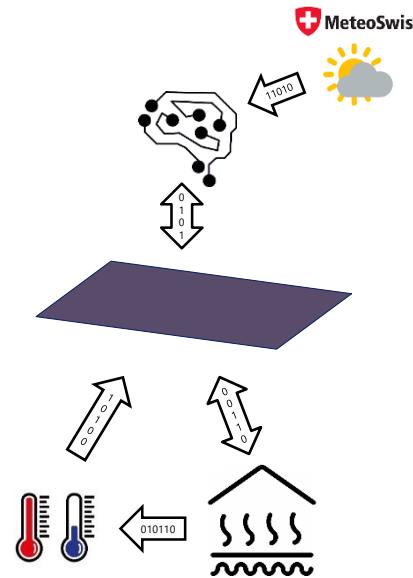
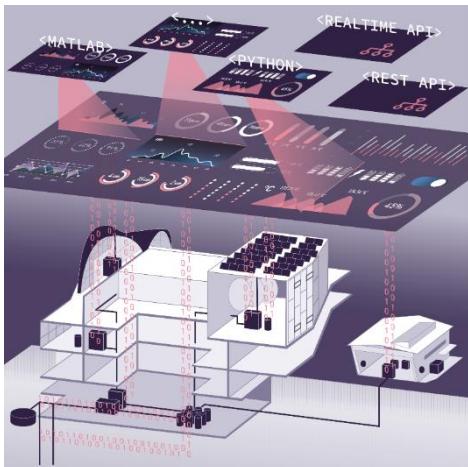
Project: Data Predictive Control – heizen und kühlen mit KI

MeteoSwiss



Büning, F., Huber, B., Heer, P., Aboudonia, A., & Lygeros, J. (2020). Experimental demonstration of data predictive control for energy optimization and thermal comfort in buildings. Energy and Buildings, 211, 109792 (8 pp.). <https://doi.org/10.1016/j.enbuild.2020.109792>

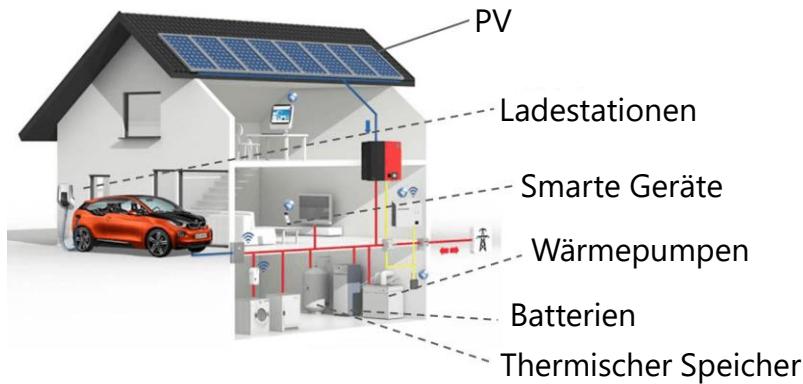
Project: Data Predictive Control – heizen und kühlen mit KI



