

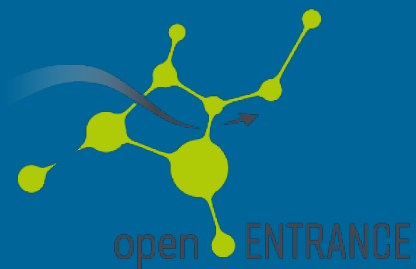
Berücksichtigung verschiedener Siedlungsmuster in der Entwicklung von lokalen Energiegemeinschaften mit Peer-to-Peer Handel

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IEWT Vienna 2021, 08.09.-10.09.2021

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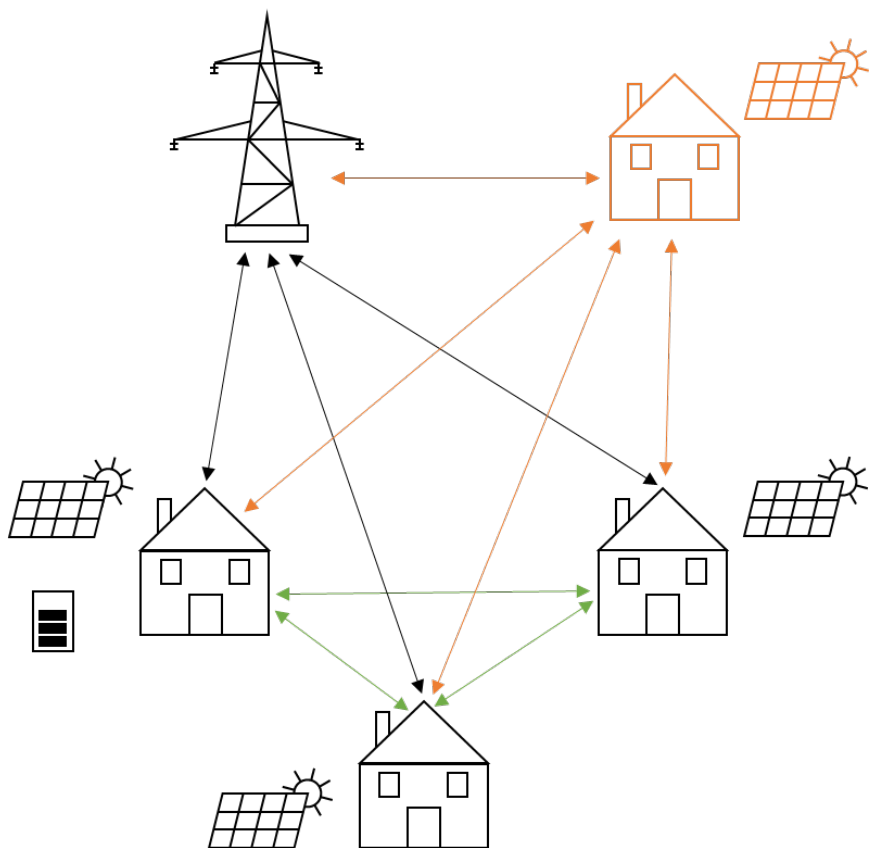
This project has received funding from the European Unions Horizon 2020 research and innovation program under grant agreement No. 835896



Motivation and Scope

- Photovoltaic (PV) systems: Decentralized electricity production and *prosumers*
- From individual self-consumption to collective self-consumption to active participants
- Trading and sharing of PV within a certain framework: Energy communities and Peer-to-Peer Trading
- Clean Energy Package (CEP) legal instruments:
 - Member states to enable the entrance of active participants into the market
 - Definition of peer-to-peer trading
- Framework:
 - Voluntary participation and consideration of individual willingness-to-pay
 - Low entry barriers: No closed systems, but part of the distribution network
 - Dynamic participation
- **Research question:**
 - Dynamic participation in peer-to-peer trading communities depending on different settlement patterns

Motivation and Scope



Scope:

- Optimizing energy communities within different settlement patterns over several years:
 - Considering phase-in/phase-out of prosumers
 - Assuming that local energy markets are more established in the future
 - Operating model of existing prosumers who want to participate in a local energy community

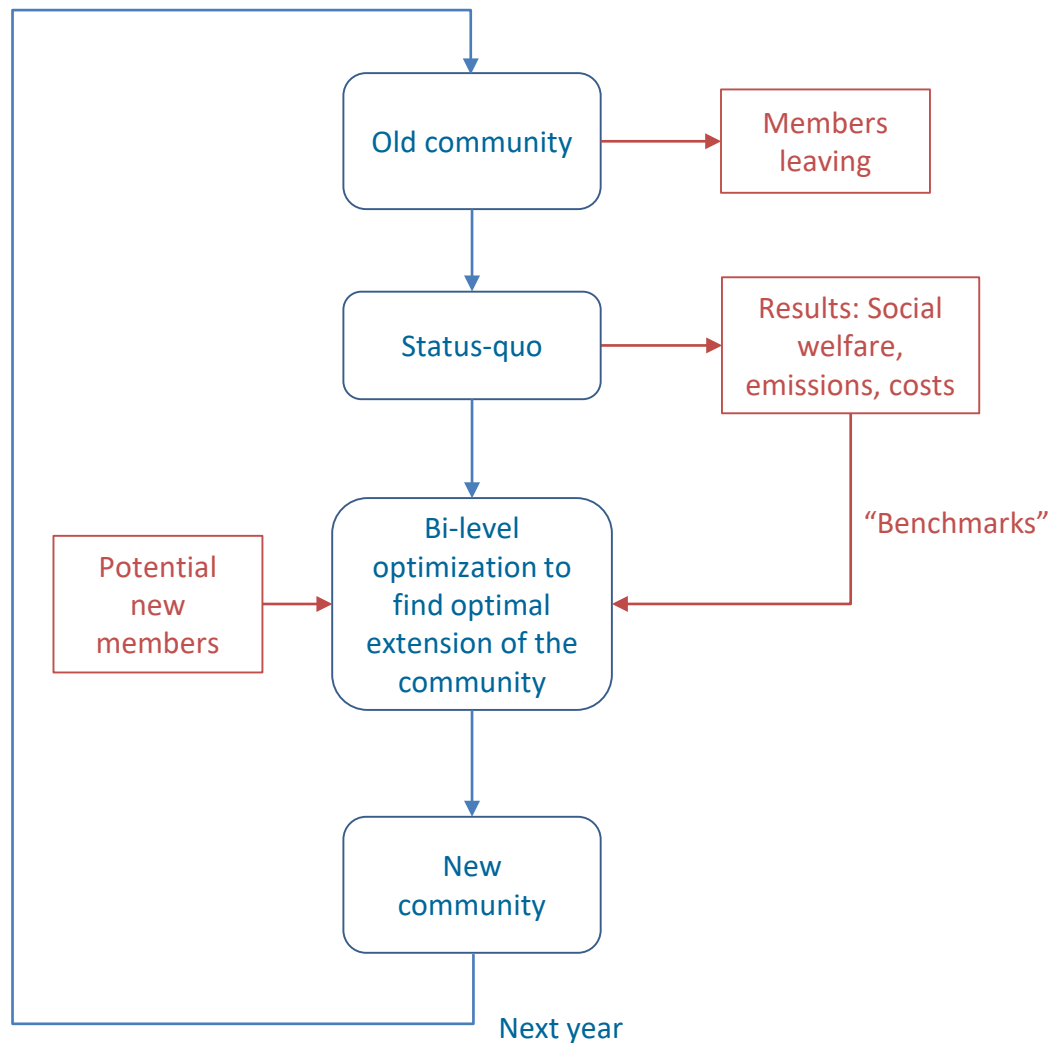
About the model:

- Linear optimization model FRESH:COM [1] maximizing the social welfare of a **local energy community**
- Allocation mechanism: Peer-to-peer trading under the consideration of each prosumer's *individual willingness-to-pay*
- Members: Private households and SMEs
 - Photovoltaic (PV) and Battery Energy Storage Systems (BESS)

Contribution:

- Extension of FRESH:COM to optimize *dynamic participation* in peer-to-peer trading communities within different settlement patterns

Modeling Approach



- Social welfare:

$$SW = \underbrace{\sum_{t \in T, i \in I} p_t^{G_{out}} q_{i,t}^{G_{out}} - \sum_{t \in T, i \in I} p_t^{G_{in}} q_{i,t}^{G_{in}}}_I + \underbrace{\sum_{t \in T, i, j \in I} wtp_{i,j,t} q_{i,j,t}^{share}}_{II}.$$

- Willingness-to-pay:

$$wtp_{i,j,t} = p_t^{G_{in}} + w_j(1 - d_{i,j}) \cdot e_t.$$

- „Benchmarks“:

$$\Delta costs_i = costs_i - costs_{i,old},$$

$$\Delta emissions_i = emissions_i - emissions_{i,old}.$$

Modeling Approach – Bi-level problem

- Upper level problem (“*leader*”):
 - Selecting the optimal electricity demand and PV capacity of new prosumers to fulfill certain requirements set by the original community members
 - Minimizing the cost-emission function CE:

$$CE = \sum_{i \in \mathcal{I}_{old}} \alpha_i \Delta costs_i + (1 - \alpha_i) \Delta emissions_i$$

- $\Delta costs_i$ and $\Delta emissions_i$ are the changes of annual costs and emissions of prosumer i , respectively.
- $\alpha_i \in [0,1]$ is individual weighting factor of prosumer i
- $b_i \in (0,1)$ are binary decision variables

$$\min_{\{load_i, PV_i, b_i, Q_{i,t}\}} \sum_{i \in \mathcal{I}_{old}} \alpha_i \Delta costs_i + (1 - \alpha_i) \Delta emissions_i$$

subject to:

$$b_i \cdot load_i^{min} \leq load_i \leq b_i \cdot load_i^{max} \quad \forall i \in \mathcal{I}_{new}$$

$$b_i \cdot PV_i^{min} \leq PV_i \leq b_i \cdot PV_i^{max} \quad \forall i \in \mathcal{I}_{new}$$

$$\sum_{i \in \mathcal{I}_{new}} b_i = n$$

Modeling Approach – Bi-level problem

- Lower level problem (“*follower*”):

- Maximizing the social welfare of the community, given the new prosumers' parameters selected in the upper problem

- Two parts in social welfare SW:

- Maximizes the overall self-consumption of the community and
 - Optimally distributes PV generation between the prosumers (peer-to-peer trading)

- Constraints:

- Covering electricity demand and PV generation
 - Battery storage operation

$$\max_{Q_{i,t}} \sum_{t \in \mathcal{T}, i \in \mathcal{I}} p_t^{G_{out}} q_{i,t}^{G_{out}} - \sum_{t \in \mathcal{T}, i \in \mathcal{I}} p_t^{G_{in}} q_{i,t}^{G_{in}} + \sum_{t \in \mathcal{T}, i, j \in \mathcal{I}} wtp_{i,j,t} q_{i,j,t}^{share}$$

subject to:

$$q_{i,t}^{G_{in}} + q_{i,t}^{B_{out}} + \sum_{j \in \mathcal{I}} q_{j,i,t}^{share} - q_{i,t}^{load} = 0 \quad (\lambda_{i,t}^{load}) \quad \forall i \in \mathcal{I}_{old}, t$$

$$q_{i,t}^{G_{out}} + q_{i,t}^{B_{in}} + \sum_{j \in \mathcal{I}} q_{i,j,t}^{share} - q_{i,t}^{PV} = 0 \quad (\lambda_{i,t}^{PV}) \quad \forall i \in \mathcal{I}_{old}, t$$

$$q_{i,t}^{G_{in}} + q_{i,t}^{B_{out}} + \sum_{j \in \mathcal{I}} q_{j,i,t}^{share} - load_i q_{i,t}^{load} = 0 \quad (\lambda_{i,t}^{load}) \quad \forall i \in \mathcal{I}_{new}, t$$

$$q_{i,t}^{G_{out}} + q_{i,t}^{B_{in}} + \sum_{j \in \mathcal{I}} q_{i,j,t}^{share} - PV_i q_{i,t}^{PV} = 0 \quad (\lambda_{i,t}^{PV}) \quad \forall i \in \mathcal{I}_{new}, t$$

