



House of  
**Energy Markets  
& Finance**

# Identifying key elements for adequate simplifications of investment choices – The case of wind energy expansion

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ESSEN**

*Open-Minded*

- ❖ A social planner seeks to identify possible and profitable sites and turbine technologies to invest in under a given economical and technological setting.
- ❖ A modeler aims to support the social planner. She therefore aims to accurately model wind energy expansion under given constraints while using limited computational and time resources.
- Leads to the challenge of minimizing the error of aggregation in electricity market modeling

## I. Definition of value components

## II. Computation of value components

1. Choice of adequate scenarios
2. Define investment choices as objects for the clustering algorithm
3. Calculate value components for various scenarios

## III. Aggregation of investment choices in limited number of clusters

4. Predefine cluster numbers using hierarchical clustering
5. Aggregation of investment choices

## Definition of value components

- Key parameter in decision-making process
- Considers multiple spatial and technological factors
- Sum of the contribution margin over time subtracted by the investment cost
- Interpreted as site (area) and unit (technology) specific excess profit per installed capacity

### Parameters

$e^{-rt}$	discounting factor
$\theta_{t,a,i}$	capacity factor
$K_{a,i}$	capacity
$p_t$	electricity price
$c_i$	investment cost

### Indices

$t$	time
$a$	area
$i$	technology

$$NPV_{a,i} = \sum_t e^{-rt} \theta_{t,a,i} K_{a,i} p_{t,a} - c_i K_{a,i}$$

Per-unit consideration:

$$Value_{a,i} = \sum_t e^{-rt} \theta_{t,a,i} p_t - c_i$$

## Definition of value components

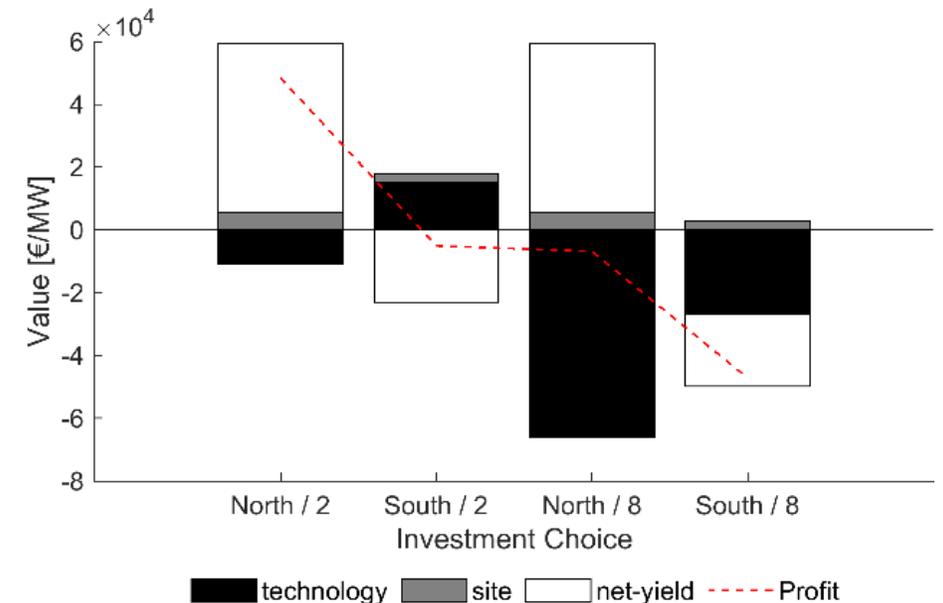
- **Yield specific value component**
  - driven by the **site-specific full load hours** (at average technology mix)
- **Resource specific value component**
  - driven by the **market value factor** of wind
  - self destructive effect of high RE shares
- **Site specific value component** (spatial heterogeneity)
  - driven by **wind profile** of the selected site
  - site vs. overall averaged profile per time
- **Technology specific value component**
  - driven by the **selected turbine type**
  - individual vs. portfolio averaged profile
- **Grid specific value component**
  - considering the **network load** and resulting nodal price differences.

