



## Specifying Grid Development Needs in Sample Living Quarters

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# Agenda

## 1. Introduction

Approach and process

## 2. Methodology

Introduction to the proposed strategy

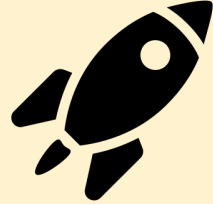
## 3. Insights gained

Results, sensitivity analysis, and discussion

## 4. Outlook

Conclusion and outlook

## Motivation



- The need of **creating a decision-making tool** dedicated to systematic and strategic regional grid development, with sample living quarters (“Musterquartiere”) as a conceptual framework, was identified.
- **Specific characteristics** are highlighted and used for **describing such “Musterquartiere.”**
- Availability of **big data** enables the use of new types of **(spatial) analysis.**



## Research Goals



1. Analyzing the **concept of “Musterquartiere”** as well as its **main characteristics**, and its suitability and applicability
2. Defining the **key requirements** for the **development of an algorithm** that can be used for clustering
3. Developing a tool for **Strategic investment planning** of grid operators

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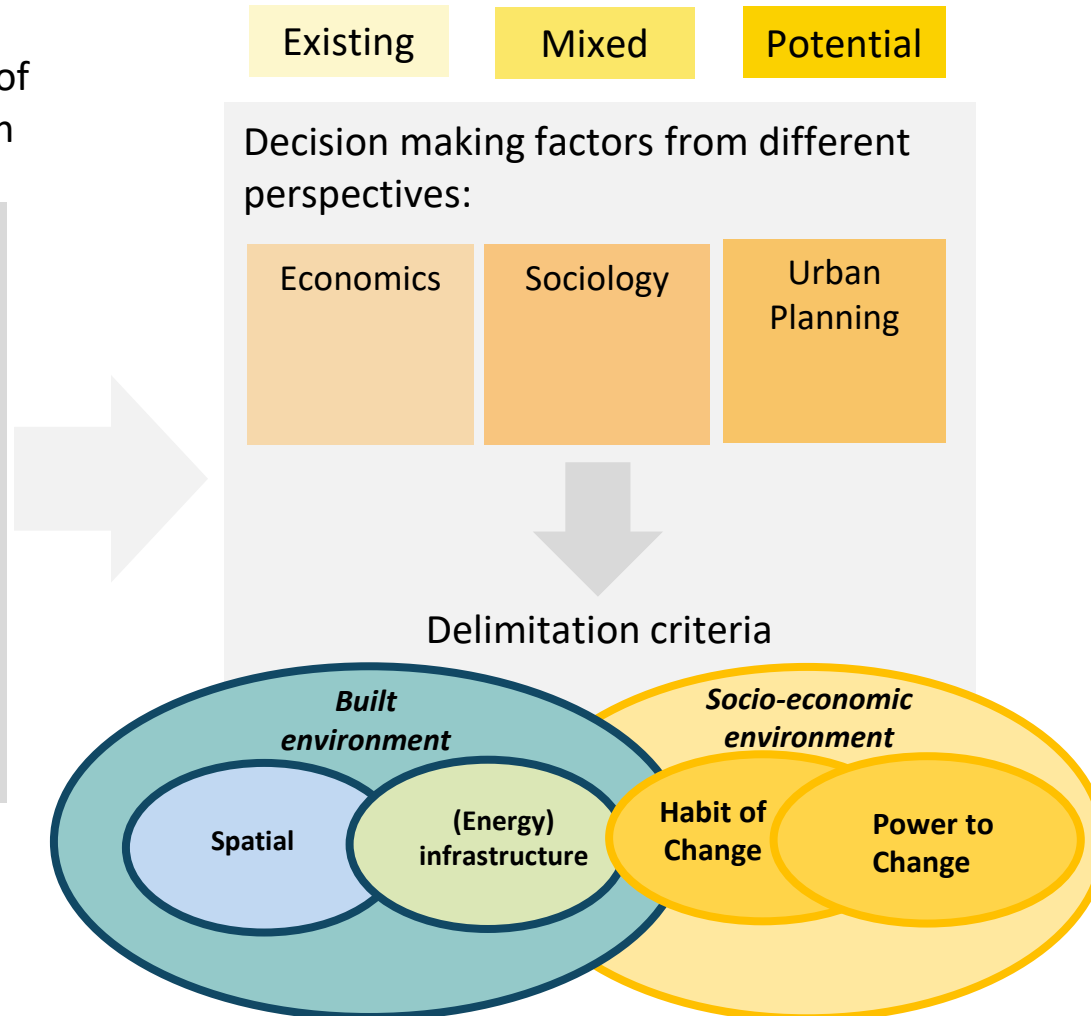
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# Frameworks for the Classification of Data

## Selected approach for the classification of data

**Available data:** 150 groups of data sets for each building in NRW, Germany

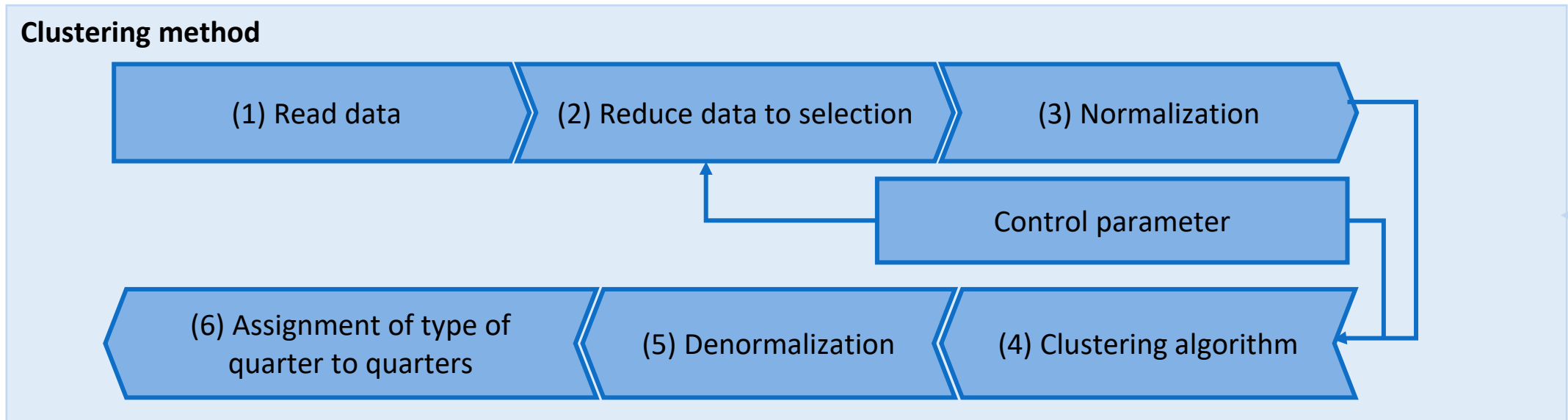
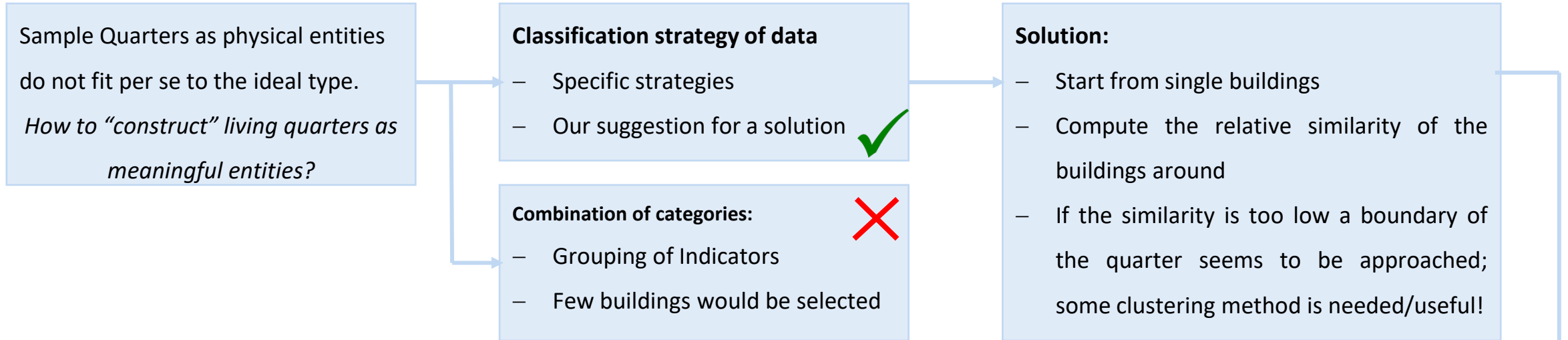
- General information of buildings
- Socio-economic data
- Energy data (e.g., electricity/heat demand, energy technologies)
- Building features (e.g., building use, location,)
- ....