$$SoC_{i,t-1} + q_{i,t}^{B_{in}} \cdot \eta^B - q_{i,t}^{B_{out}} / \eta^B - SoC_{i,t} = 0 \quad (\lambda_{i,t}^{SoC}) \quad \forall i, t > t_0$$

$$SoC_{i,t=t_{end}} + q_{i,t_0}^{B_{in}} \cdot \eta^B - q_{i,t_0}^{B_{out}} / \eta^B - SoC_{i,t_0} = 0 \quad (\lambda_{i,t_0}^{SoC}) \quad \forall i, t = t_0$$

$$SoC_{i,t} - SoC_i^{max} \leq 0 \quad (\mu_{i,t}^{SoC^{max}}) \quad \forall i, t$$

$$q_{i,t}^{B_{in}} - q_i^{B^{max}} \leq 0 \quad (\mu_{i,t}^{B_{in}^{max}}) \quad \forall i, t$$

$$q_{i,t}^{B_{out}} - q_i^{B^{max}} \leq 0 \quad (\mu_{i,t}^{B_{out}^{max}}) \quad \forall i, t$$

Modeling Approach – Bi-level problem

How is the bi-level problem solved?

- Transformation of the lower level problem with its corresponding KKT conditions (“*Karush-Kuhn-Tucker*”):
- Mathematical program with equilibrium constraints (MPEC)
- The equilibrium problem of the follower is parametrized by the leader’s decisions variables
- Formulation of a set of complementarity conditions
- Big-M transformation

Modeling approach – Settlement patterns

Characteristics of the different settlement patterns:

1. City areas (high population density)

- Multi-apartment buildings
 - Assuming voluntary participation of tenants
 - Aggregation of tenants' load profiles
 - Possibly with different types of businesses in the buildings (shops on the first floor, offices, ...)
 - Limited rooftop area for PV systems

2. Suburban areas (medium density)

- Mix of multi-apartment buildings and single family houses
- Some businesses included (e.g. shops, bakery, ...)

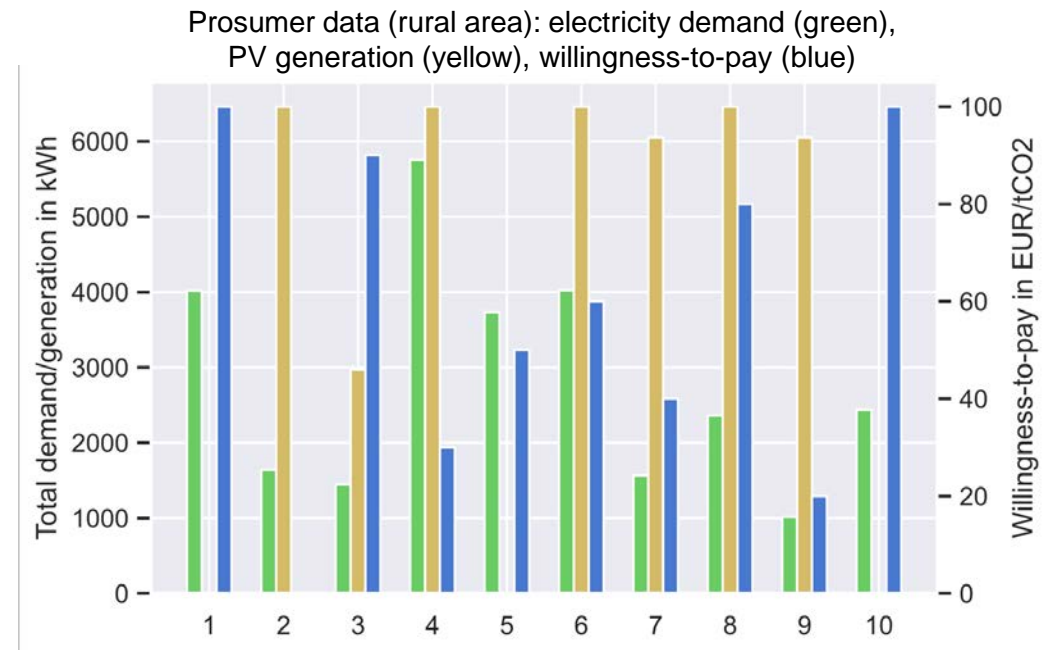
3. Rural areas (low population density)

- Mostly single family houses
- Sufficient rooftop area available

Modeling Approach – Data and assumptions

Case study:

- Model implemented in *Python* using *Pyomo*
- Small community set-up consisting of 10 prosumer + new prosumer
- Electricity demand: Modular households or houses from *Load Profile Generator* [1]
- PV generation: PV modules with different orientations (location: Vienna) from *renewables.ninja* [2]
- Annual hourly data is clustered in representative time periods using Python module *sklearn.cluster.Kmeans* [3]
- New prosumer:
 - Apartment building: $PV = 5 \text{ kW}_{peak}, load = 39000 \frac{kWh}{yr}$
 - Single house: $PV = 3 \text{ kW}_{peak}, load = 1400 \frac{kWh}{yr}$