For further analysis  
summed up as  
**“net yield”**  
component

# Exemplary value components (VC) for a future portfolio

## Definition of value components

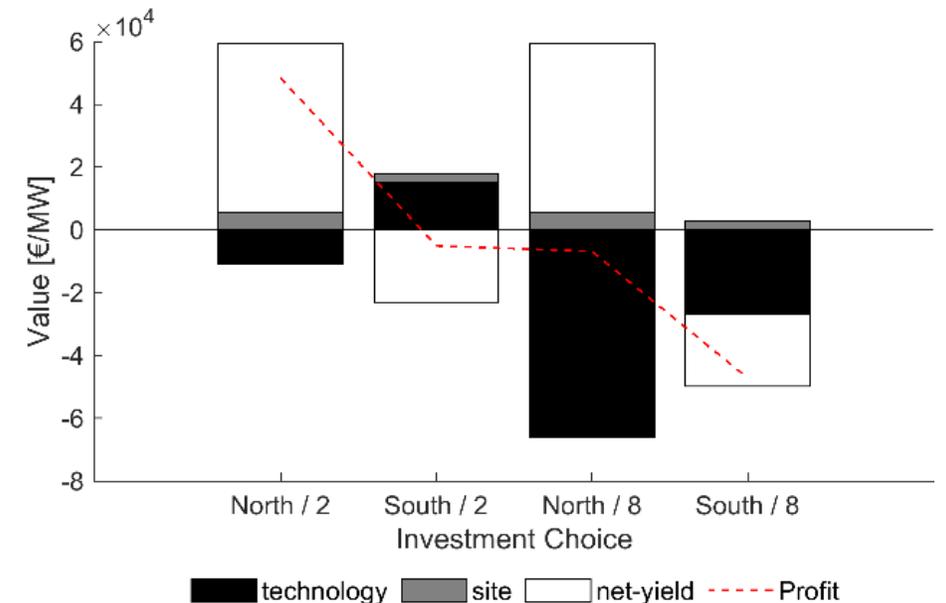
- VC based on the result of an optimized future scenario
- Four investment choices (IC ) as combinations of site and technology
  - sites: DE132 → South Germany  
DEF02 → North Germany
  - technologies: Turbine 2 → Onshore low speed  
Turbine 8 → Onshore high speed
- Net yield component positive in the north, negative in the south, reflective of FLH independent from technology
- Positive effect of the site component even at the exemplary southern site
- Highest variation in the technology component
  - Onshore high speed turbine is not beneficial at the Northern site
  - the additional revenues do not compensate the additional costs compared to the average portfolio (cf. also next slide)



# Exemplary value components (VC) for a future portfolio

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Note: In the optimized future portfolio

- If at one site no technology has positive profit, the site remains unused
- If there is at least one profitable technology at one site, the technology with the highest profit is built up to the site capacity limit