## Framework for data classification:

1. Categorization of the given data if relevant for describing the existing situation, or essential for a possible potential, or both
2. Allocation of the available data to the scientific disciplines concerned to describe the decision-making factors
3. Elaboration of four main categories to map the current situation and the possibility of change

# Quarter Identification Strategy: Combining or Clustering?



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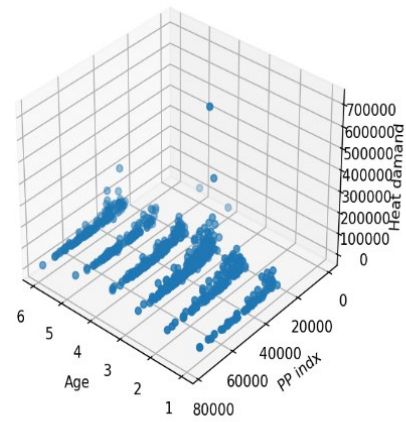
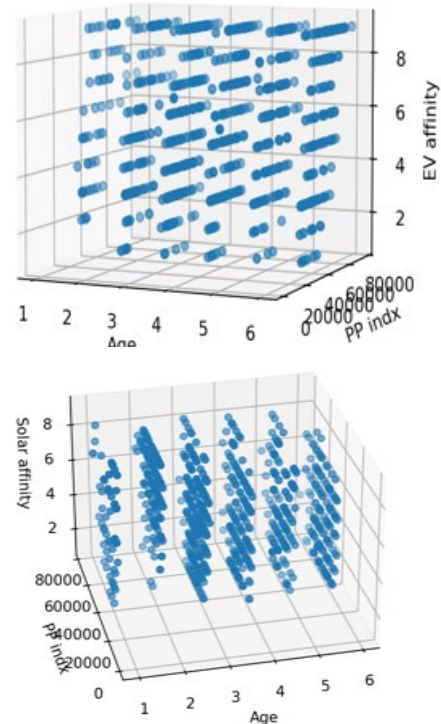
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# Evaluation of Selected Indicators

## Factor analysis:

- Solar affinity is impacted by age group.
- In contrast to the purchasing power, the age group does not influence heat demand.
- Factor analysis can be used for continuous indicators (e.g., density, heat demand, electricity demand, power purchase).



## Correlation matrix:

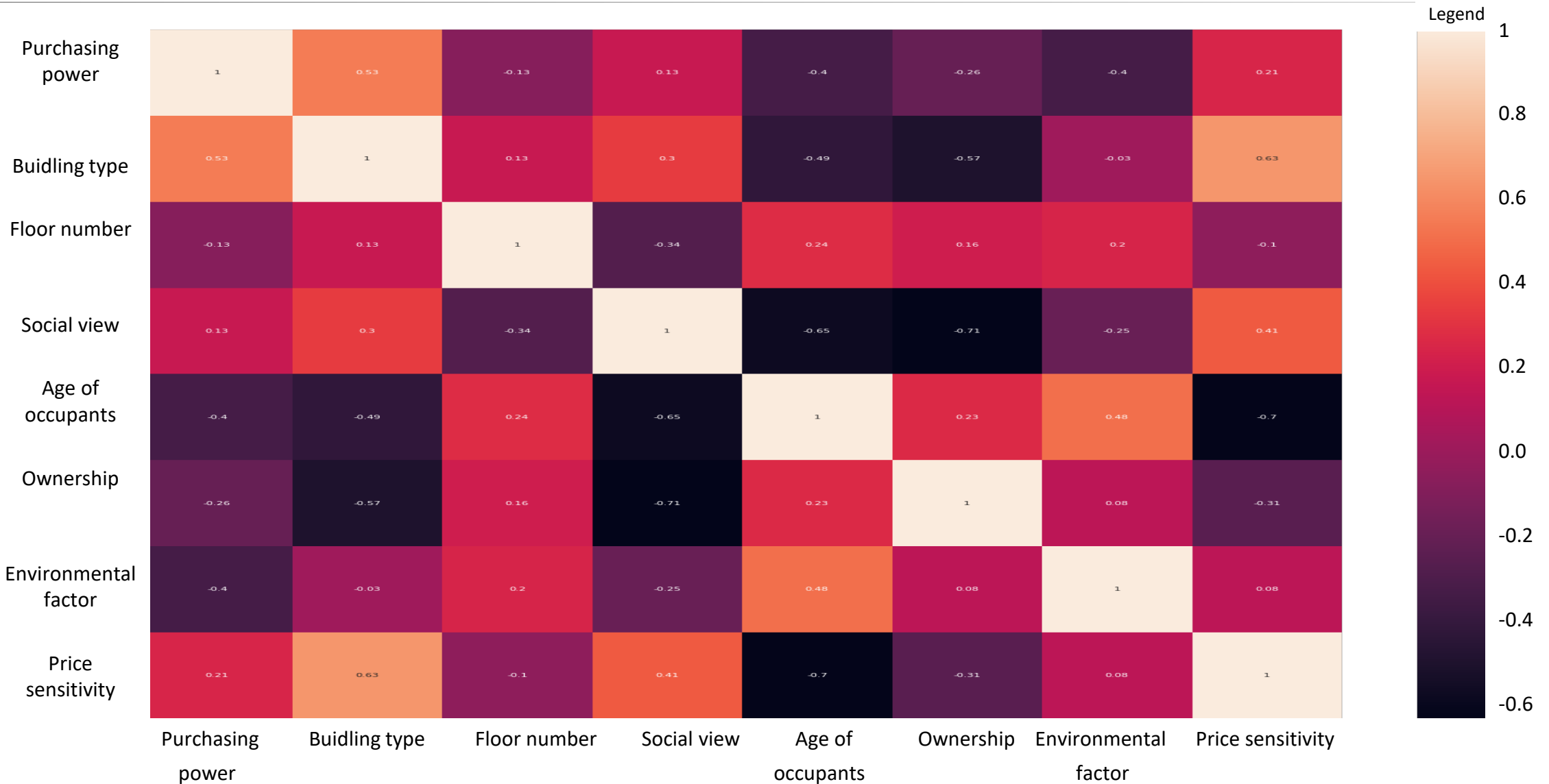
- Two elements of habit of change have a correlation of 0.08 with each other.
- Power to change: PP has a negative correlation with ownership and price sensitivity; ownership and age of inhabitant have correlation of 0.2.



Figure: Thermal image of socio-economic properties



# Evaluation of Selected Indicators



# Indicators



Which building properties are significant for the categorization of a living quarter?

Which social factors of people in a living quarter are significant?

## Spatial

1



Location

2



Characteristics

3



Age of building / Construction

## Socio-Economic

1



Purchasing power

2



Age of occupants

3



Ownership

4



Price sensitivity

5



Environmental concern

Power to change

Habit of Change

## Energy

1



Solar affinity

2



E-Mobility affinity

3



EV stations

4



No. of garages

5



Solar capacity

6



Heating system

# Clustering Strategy: Denormalization of Outputs

- How to elaborate results quantitatively?