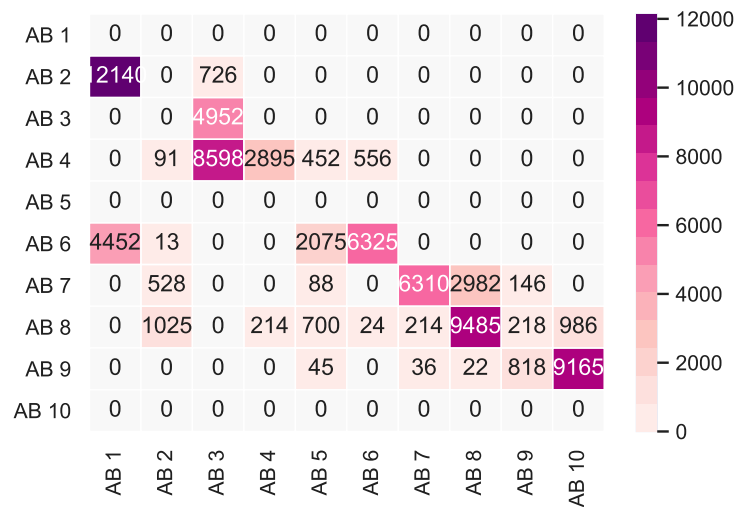


Results – Comparison of settlement patterns

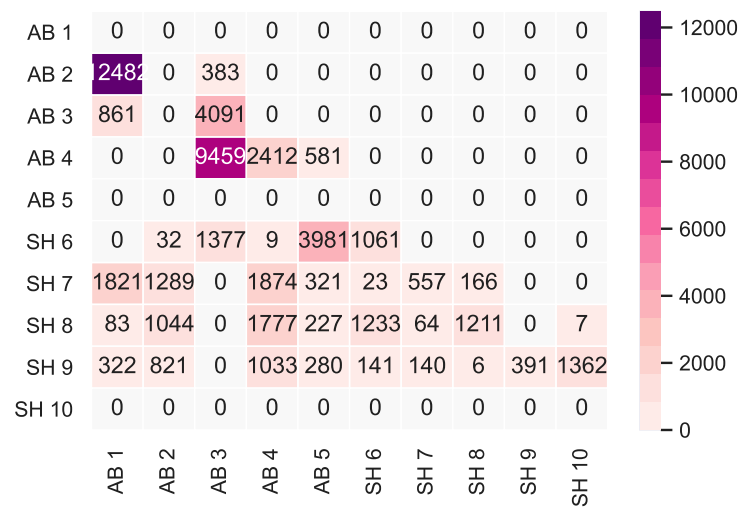
Comparison of peer-to-peer trading with 10 members:

- Annual results: **electricity trades with the peers**
- Different settlement patterns: city – suburban – rural

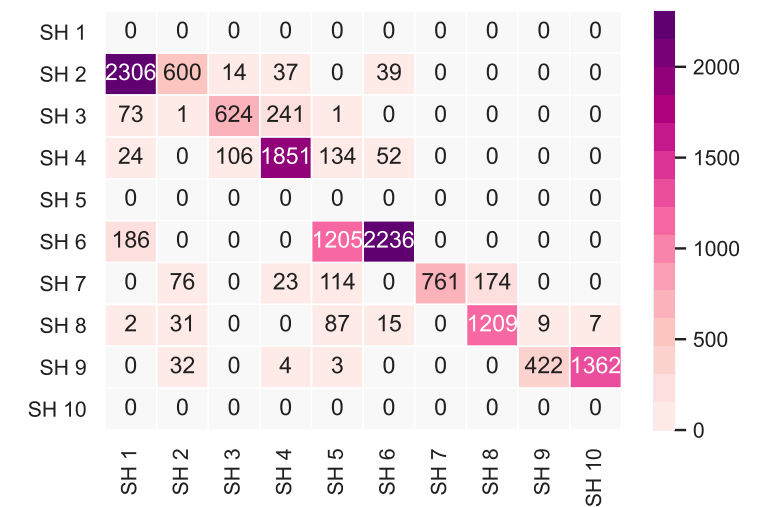
City



Suburban area



Rural area

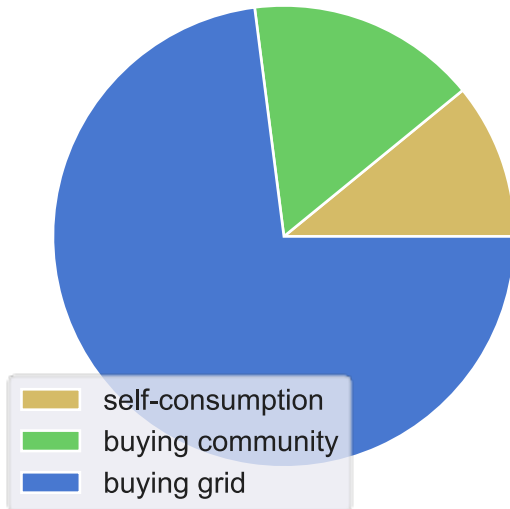


Results – Comparison of settlement patterns

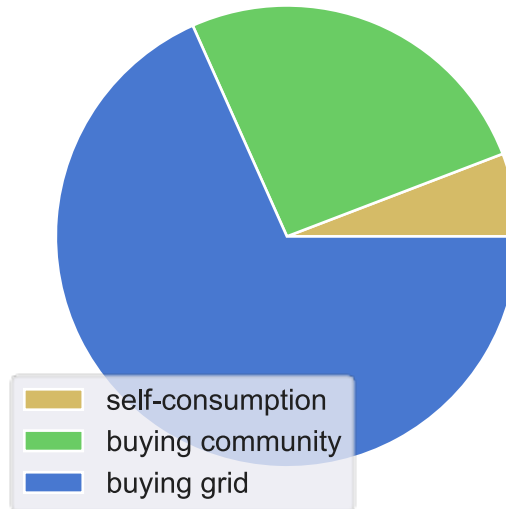
Comparison of peer-to-peer trading with 10 members:

- Annual results: self-consumption/trading with community/trading with grid
- Different settlement patterns: city – suburban – rural