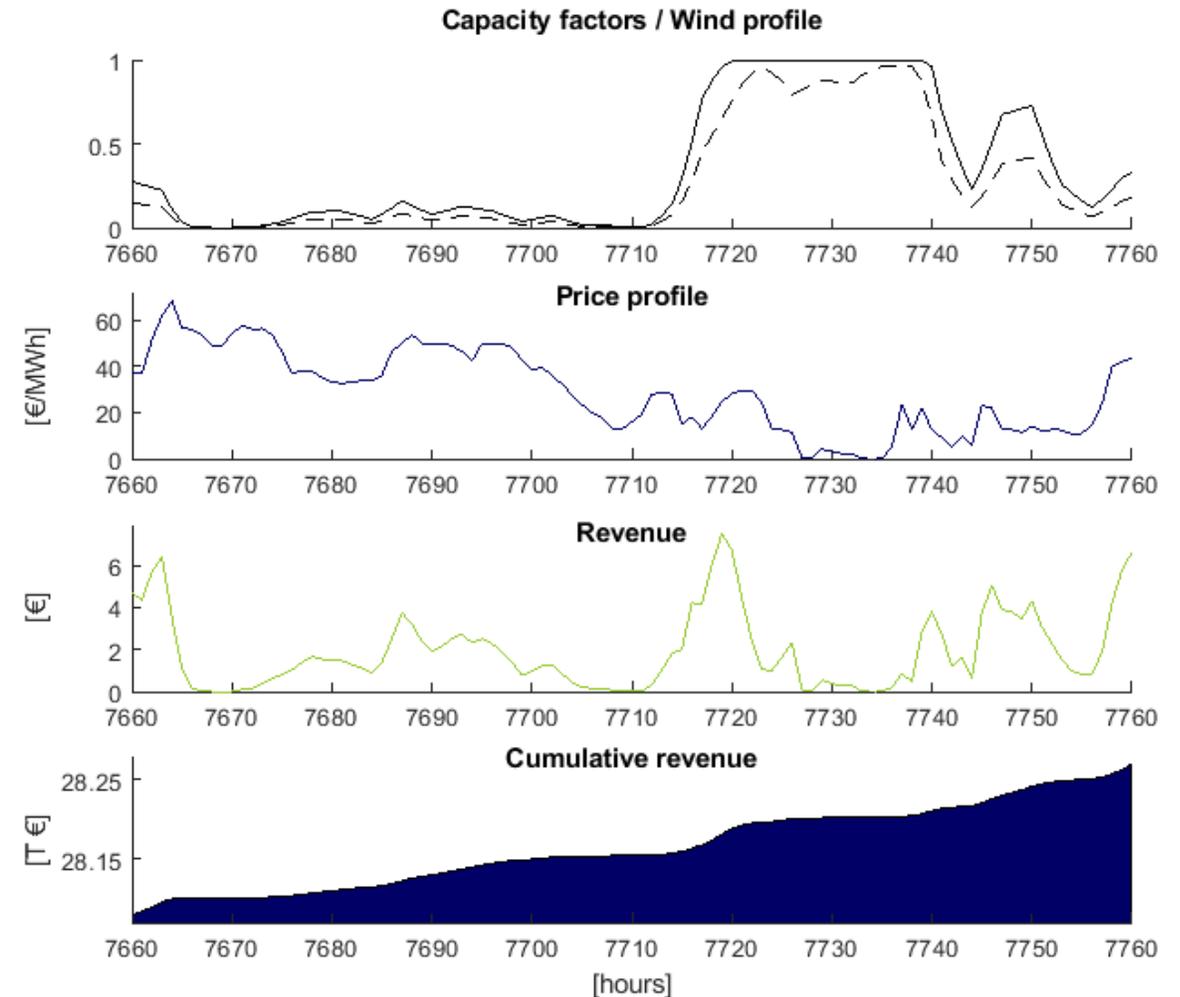
# Exemplary analysis of the technology component

## Definition of value components

- VC based 2017 scenario
- Investment choice South (DE132) / WEA 2

$$VC_{a,i}^{technology} = \sum_t \underbrace{(\theta_{t,a,i} - \bar{\theta}_{t,a,\cdot})}_{\text{Difference in capacity factors}} p_t - \underbrace{(c_i - c.)}_{\text{Cost difference}}$$

- Cumulated yearly
  - revenue 33,191.00 €
  - cost difference 39,806.00 €
  - technology component -6,615.00 €