<b>Socio-Economic</b>	Very low power to change Very low habit of change	
<b>Spatial</b>	Local Site, multi family houses	
<b>Energy</b>	NOW	FUTURE DEVELOPMENT
PV	Very low	Very low
EV	Very low	Very low
Heat options	Floor heating, central heating	Heat pump

Sample

- Output of each indicator is allocated dynamically into five groups:

very low | low | mid | high | very high

**Power to change:** {Age of inhabitant}\*{Purchasing power}

**Habit of change:** {Price sensitivity}\*{Environmental concern}\*{Home ownership}

**PV NOW:** Present PV installed capacity

**EV NOW:** Present No. of EVs

**PV Future:** {PV affinity}\*{Power to change}\*{Habit of Change}

**EV Future:** {EV affinity}\*{Power to change}\*{Habit of Change}\*{No. Of Garages}

**Heat options Now:** Two technologies with the highest shares

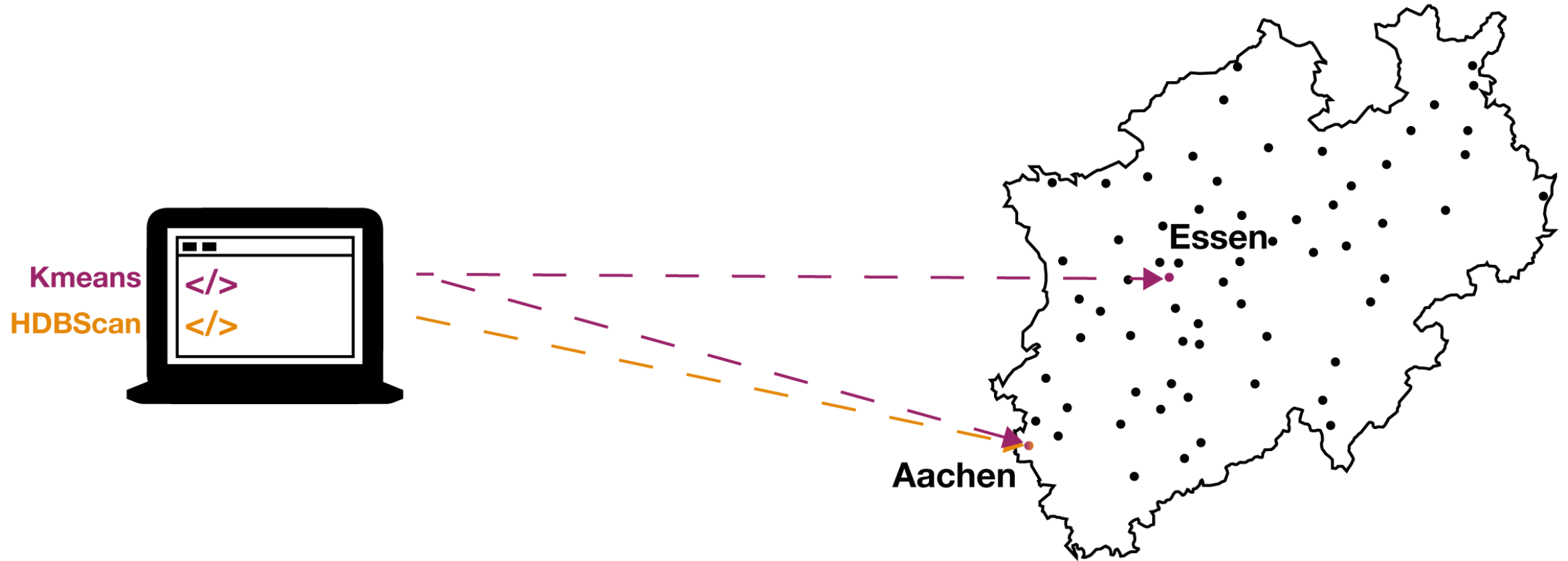
**Heat options Future\*:**

If { District heating share > 5% → District heating  
District heating share < 5% → Heat pump }

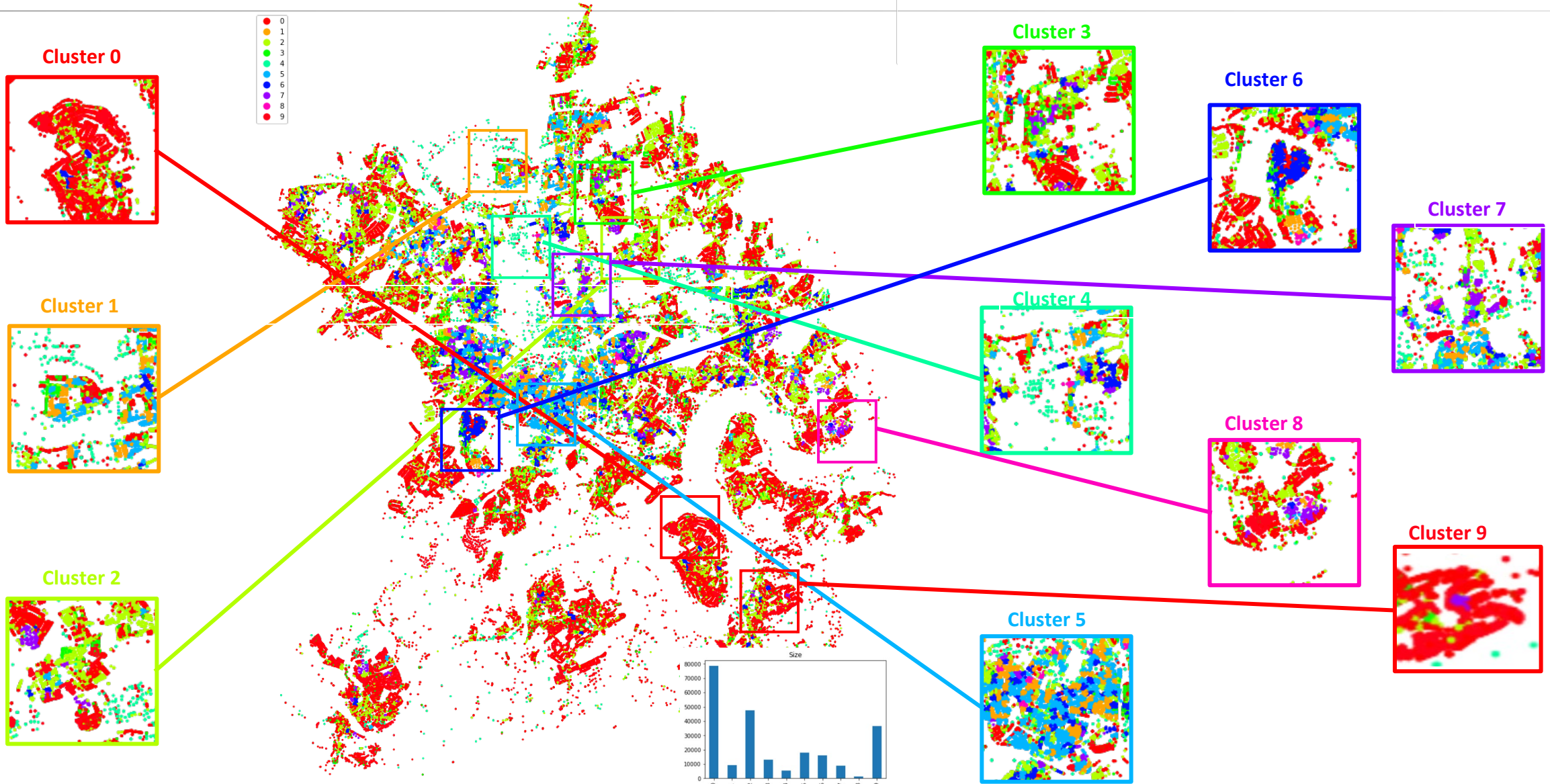
\*Based on scenarios presented in E.ON Energiewende im Wärmesektor