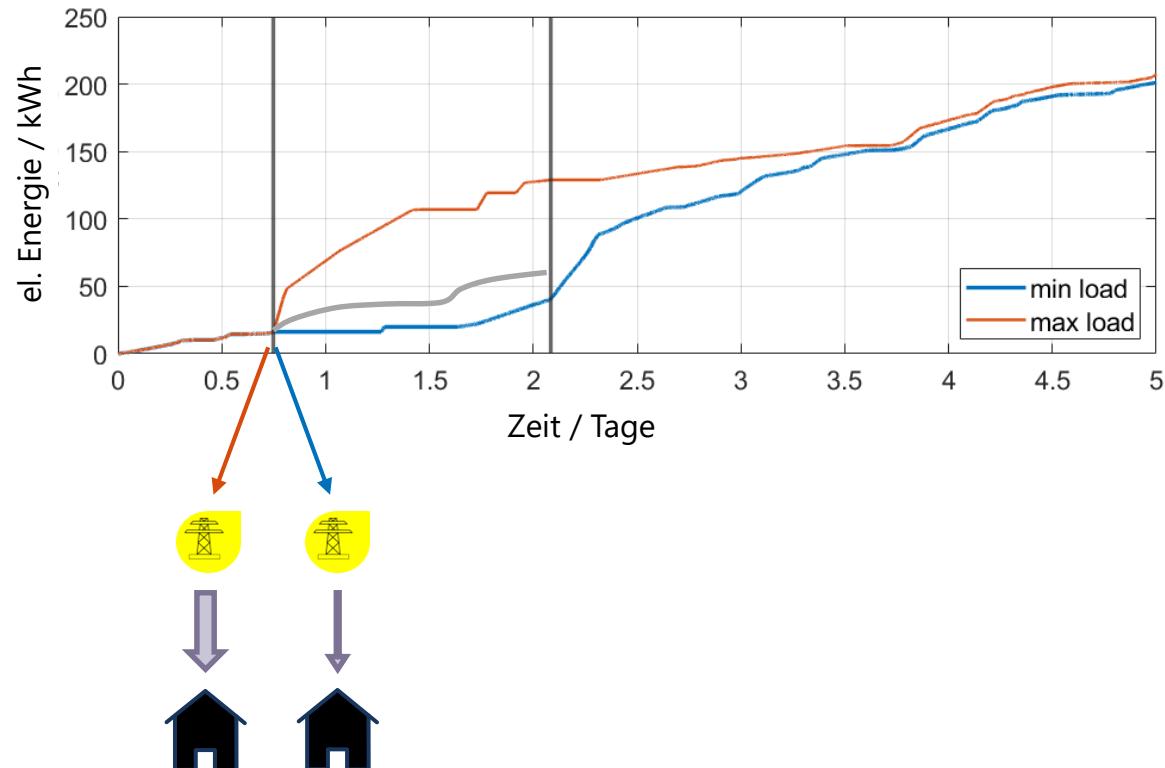
25% of heating and cooling energy can be saved with a predictive controller.

It is possible to achieve both objectives at the same time:
reducing energy cost
increasing comfort

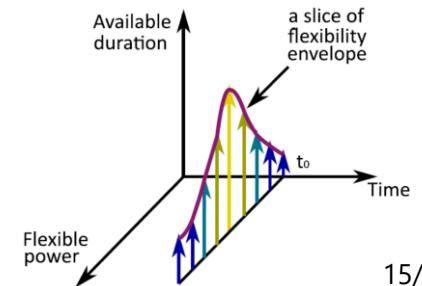
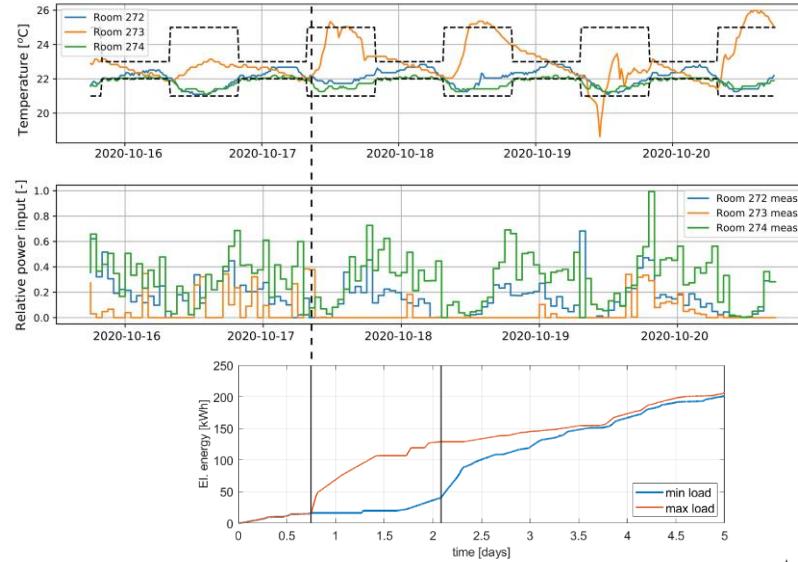
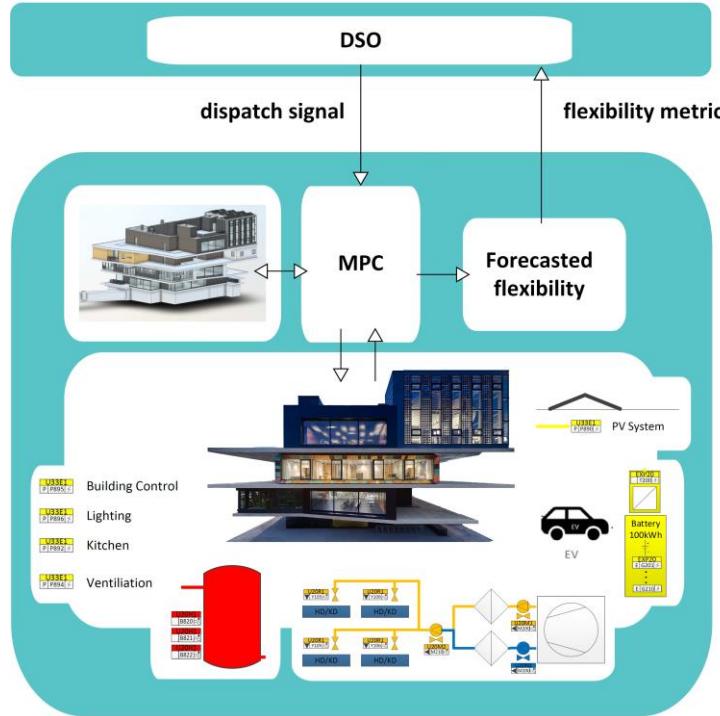
Von Gebäuden zu Quartieren und Städten