# Exemplary analysis of the technology component

## Definition of value components

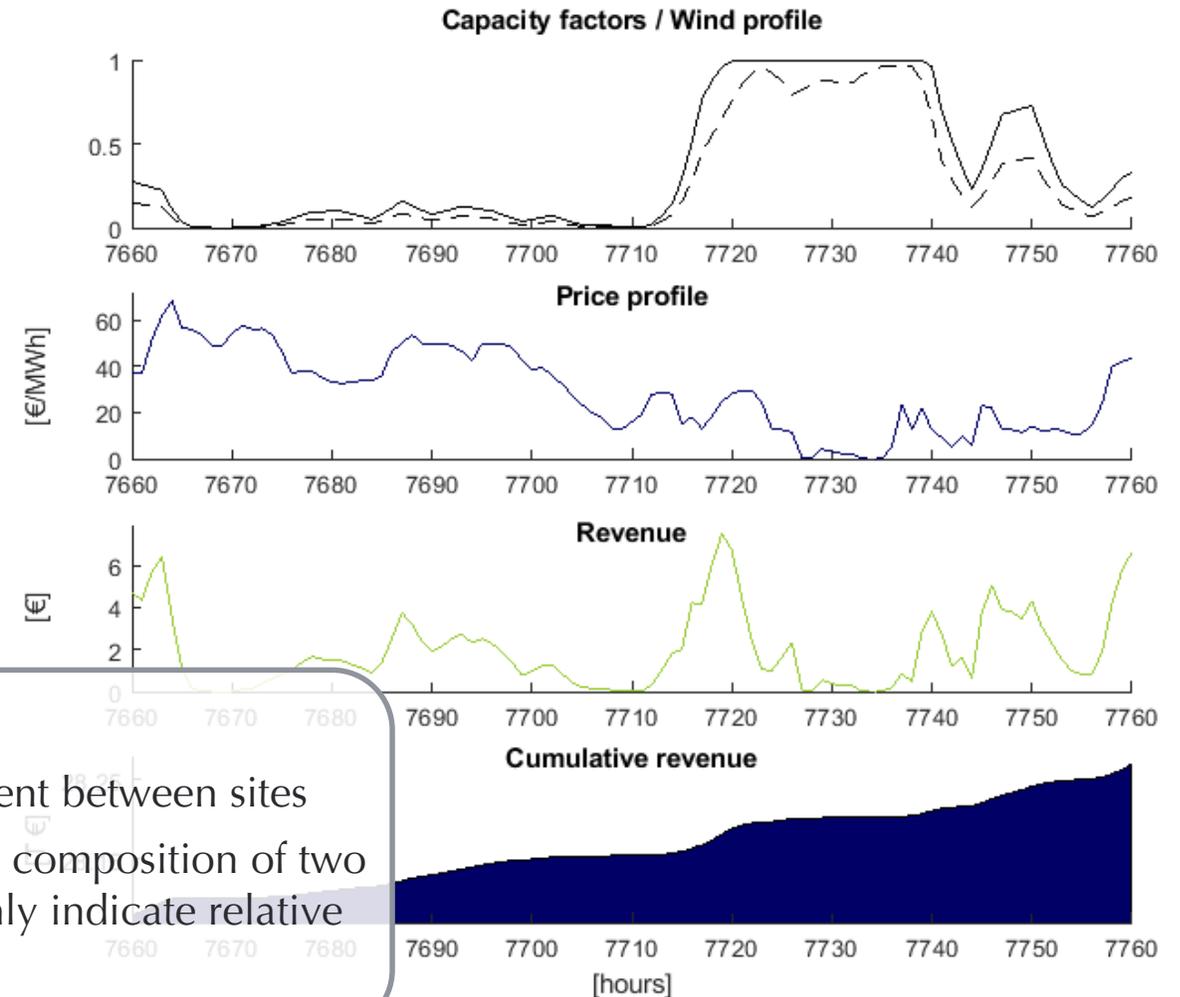
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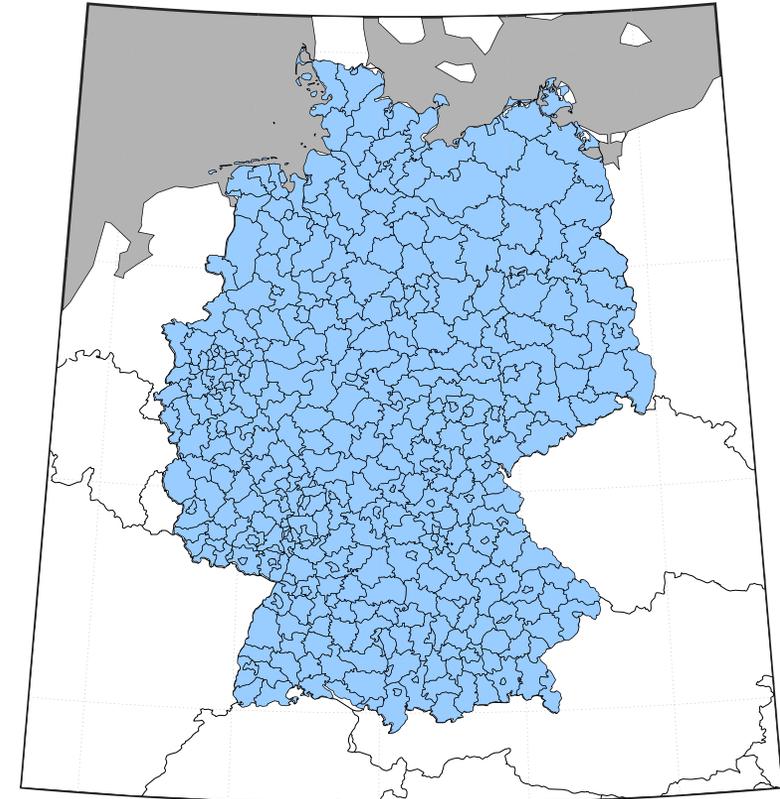