# Results

- Implementing one algorithm on two cities (Essen, Aachen)

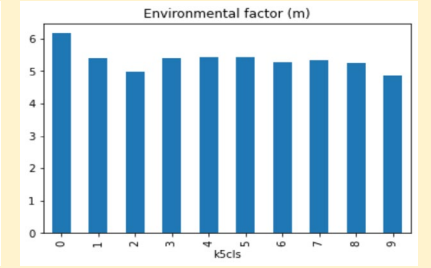
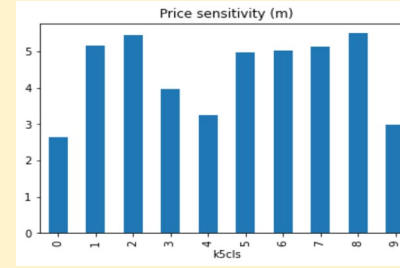
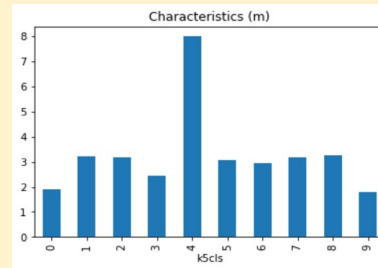
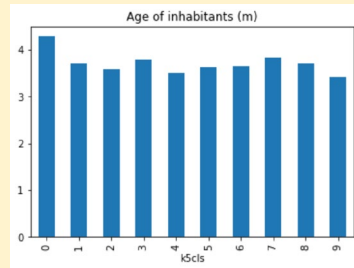
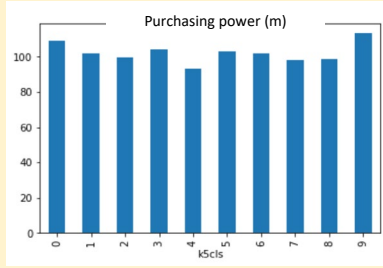


# K-means Clustering: Essen

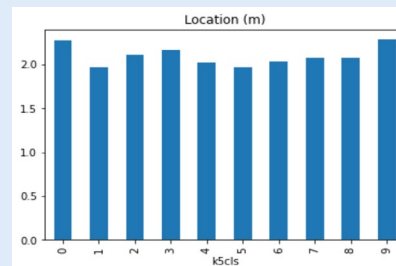
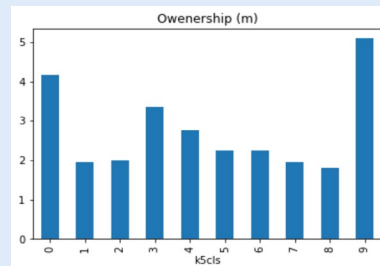
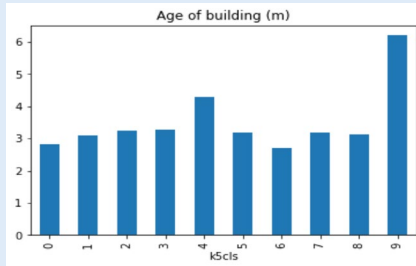


# Statistical Analysis of K-means Method: Essen

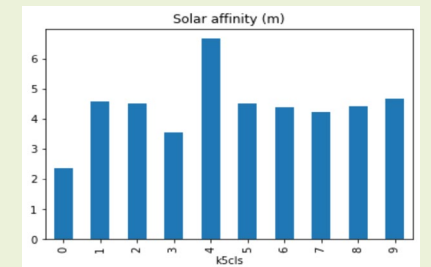
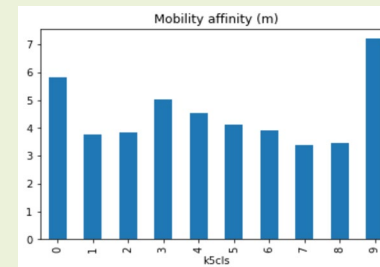
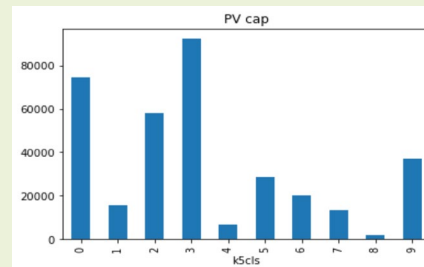
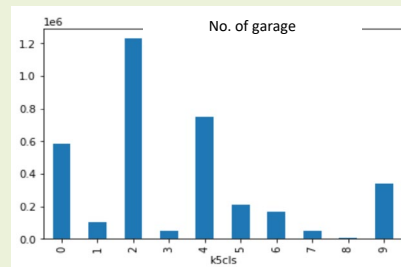
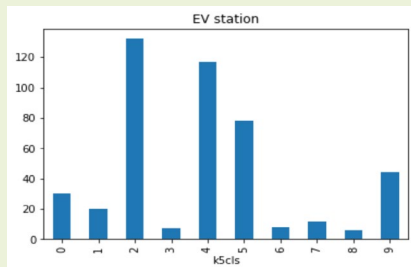
## Socio-Economic



