





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

Theresia Perger TU Wien

Energy Economics Group (EEG) IEWT Vienna 2021, 08.09.-10.09.2021

□ perger@eeg.tuwien.ac.at







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: <u>Energy communities</u> and <u>Peer-to-Peer Trading</u>
- 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

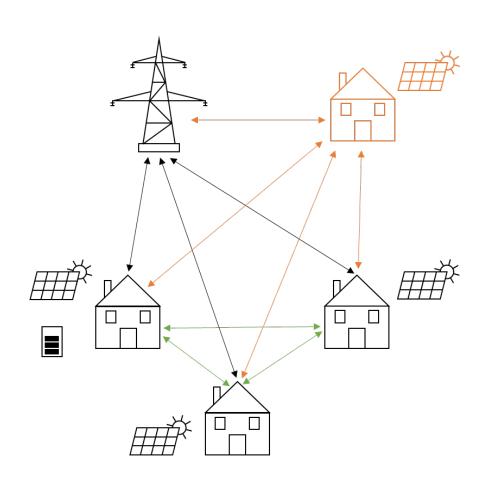
Research question:

 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)

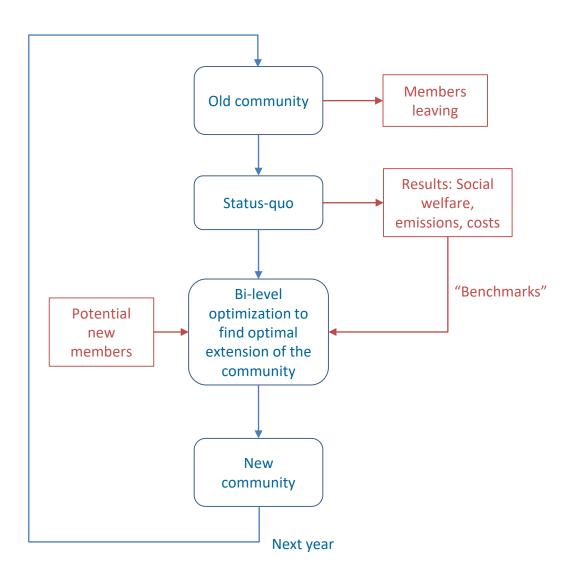
Contribution:

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



Modeling Approach





Social welfare:

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

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

$$\begin{aligned} & \min_{\{load_i, PV_i, b_i, Q_{i,t}\}} & \sum_{i \in \mathcal{I}_{old}} \alpha_i \Delta costs_i + (1 - \alpha_i) \Delta emissions_i \\ & \text{subject to:} \\ & b_i \cdot load_i^{min} \leq load_i \leq b_i \cdot load_i^{max} & \forall i \in \mathcal{I}_{new} \\ & b_i \cdot PV_i^{min} \leq PV_i \leq b_i \cdot PV_i^{max} & \forall i \in \mathcal{I}_{new} \\ & \sum_{i \in \mathcal{I}_{old}} b_i = n \end{aligned}$$



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 selfconsumption 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

$$\begin{aligned} & \max_{Q_{i,t}} \sum_{t \in \mathcal{T}, i \in \mathcal{I}} p_{t}^{G_{out}} q_{i,t}^{G_{out}} - \sum_{t \in \mathcal{T}, i \in \mathcal{I}} p_{t}^{G_{in}} q_{i,t}^{G_{in}} + \sum_{t \in \mathcal{T}, i, j \in \mathcal{I}} wtp_{i,j,t} q_{i,j,t}^{share} \\ & \text{subject to:} \\ & q_{i,t}^{G_{in}} + q_{i,t}^{B_{out}} + \sum_{j \in \mathcal{I}} q_{j,i,t}^{share} - q_{i,t}^{load} = 0 \qquad (\lambda_{i,t}^{load}) \quad \forall i \in \mathcal{I}_{old}, t \\ & q_{i,t}^{G_{out}} + q_{i,t}^{B_{in}} + \sum_{j \in \mathcal{I}} q_{i,j,t}^{share} - q_{i,t}^{PV} = 0 \qquad (\lambda_{i,t}^{PV}) \quad \forall i \in \mathcal{I}_{old}, t \\ & q_{i,t}^{G_{in}} + q_{i,t}^{B_{out}} + \sum_{j \in \mathcal{I}} q_{j,i,t}^{share} - load_i q_{i,t}^{load} = 0 \qquad (\lambda_{i,t}^{load}) \quad \forall i \in \mathcal{I}_{new}, t \\ & q_{i,t}^{G_{out}} + q_{i,t}^{B_{in}} + \sum_{j \in \mathcal{I}} q_{i,j,t}^{share} - PV_i q_{i,t}^{PV} = 0 \qquad (\lambda_{i,t}^{PV}) \quad \forall i \in \mathcal{I}_{new}, t \\ & SoC_{i,t-1} + q_{i,t}^{B_{in}} \cdot \gamma^B - q_{i,t}^{B_{out}} / \gamma^B - SoC_{i,t} = 0 \qquad (\lambda_{i,t}^{SoC}) \quad \forall i, t > t_0 \\ & SoC_{i,t=t_{end}} + q_{i,t_0}^{B_{in}} \cdot \gamma^B - q_{i,t_0}^{B_{out}} / \gamma^B - SoC_{i,t_0} = 0 \qquad (\lambda_{i,t_0}^{SoC}) \quad \forall i, t = t_0 \\ & SoC_{i,t} - SoC_{i}^{max} \leq 0 \qquad (\mu_{i,t}^{SoC^{max}}) \quad \forall i, t \\ & q_{i,t}^{B_{in}} - q_{i}^{B_{max}} \leq 0 \qquad (\mu_{i,t}^{B_{max}}) \quad \forall i, t \\ & q_{i,t}^{B_{out}} - q_{i}^{B_{max}} \leq 0 \qquad (\mu_{i,t}^{B_{out}}) \quad \forall i, t \end{aligned}$$