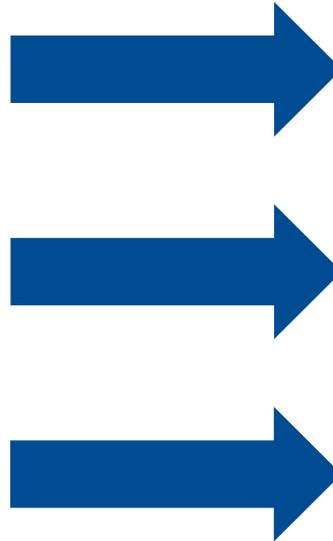
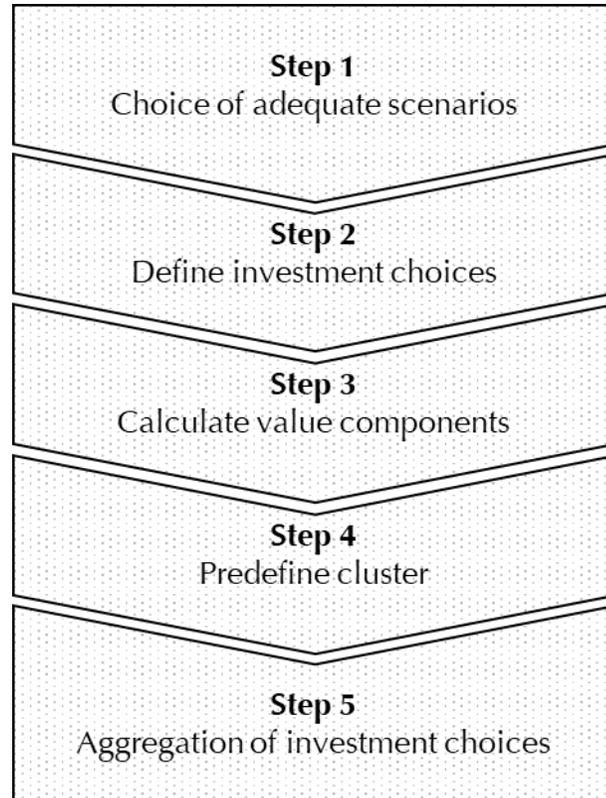
- Cumulated yearly
  - revenue 33,191.00 €
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Note: For the technology component