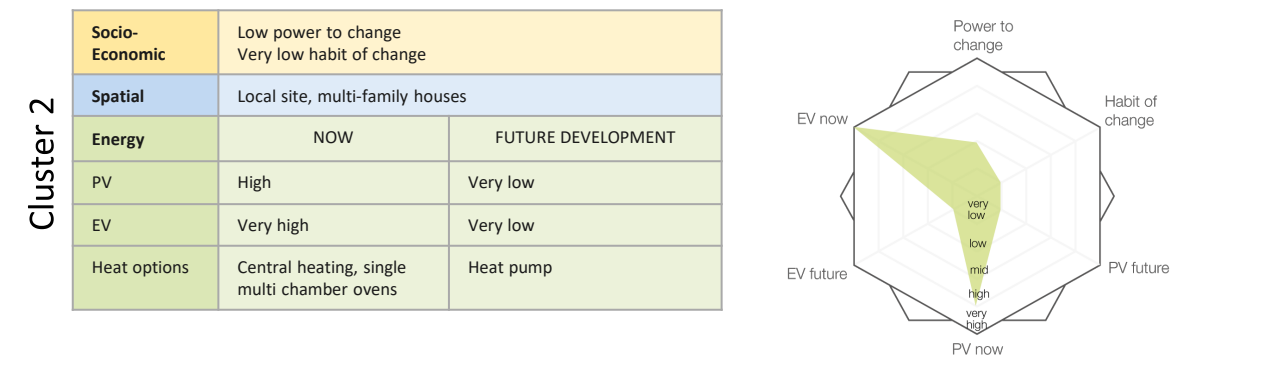
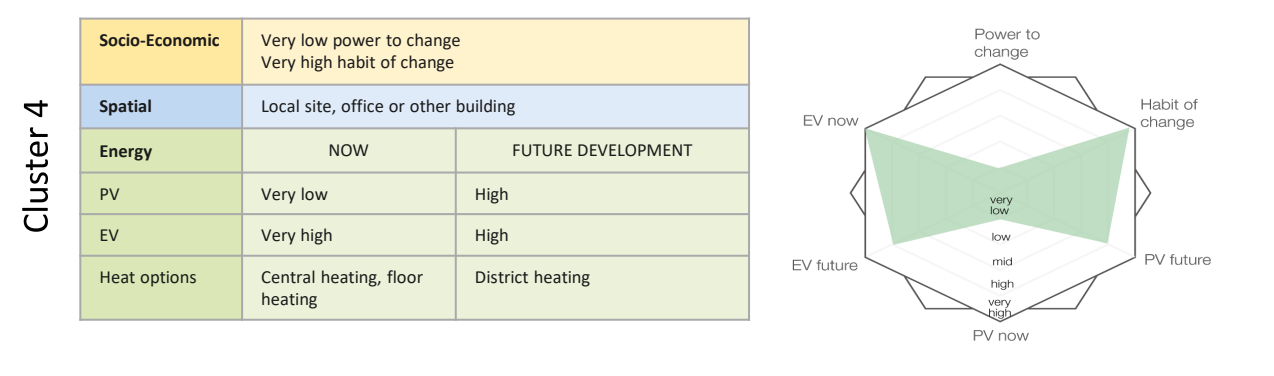
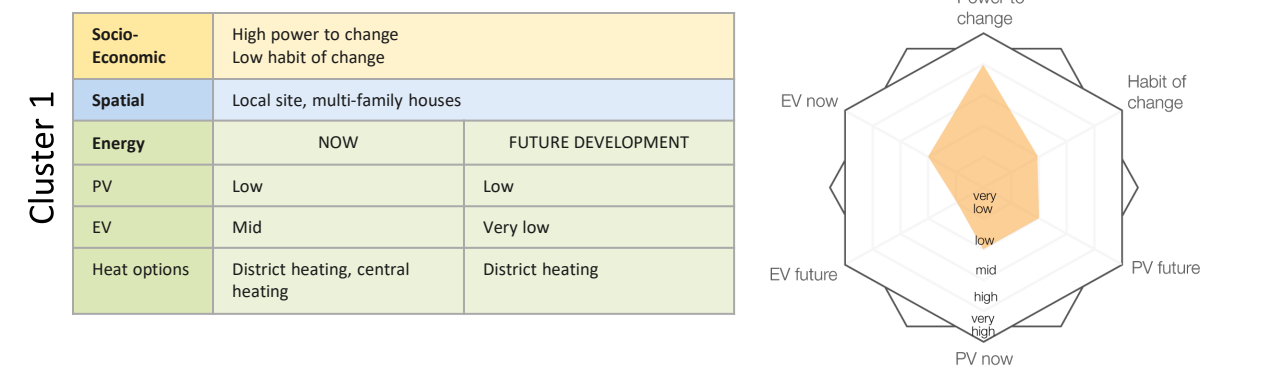
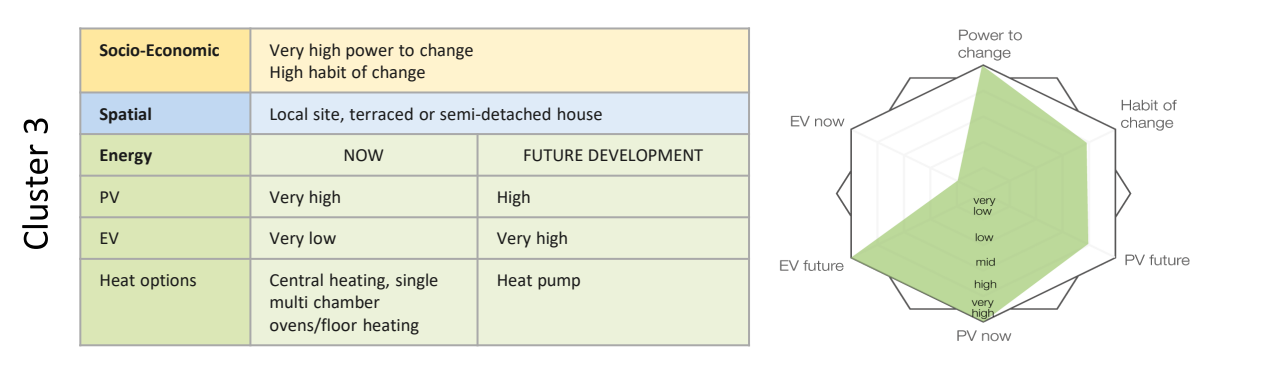
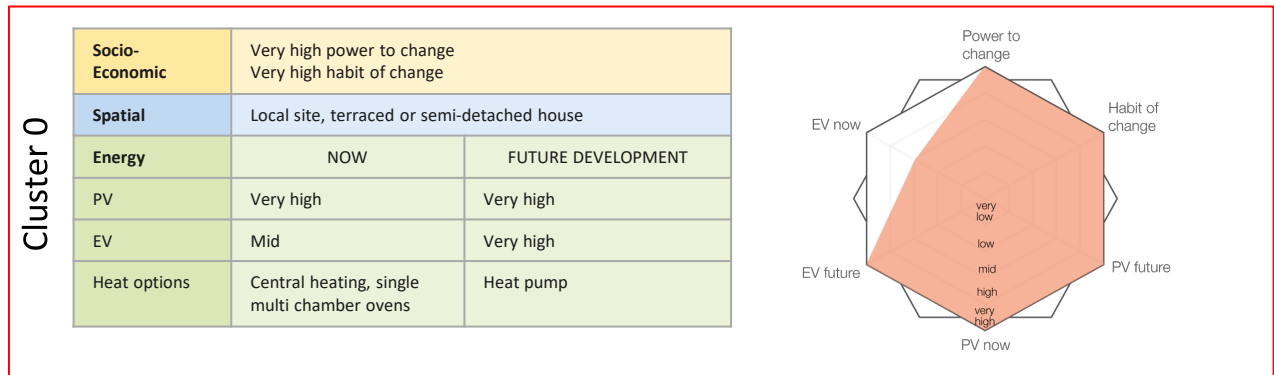
## Spatial



## Energy



# Cluster Profile: Clusters 0-4 (K-means)

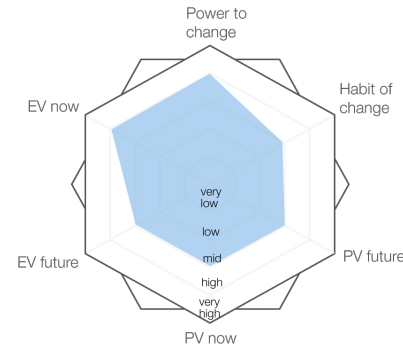




# Cluster Profile: Clusters 5-9 (K-means)

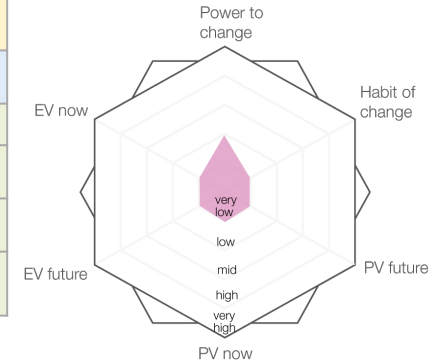
Cluster 5

<b>Socio-Economic</b>	High power to change Mid habit of change	
<b>Spatial</b>	Local site, multi-family houses	
<b>Energy</b>	NOW	FUTURE DEVELOPMENT
PV	Mid	Mid
EV	High	Mid
Heat options	Central heating, district heating	District heating



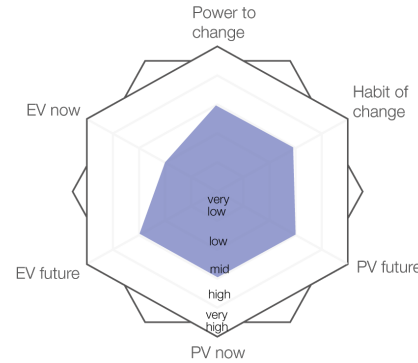
Cluster 8

<b>Socio-Economic</b>	Low power to change Very low habit of change	
<b>Spatial</b>	Local site, multi-family houses	
<b>Energy</b>	NOW	FUTURE DEVELOPMENT
PV	Very low	Very low
EV	Very low	Very low
Heat options	District heating, central heating	District hating