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

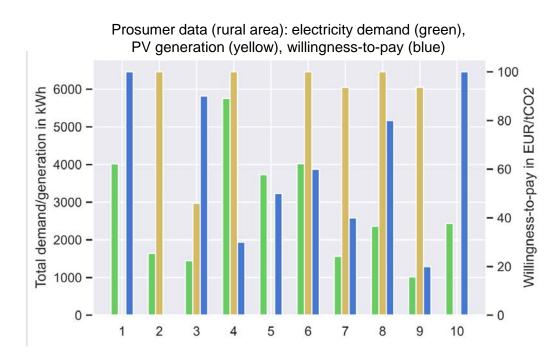


Modeling Approach – Data and assumptions



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 = $5 \, kW_{peak}$, $load = 39000 \frac{kWh}{vr}$
 - Single house: $PV = 3 kW_{peak}$, load = $1400 \frac{kWh}{}$



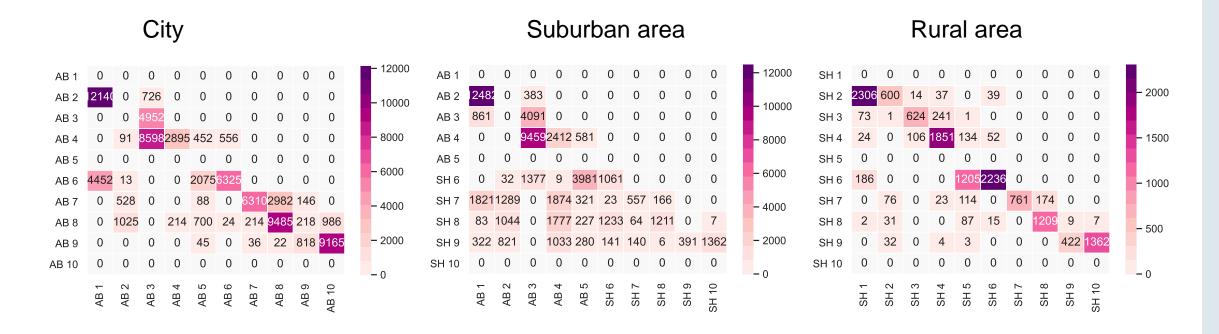


Results – Comparison of settlement patterns



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

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



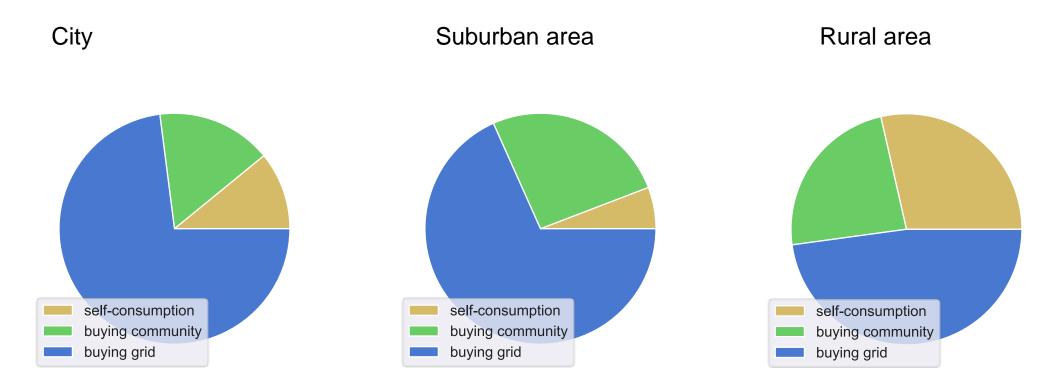


Results – Comparison of settlement patterns



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



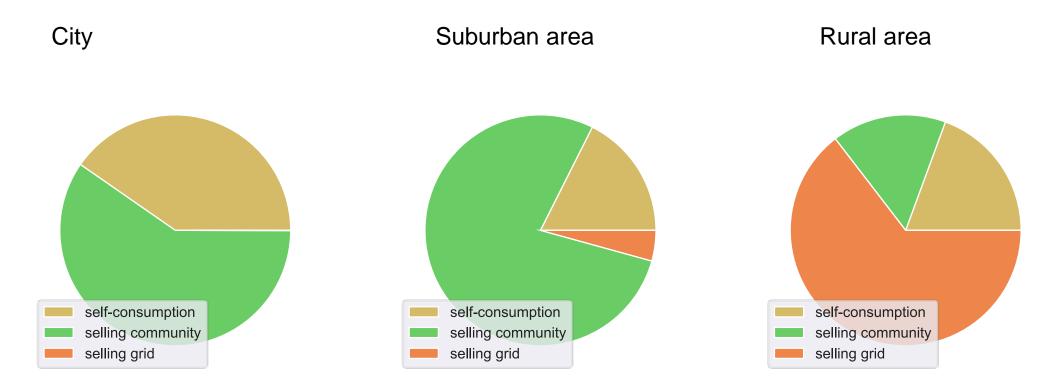


Results – Comparison of settlement patterns