- the average capacity factor is different between sites
- hence, the comparison of the value composition of two different investment choices can only indicate relative differences



# The case of wind energy in Germany



# 1. Choice of adequate scenarios

The case of wind energy in Germany

- Scenario A or “2017”
  - real capacities of wind power 2017
  - historical day-ahead spot prices
- Scenario B or “2020 nodal”
  - ensure the consideration of congestion effects
  - scaled capacities of 2017
  - approximated prices for 2020 adopted from Felling and Weber (2018)
- Scenario C or “future”
  - result of a simplified electricity market model optimization
  - greenfield approach with one conventional backup technology and three renewable sources
  - low-emission scenario with a renewable share of about 65 %
  - demand held constant from 2017

## 2. Define investment choices

The case of wind energy in Germany

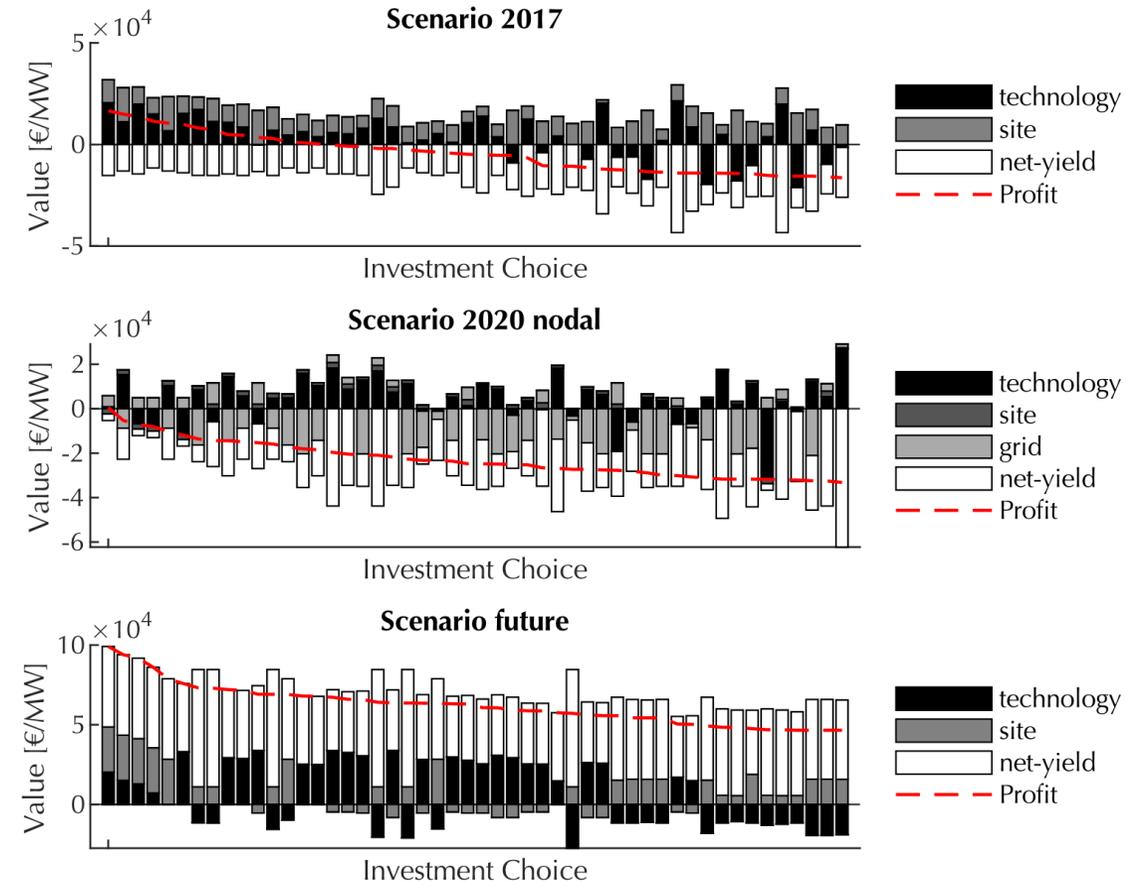
- Combination of site and technology type:
  - spatial resolution: 402 NUTS3 regions in Germany
  - level of technological detail: 8 wind turbines representing the variety of wind turbines in Germany
- 402 sites and 8 technology types lead to an overall of 3216 investment choices (cluster objects)

Turbine Type	Hub height [m]	Rotor Diameter [m]	Power [kW]	Type	Capex [€/kW]
WEA 1	72	53	800	High speed	1.047
WEA 2	139	121	2.530	Low speed	1.571
WEA 3	109	92	2.350	High speed	1.155
WEA 4	142	114	3.170	Low speed	1.290
WEA 5	110	109	3.000	Low speed	1.169
WEA 6	150	140	4.000	Low speed	1.573
WEA 7	120	124	4.500	High speed	1.363
WEA 8	120	140	6.000	High speed	1.483