Von Gebäuden zu Quartieren und Städten

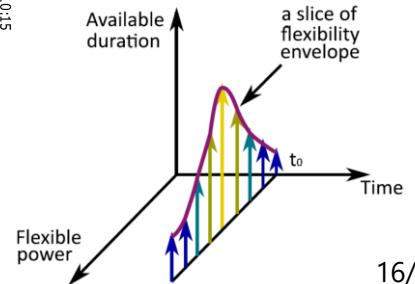
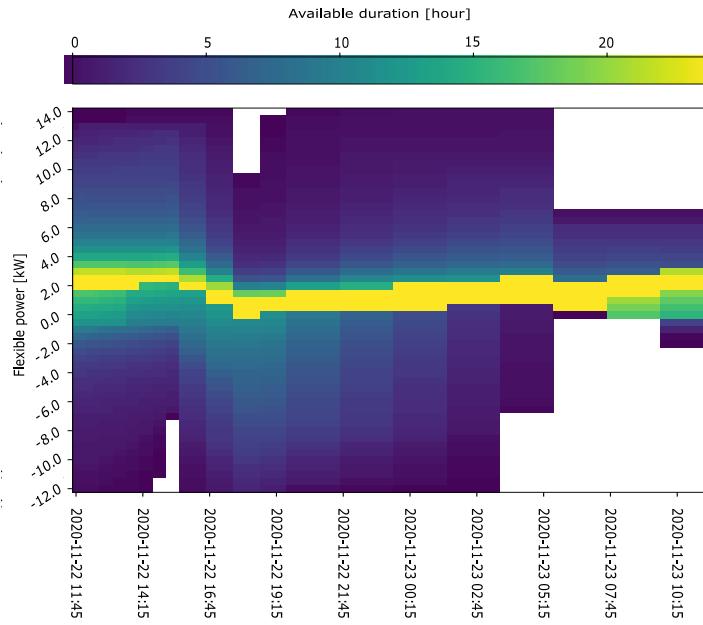
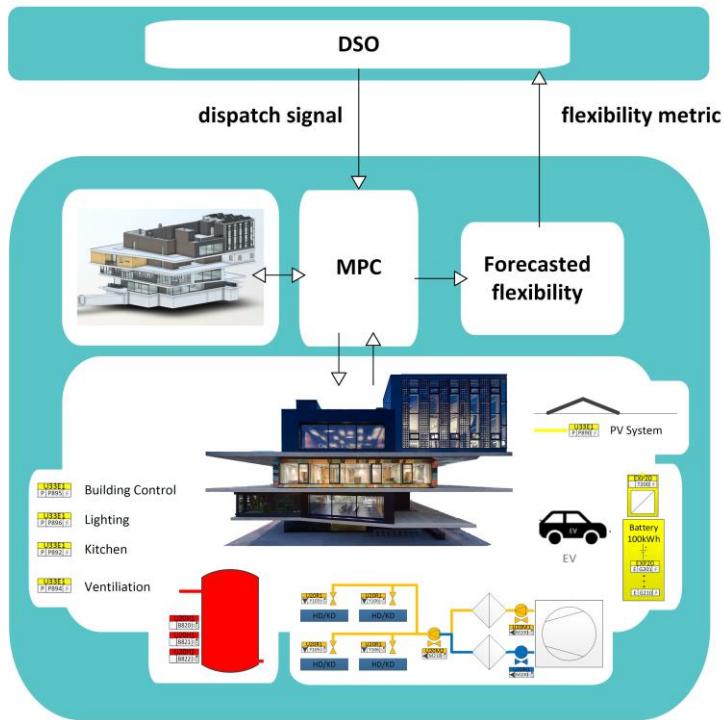