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





Results - Rural area



All prosumers want to minimize their individual costs:

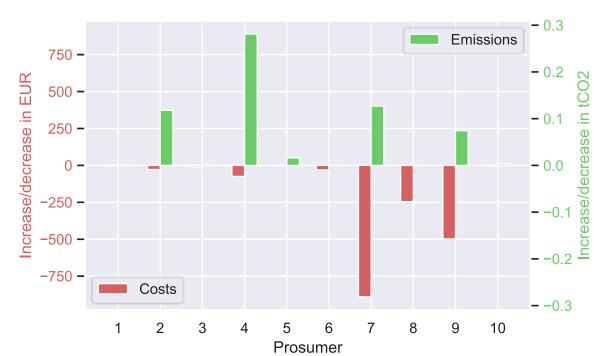
$$\alpha_i = 1, \forall i \in I_{old}$$

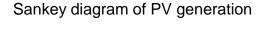
Results:

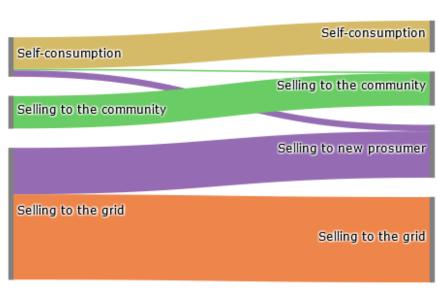
✓ Apartment building

✓ Opportunity to sell to the new prosumer (high demand, no PV installed) and lower

annual costs









Results – City area

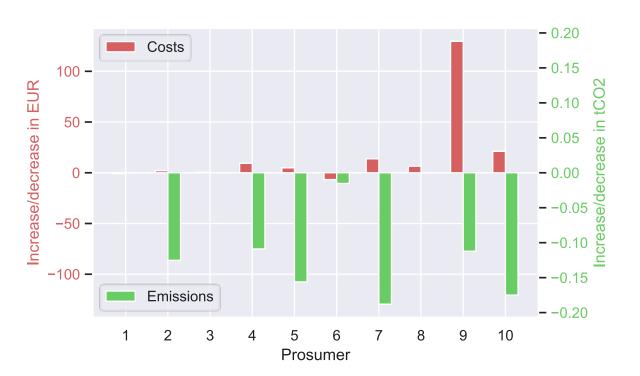


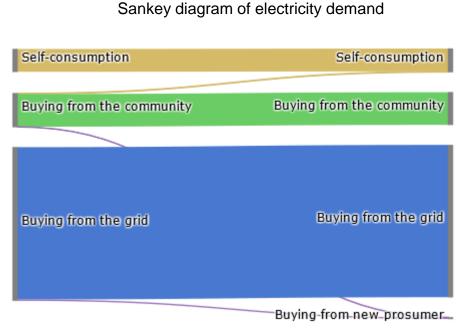
All prosumers want to minimize their individual emissions:

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

Results:

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







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!