# 3. Calculate value components for three scenarios

The case of wind energy in Germany

- Shown results: 50 most profitable investment choices (ICs) of each scenario in descending order
- Observations:
  - Number of ICs with positive profit: 15 - 1 - 245
  - Number of ICs with a strictly positive realized capacity: 991 - 927 – 52
  - Strong changes in profitable investment choices and in capacities greater zero between 2017/2020 nodal and future
    - fundamental difference prices and capacities
  - reduction in overall profitability from 2017 to 2020 nodal
    - price influence of grid congestion
  - high profitability in future
    - positive contributions of the net-yield component

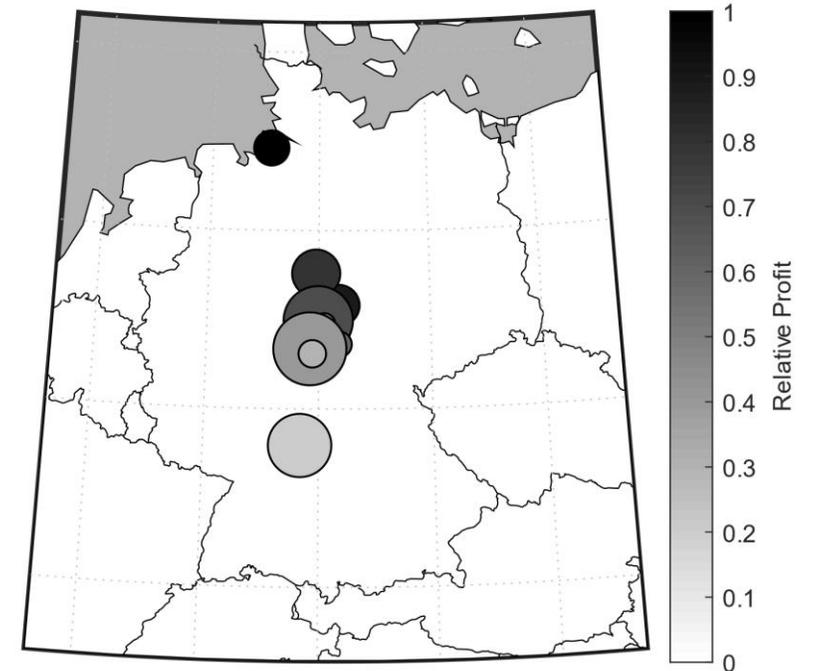
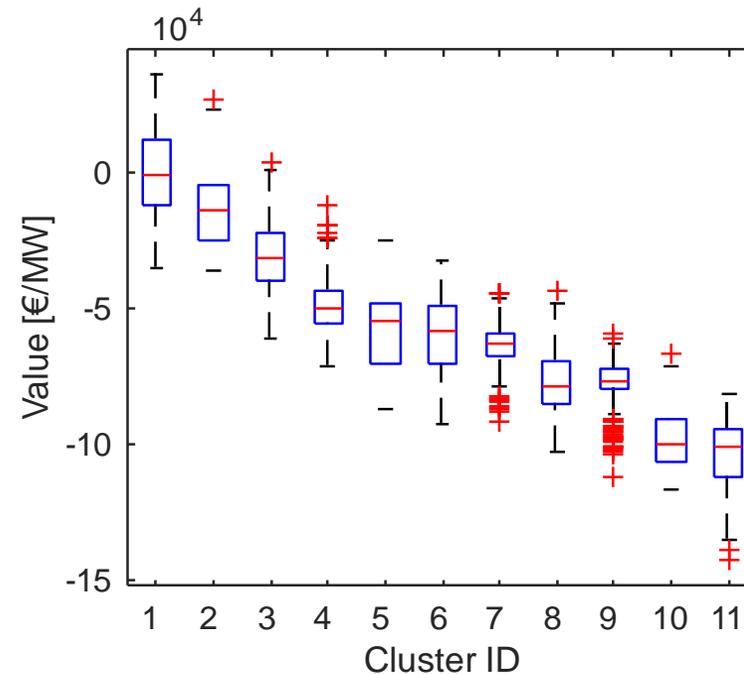