Project: Benefits in districts – quantify and predict energetic flexibility



Gasser, J., Cai, H., Karagiannopoulos, S., Heer, P., & Hug, G. (2021). Predictive energy management of residential buildings while self-reporting flexibility envelope. *Applied Energy*, 288, 116653 (14 pp.).
<https://doi.org/10.1016/j.apenergy.2021.116653>

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<https://doi.org/10.1016/j.apenergy.2021.116653>

Schweizer Haushalt: 4'500kWh/a -> $\varnothing 12.33\text{kWh/d}$



17kW Sicherung (GLZ 0.6)

Tesla Model 3^{*):} 50kWh



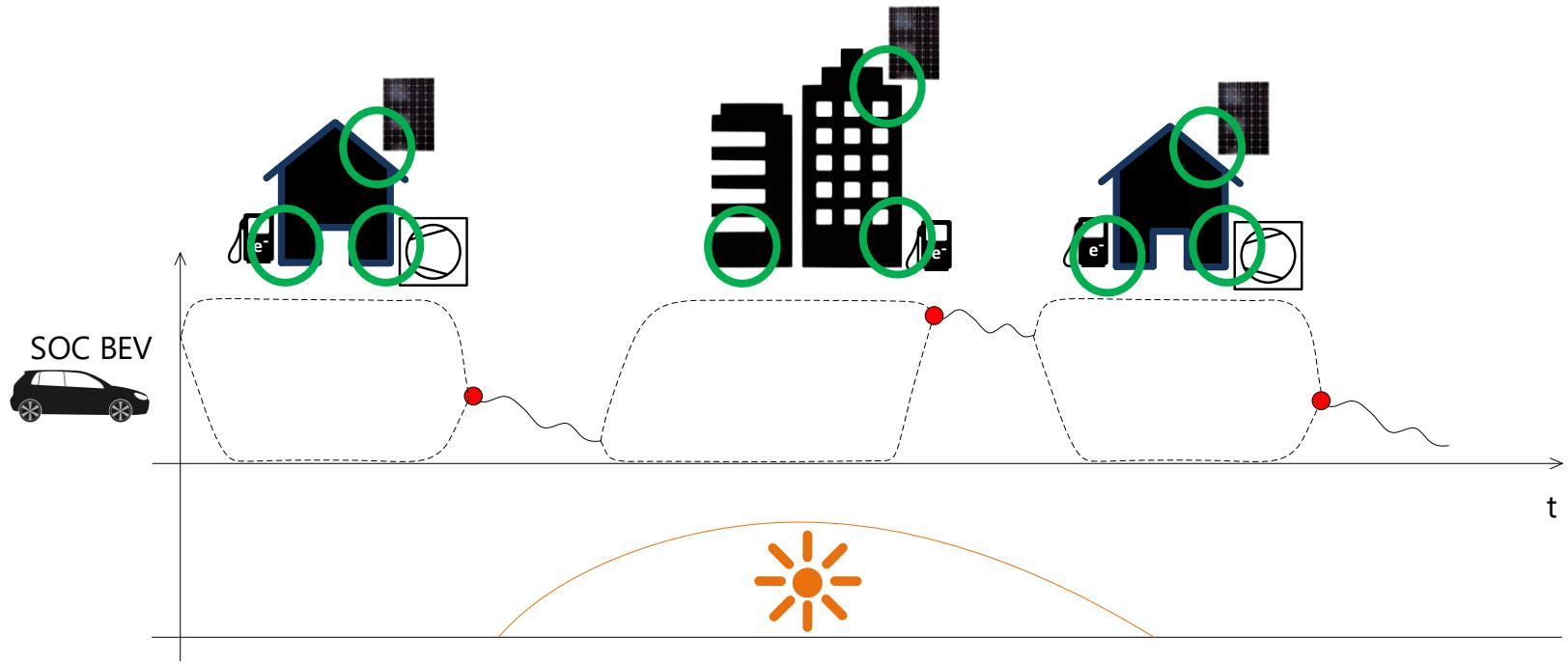
14,1 kWh/100km

3, 11, 22 kW

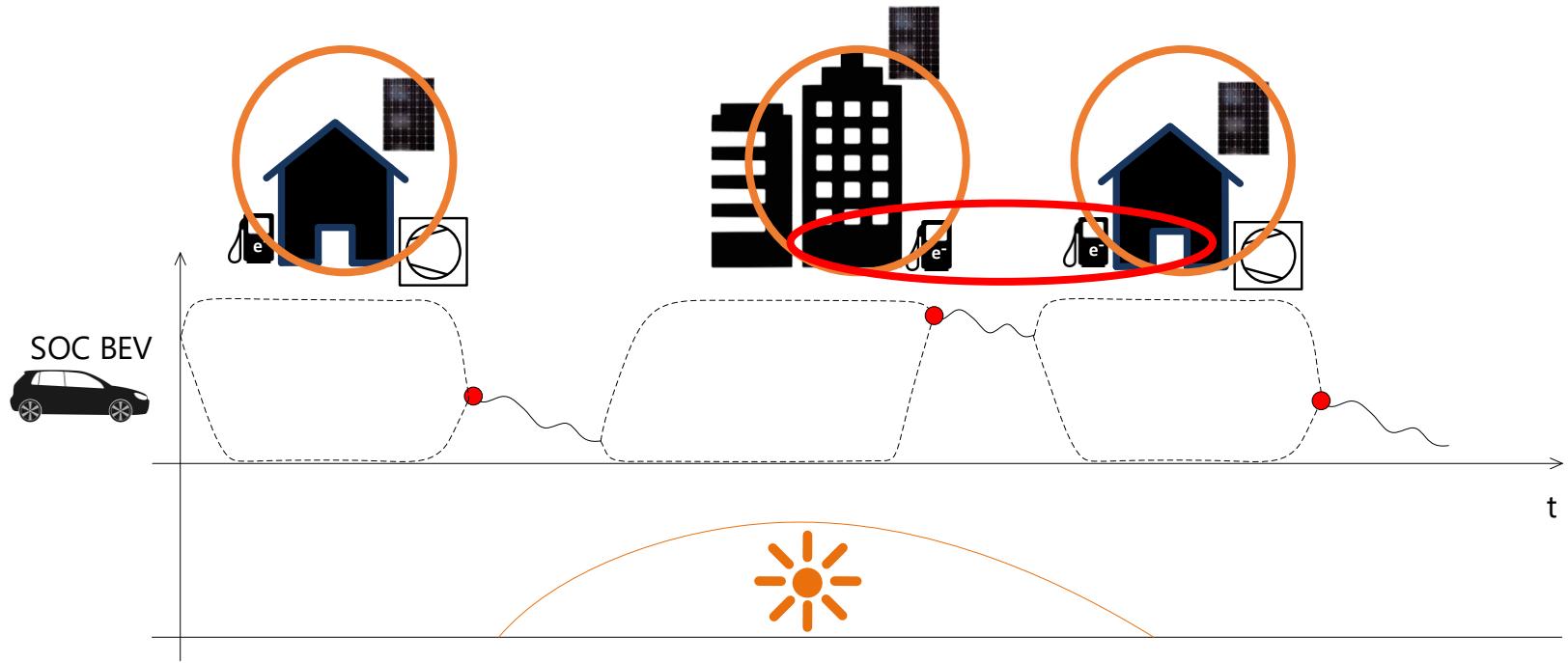
Unterschiedliche Lastprofile + nicht 100% Auslastung -> Potential für Flexibilität