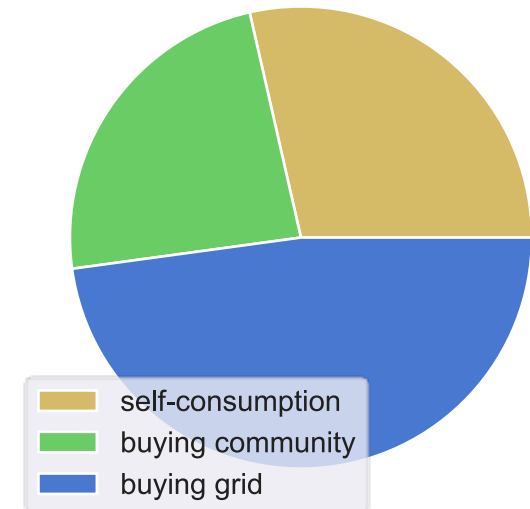
City



Suburban area



Rural area

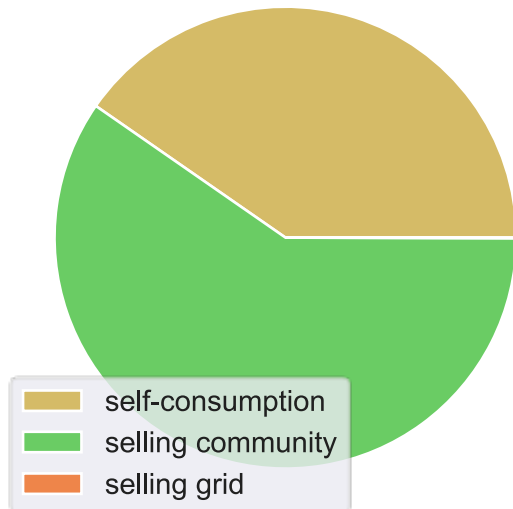


Results – Comparison of settlement patterns

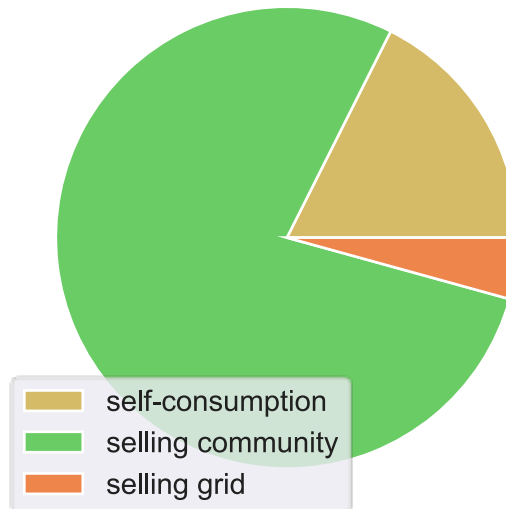
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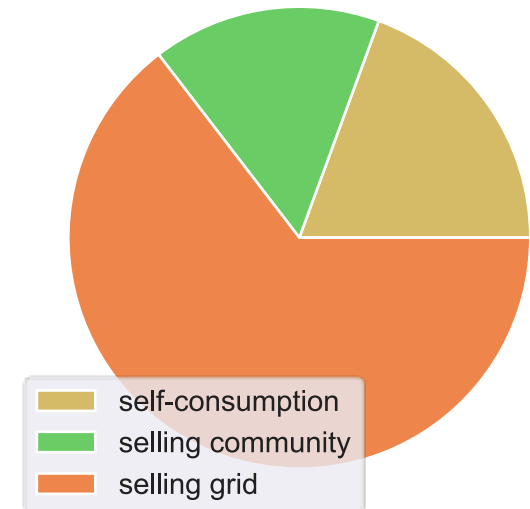
City



Suburban area



Rural area



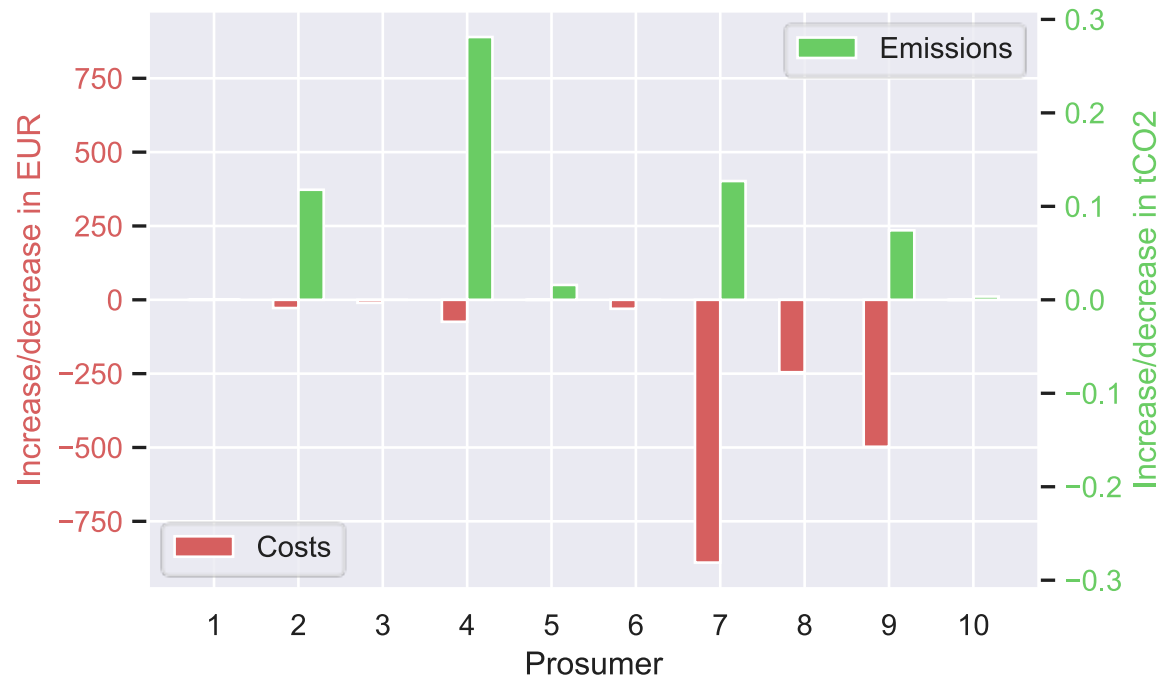
Results – Rural area

All prosumers want to minimize their individual costs:

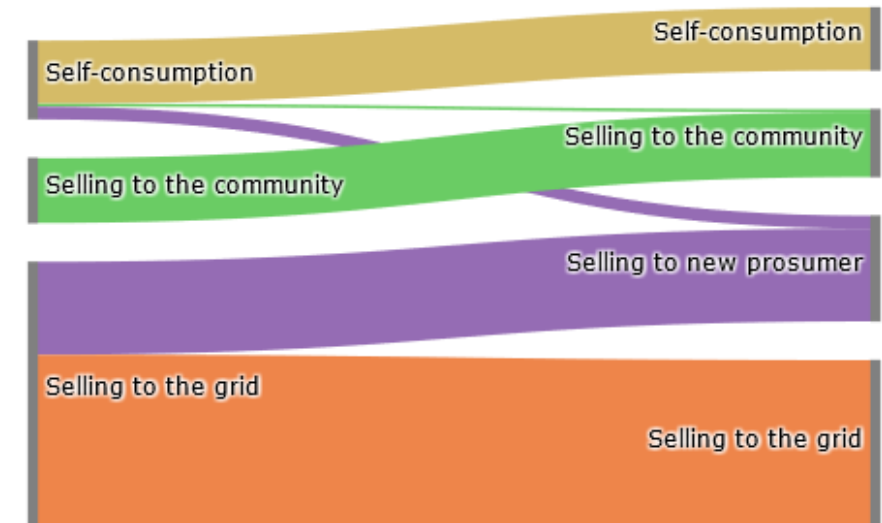
$$\alpha_i = 1, \forall i \in I_{old}$$

Results:

- ✓ Apartment building
- ✓ Opportunity to sell to the new prosumer (high demand, no PV installed) and lower annual costs



Sankey diagram of PV generation



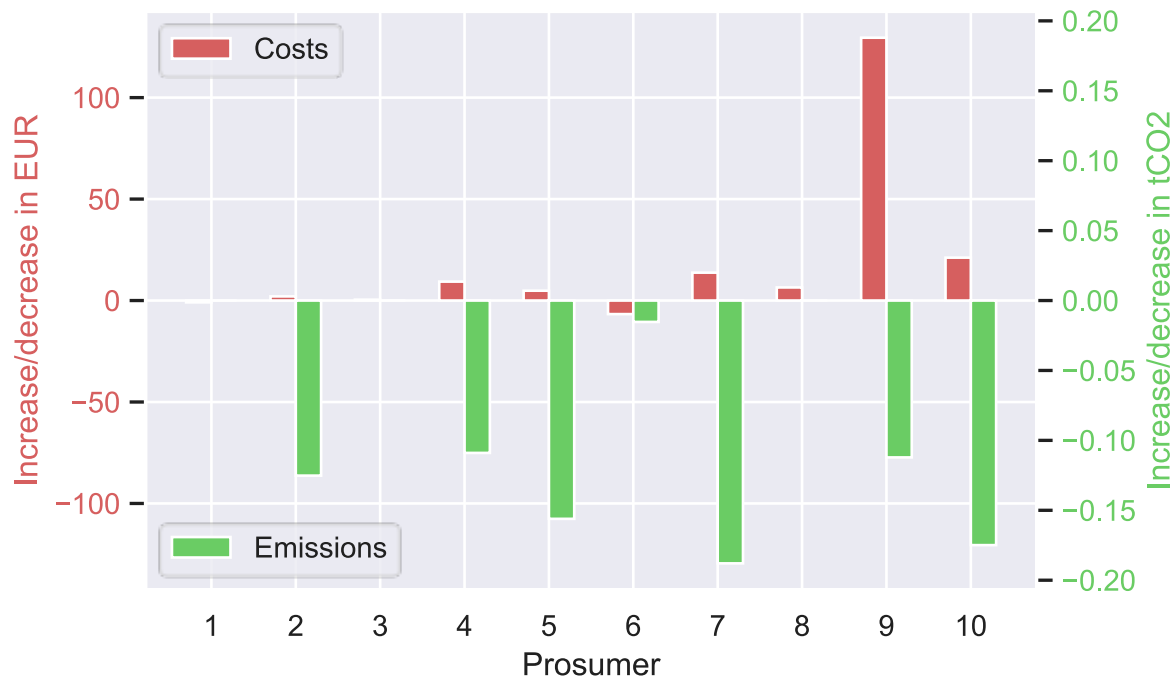
Results – City area

All prosumers want to minimize their individual emissions:

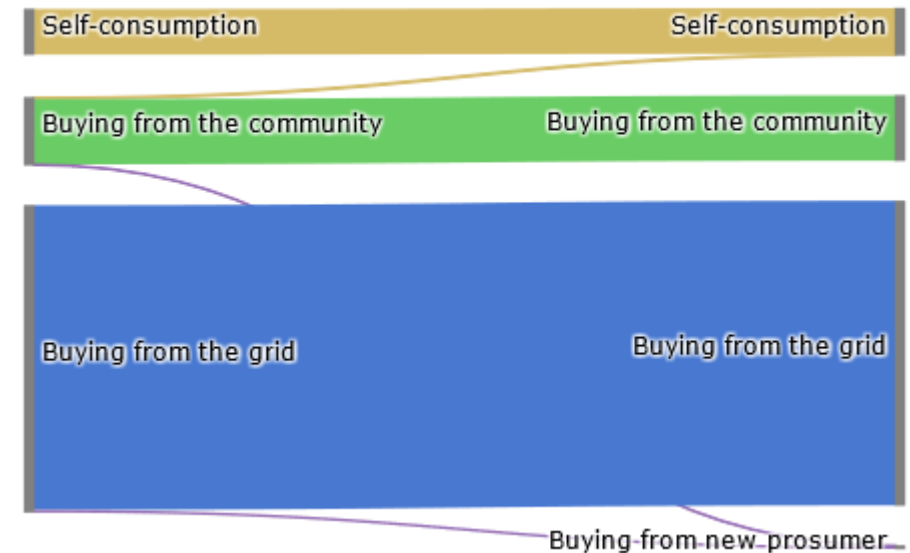
$$\alpha_i = 0, \forall i \in I_{old}$$

Results:

- ✓ Single house
- ✓ Emissions decrease, some costs increase



Sankey diagram of electricity demand



Results – Apartment building vs. single house

Influence of the willingness-to-pay:

	Minimize ind. emissions		Minimize ind. costs	
	AB	SH	AB	SH
City		✓	✓	
Suburban area		✓		✓
Rural area		✓	✓	

Conclusions

Findings:

- The model is able to choose between potential prosumer
- Balancing the needs of environmental- and profit-oriented members
- Aiming for a diverse set-up of actors
- Ultimately, the energy community has to be able to attract suitable potential new members to guarantee its performance over the years

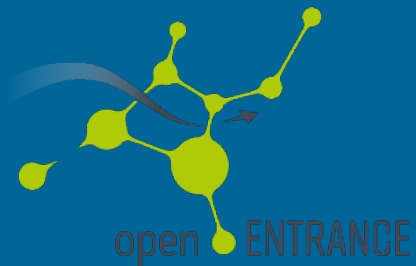
Future outlook:

- Analysis of the effects on the DSO and the community manager
- Behavior of prosumers in urban areas vs. rural areas

Thank you for your attention!