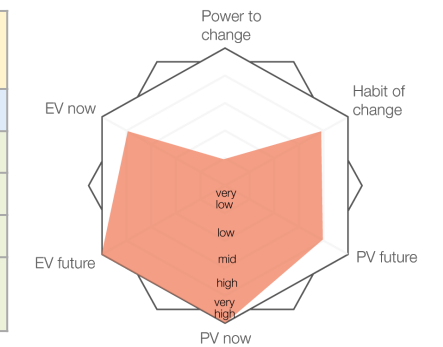
Cluster 6

<b>Socio-Economic</b>	High power to change Mid habit of change	
<b>Spatial</b>	Local site, multi-family houses	
<b>Energy</b>	NOW	FUTURE DEVELOPMENT
PV	Mid	Mid
EV	High	Mid
Heat options	Central heating, district heating	District heating



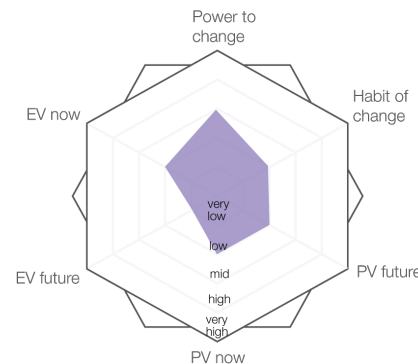
Cluster 9

<b>Socio-Economic</b>	Very low power to change High habit of change	
<b>Spatial</b>	Local site, terraced or semi-detached house	
<b>Energy</b>	NOW	FUTURE DEVELOPMENT
PV	High	Very high
EV	High	Very high
Heat options	Central heating, single multi chamber ovens	Heat pump



Cluster 7

<b>Socio-Economic</b>	Mid power to change Low habit of change	
<b>Spatial</b>	Local site, multi-family houses	
<b>Energy</b>	NOW	FUTURE DEVELOPMENT
PV	Low	Low
EV	Low	Very low
Heat options	Single multi chamber ovens, central heating	Heat pump

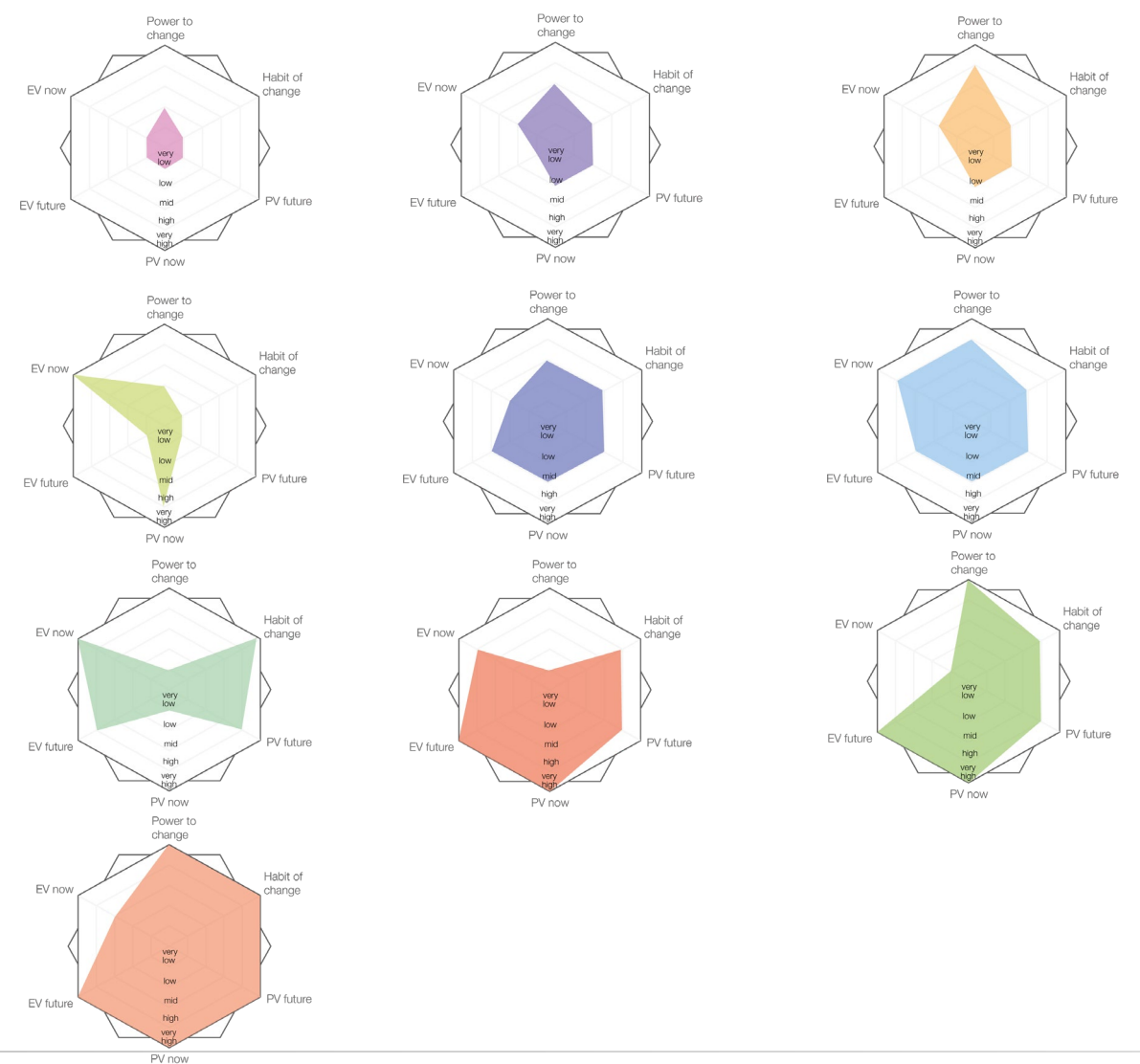
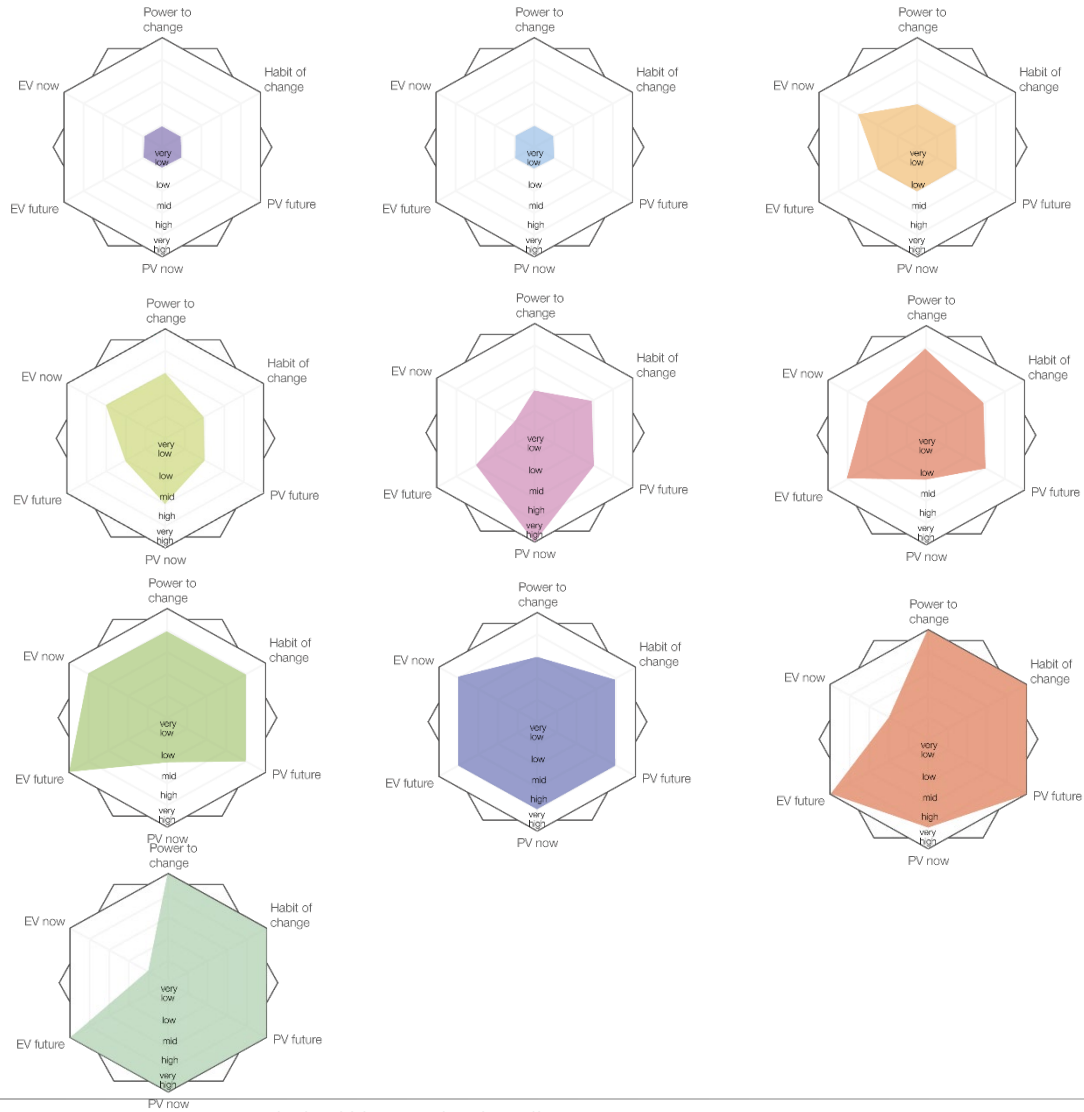