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



open ENergy TRansition ANalyses for a low-Carbon Economy

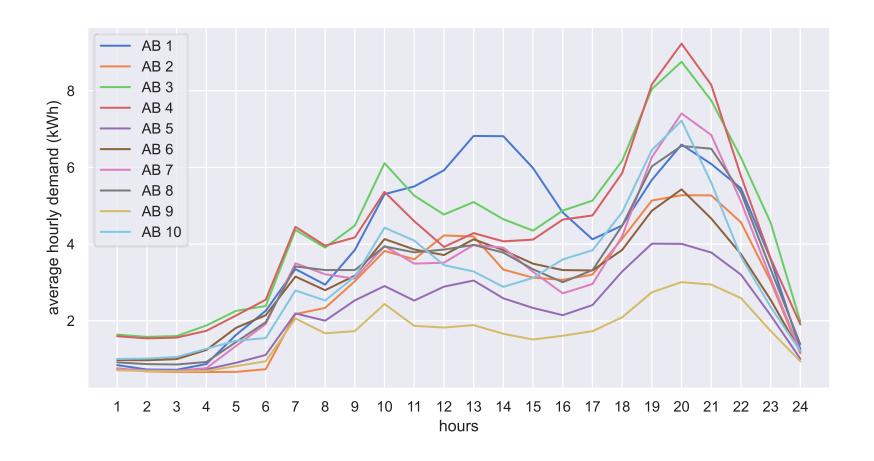
https://openentrance.eu/







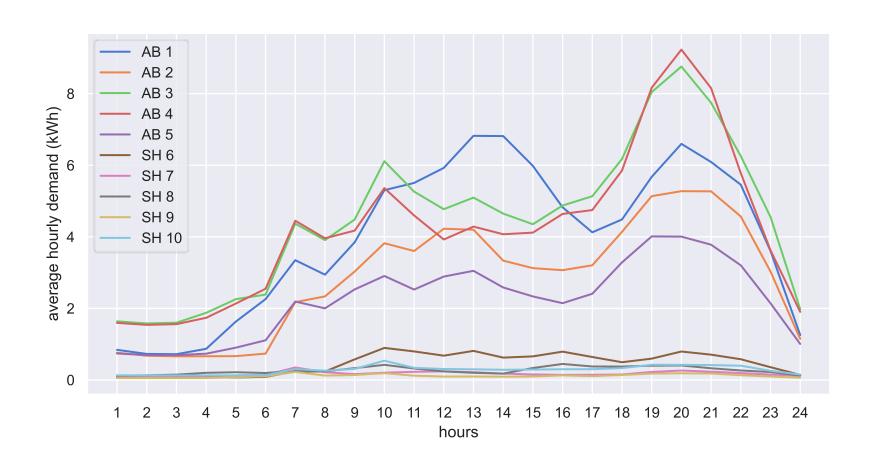
City – Average hourly electricity demand values







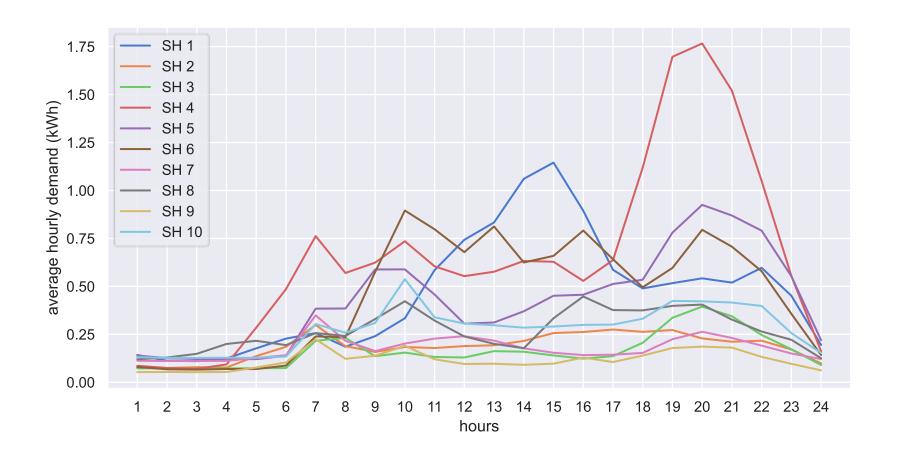
Suburban area – Average hourly electricity demand values







Rural area – Average hourly electricity demand values







Prosumers' data: City area