GitHub

<https://github.com/tperger/FRESH-COM>



open ENergy TRansition ANalyses for a
low-Carbon Economy

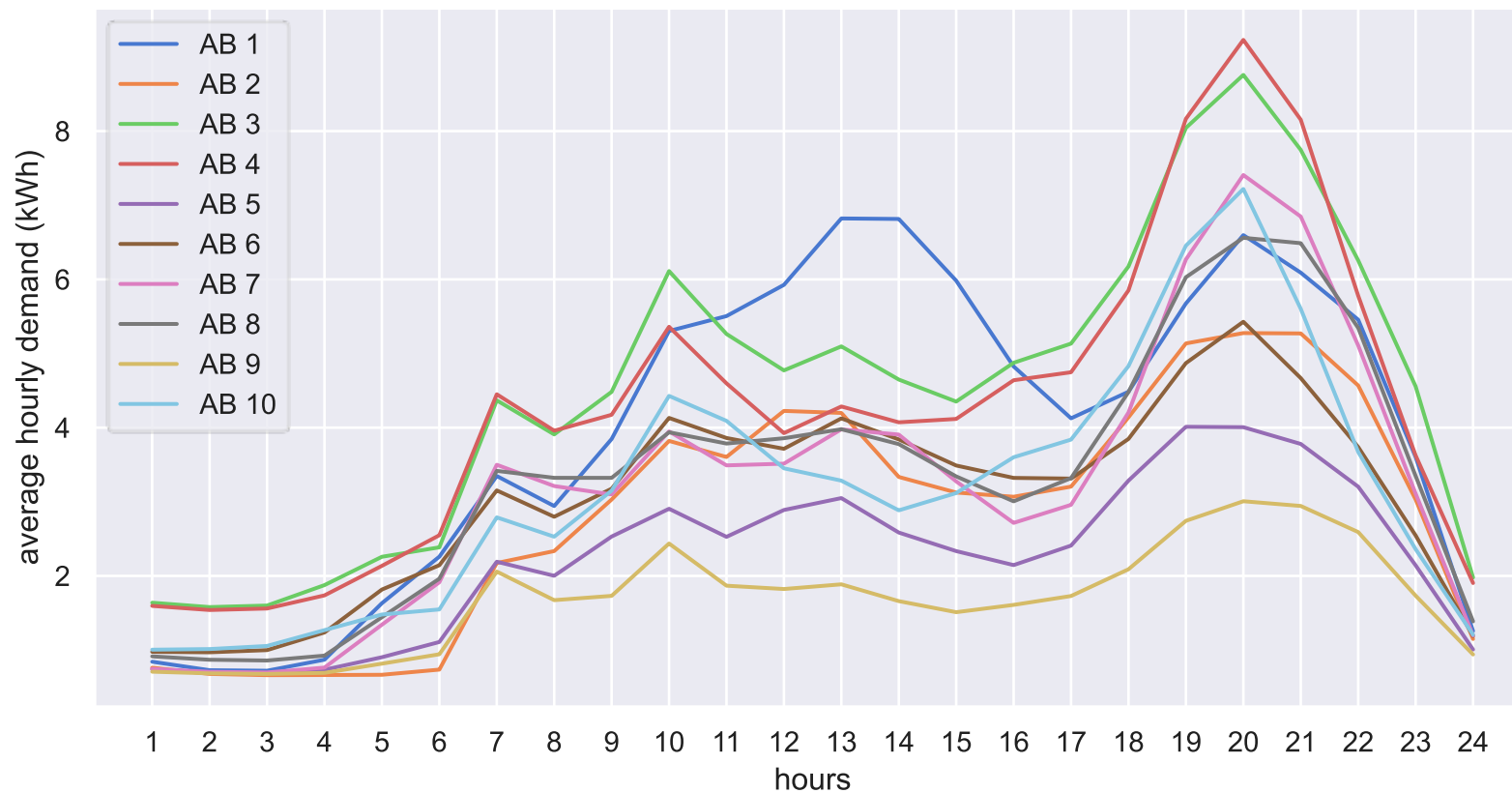
<https://openentrance.eu/>

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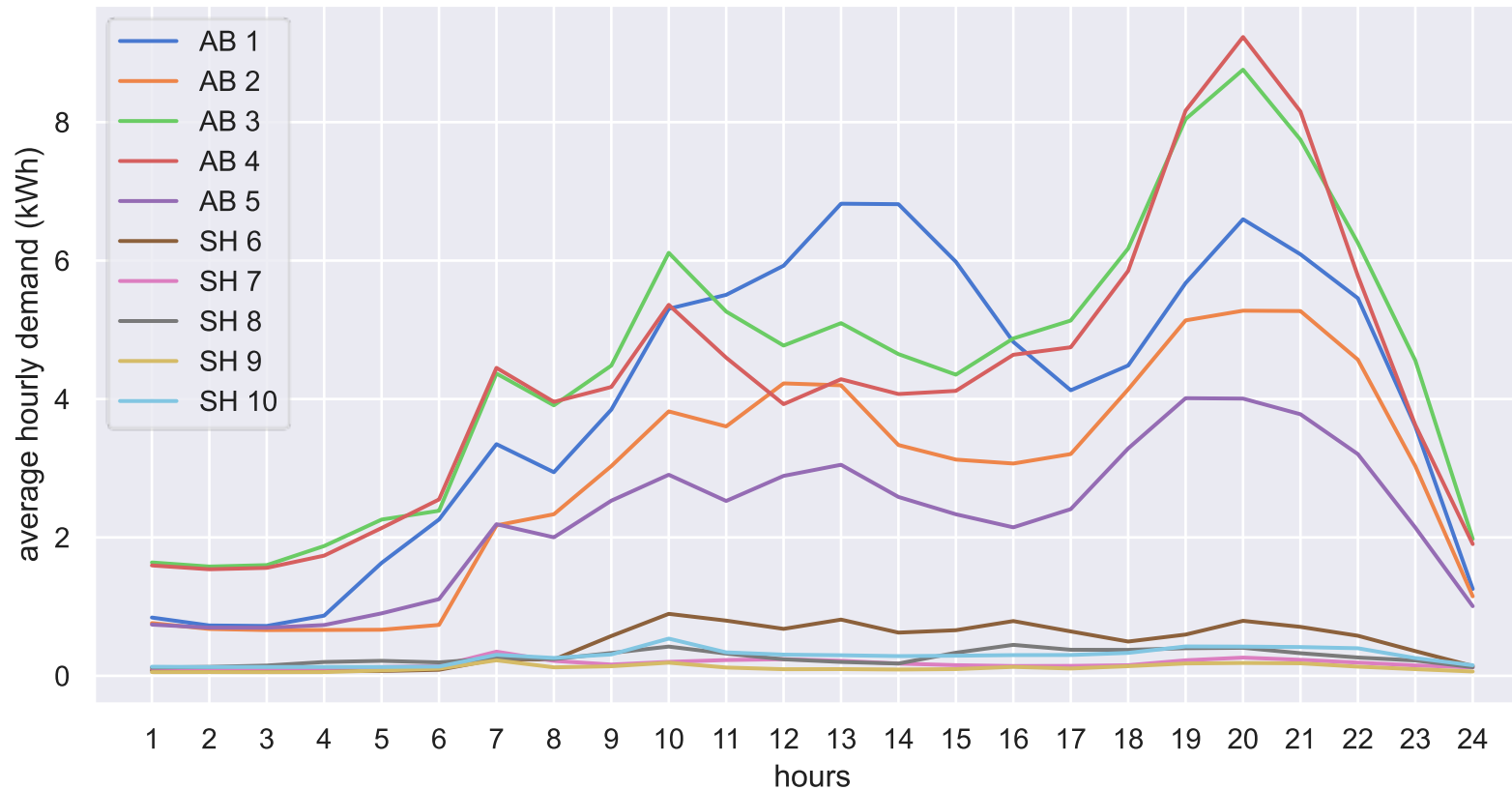