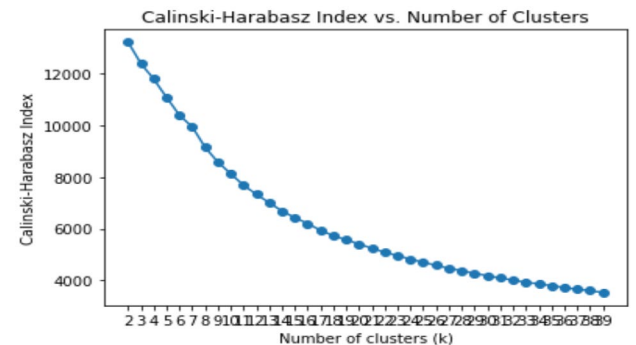
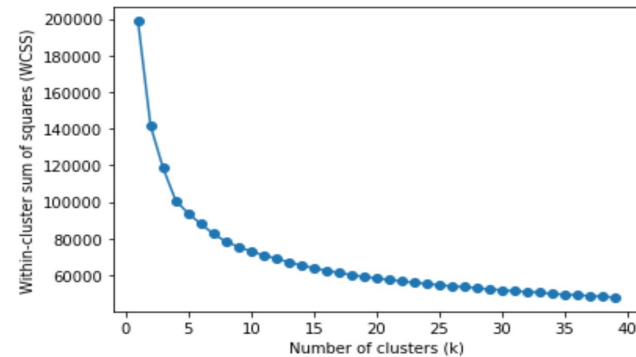
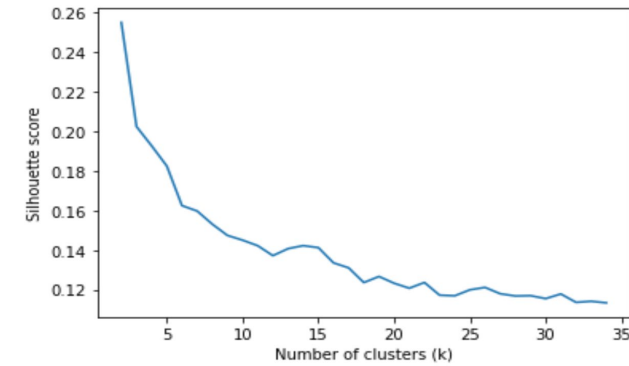
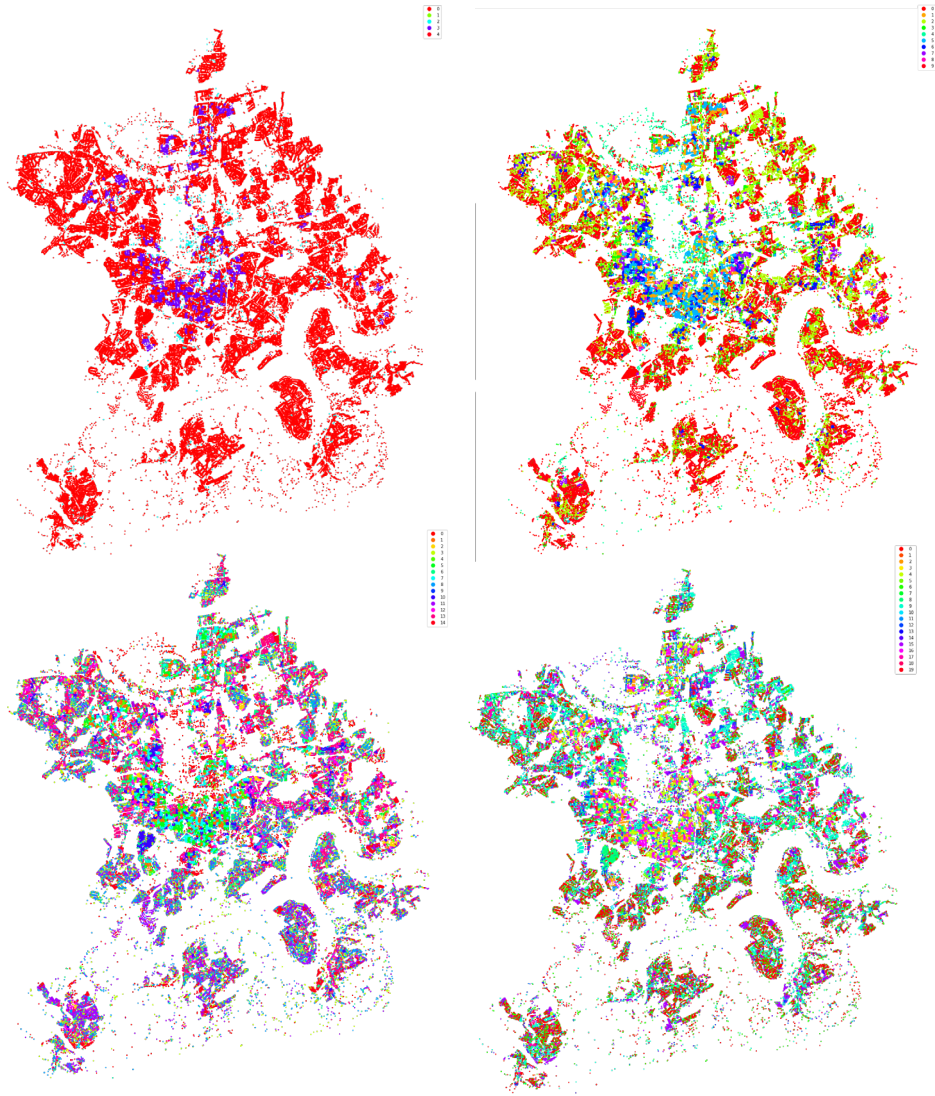
# Sensitivity Analysis: Comparing Clusters of Aachen and Essen

## Aachen

## Essen



# Sensitivity Analysis: K-means Method



## Sensitivity analysis approaches:

- ❖ Cluster numbers between 10 and 20 seem a good choice.
- ❖ Impact of an increasing number of clusters on the quality of the outputs:
  - Increases the noise (outliers) in the outputs.
  - Creates subclusters that can have trivial differences with other clusters.
  - Lower visualized quality
  - Worse statistical index (e.g., WCSS, Silhouette score)