^{*)} Meist verkauftes BEV 2019 in CH

Mobilität für das Energiesystem



Mobilität für das Energiesystem



RDF – accessible and understandable data

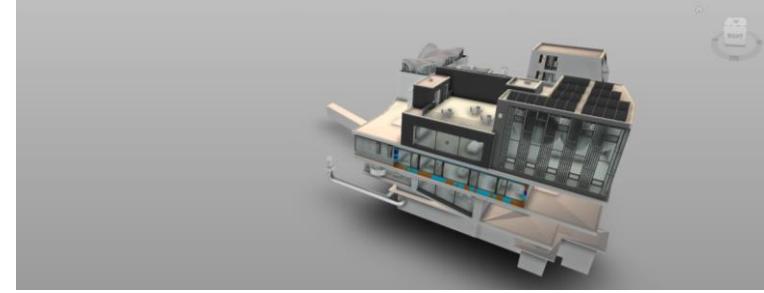


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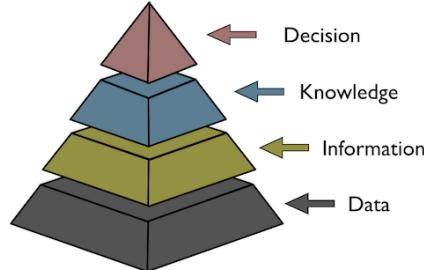
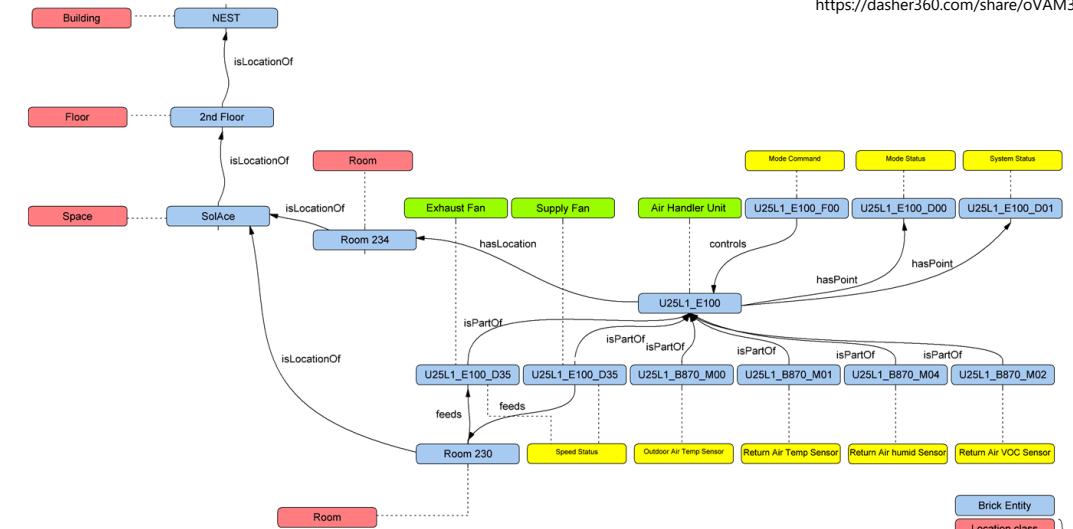
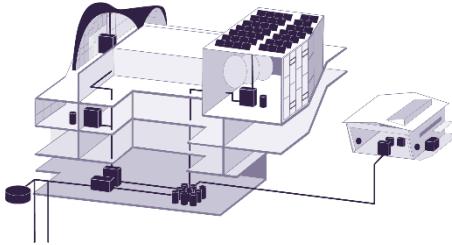
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<https://visualizer.nestcollaboration.ch/Realtime/data/3200000>

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<https://dasher360.com/share/oVAM3dF7M>



Key Take aways

- Dezentralisierung, Elektrifizierung, und Digitalisierung **verändern das Energiesystem** drastisch.
- Digitalisierung erlaubt eine **flexiblere Nutzung** von Technologie!
- **Gemessene Daten** können genutzt werden um das Verhalten von Systemen zu verbessern und anzupassen.
- **Lernende Systeme** können die Verbreitung Smarter Systeme stark beschleunigen und helfen die Ziele der Energiestrategie zu erreichen.
- Es braucht vereinheitlichte Beschreibungen von Daten damit digitalisierte Lösungen schneller **in den Markt** kommen.

Vielen Dank für ihre Aufmerksamkeit!

Philipp Heer
Deputy Head Urban Energy Systems Lab

philipp.heer@empa.ch
ehub.empa.ch
empa.ch/web/s313