Appendix

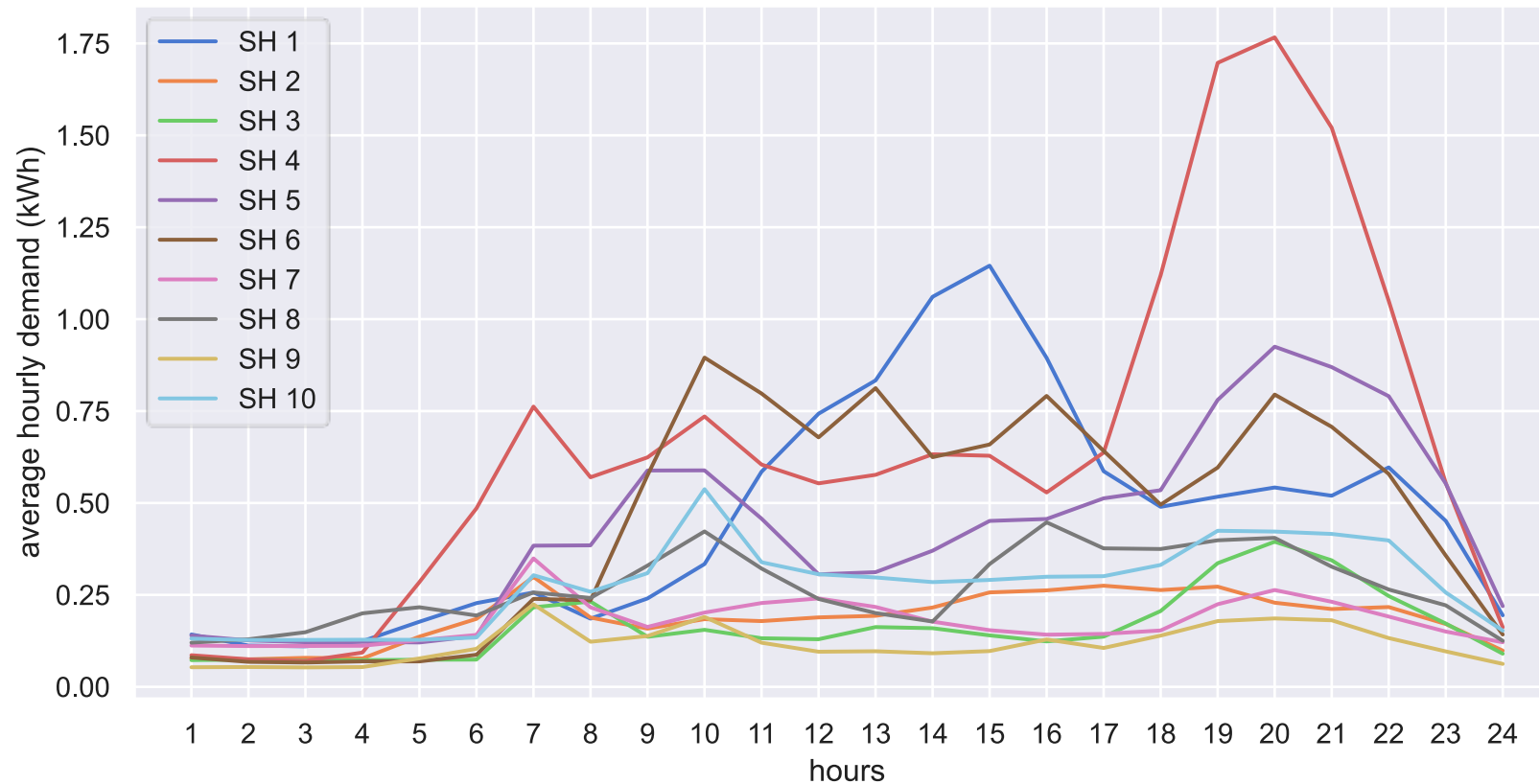
City – Average hourly electricity demand values



Suburban area – Average hourly electricity demand values



Rural area – Average hourly electricity demand values



Prosumers' data: City area