# Discussion of key results

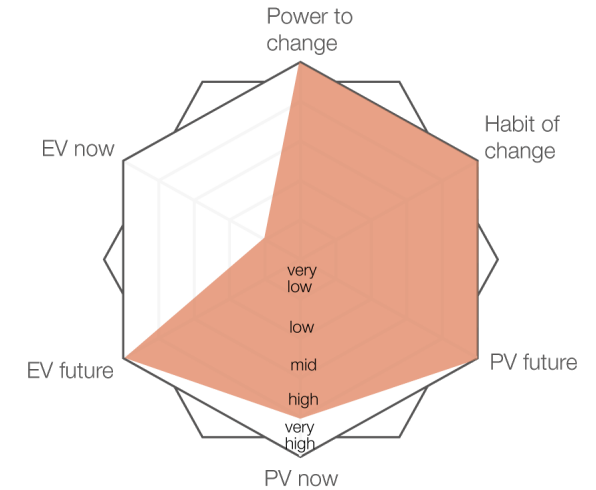
- High ownership
- Low price sensitivity
- Highest environmental concern
- Few EV stations
- High E-mobility affinity
- Medium solar capacity

- Future development on the part of residents high
- Inhabitants do have the power and the means to create change
- Change potential
- Provide concentrated energy (EV, PV)

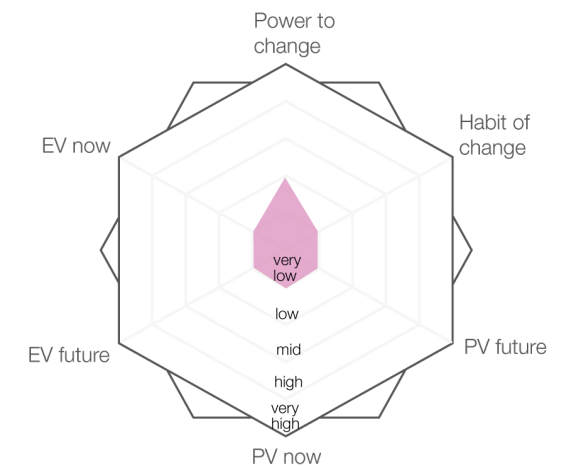
- Low ownership
- Highest price sensitivity
- Medium environmental concern
- No EV stations
- Lower E-mobility affinity
- Very low solar capacity

- Future development on the part of residents low
- Inhabitants do not have the power and the means to create change
- Other actors (like the landlord) or institutions (like the city municipality) could create change in this area

<b>Socio-Economic</b>	Very high power to change Very high habit of change	
<b>Spatial</b>	Local site, Terraced or semi-detached house	
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# Conclusion

## Conclusion:

- The optimal number of clusters depends on several parameters: indicators, clustering algorithm, aims of clustering, number of key outputs.
- There is **no single optimal number** of clusters in all cities/regions. However, the proposed heuristics can help us to find the optimal number of clusters for any region.
- We determined energy characteristics of clusters for the **current** and the **future** status.
- We found **10** meaningful clusters for city of Essen, Germany.
- For EVs, solar PV, heating systems, socio-economic factors and building types, the proposed **clustering algorithm** works.

## Outlook:

- Exploring contrasting situations like rural areas or ones that are of a suburban structure.
- Investigate whether clusters can also be described with existing planning instruments (such as the land use plan and regional planning).
- Carbon emissions and mitigation costs analysis.
- **Scalability** of the algorithm: Using other clustering algorithms (e.g., HDBSCAN)
- Explicit consideration of spatial dependencies (e.g., **contiguity**).





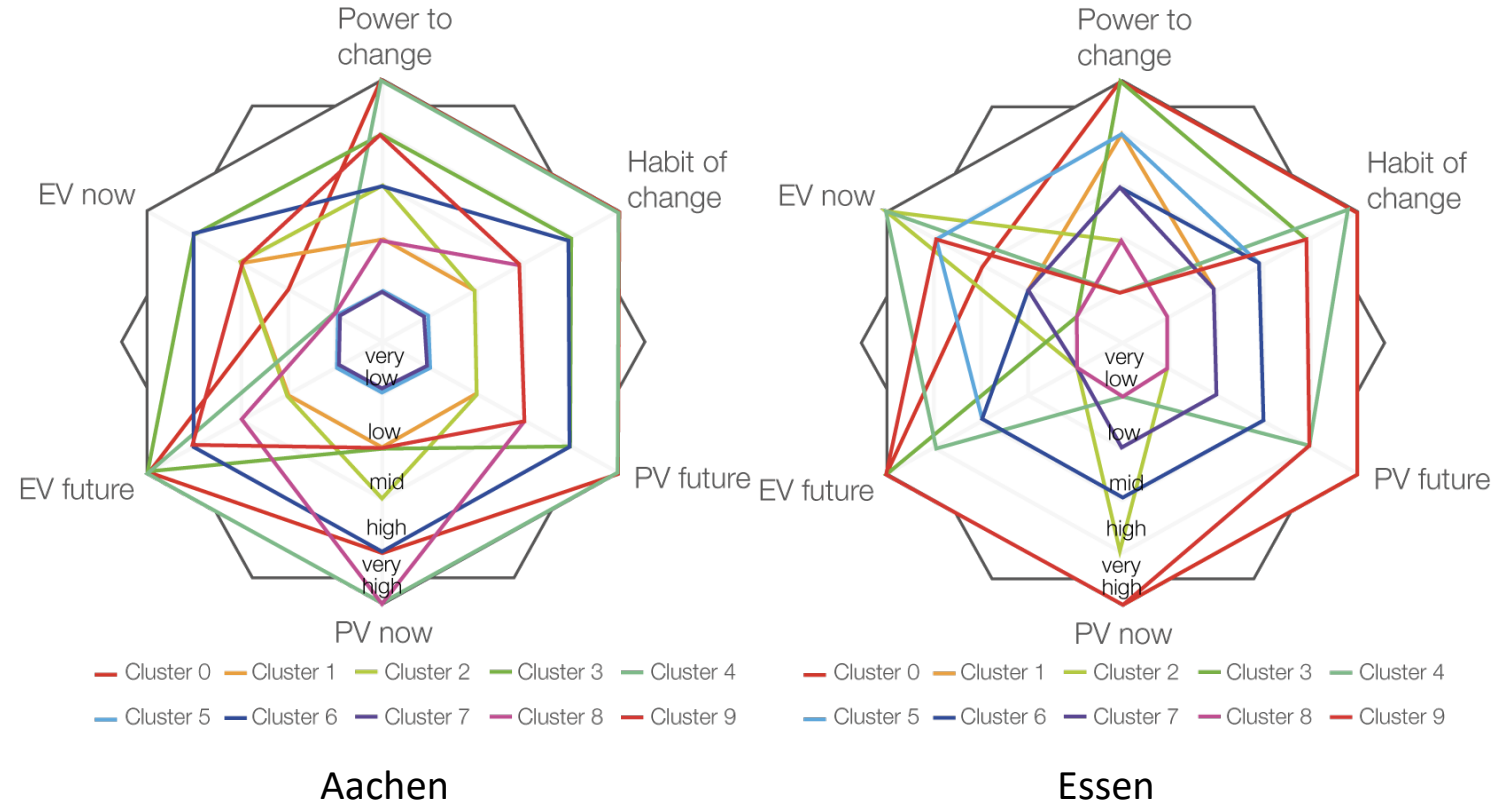
**Thank you for your attention!**  
**Questions?**

**Contact**

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# Summary of Clusters Profile: Aachen vs. Essen



- Differences can be attributed to different types of features in cities. For example, Essen has a higher EV potential than Aachen. However, both have a relatively similar EV future.
- In Essen, there is a positive correlation between EV and PV potential for the future in clusters.
- In clusters where the share of multi-family houses is high, the expected future solar PV and e-vehicle (EV) potential is low.
- Terraced or semi-detached houses has the highest PV and EV potential.