The case of wind energy in Germany

- Using:
  - k-means clustering
  - predefined number of clusters using hierarchical clustering
  - squared Euclidean Distance
- Done for:
  - 402 sites (NUTS3 regions)
  - 8 technologies
  - 3216 investment choices
- Based on:
  - 4 to 5 value components in
  - 3 scenarios → indicating that aggregation of decision alternatives is robust under different scenario settings
  - 13 attributes
  - 11 clusters

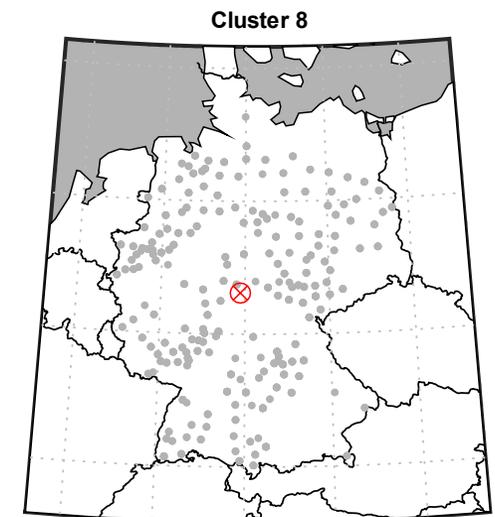
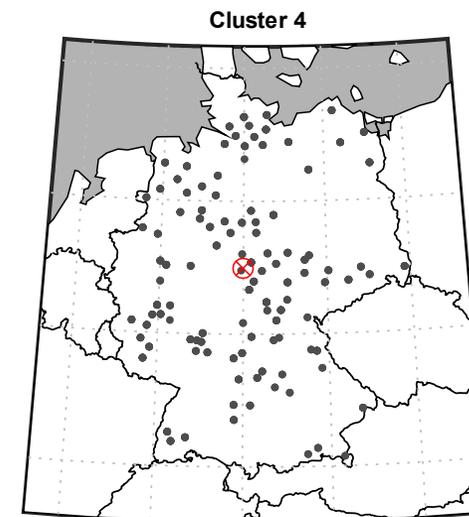
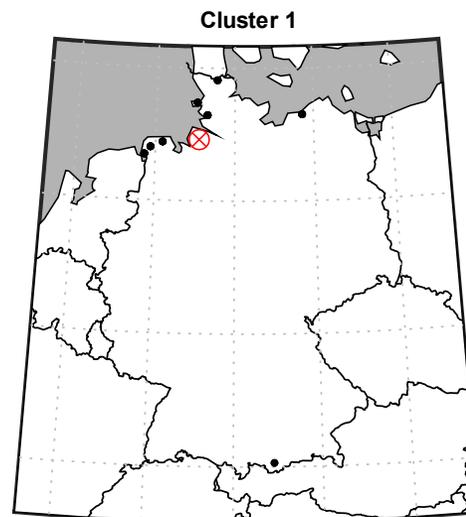
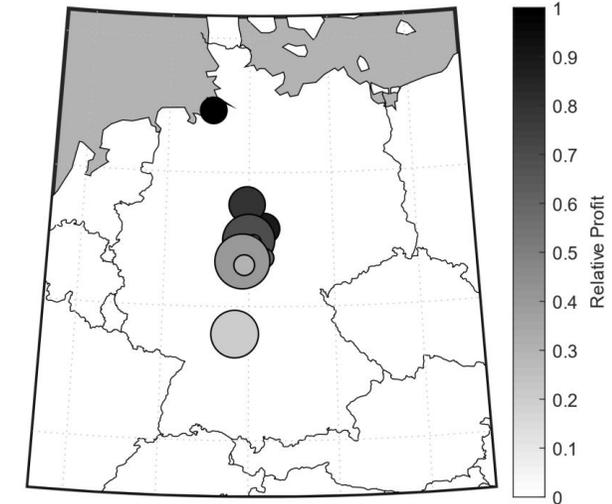
The case of wind energy in Germany

- Two key aspects of interest
  - the geographical and technological diversification within the clusters
  - general characteristics of cluster specific value components



The case of wind energy in Germany - Aggregation of investment choices

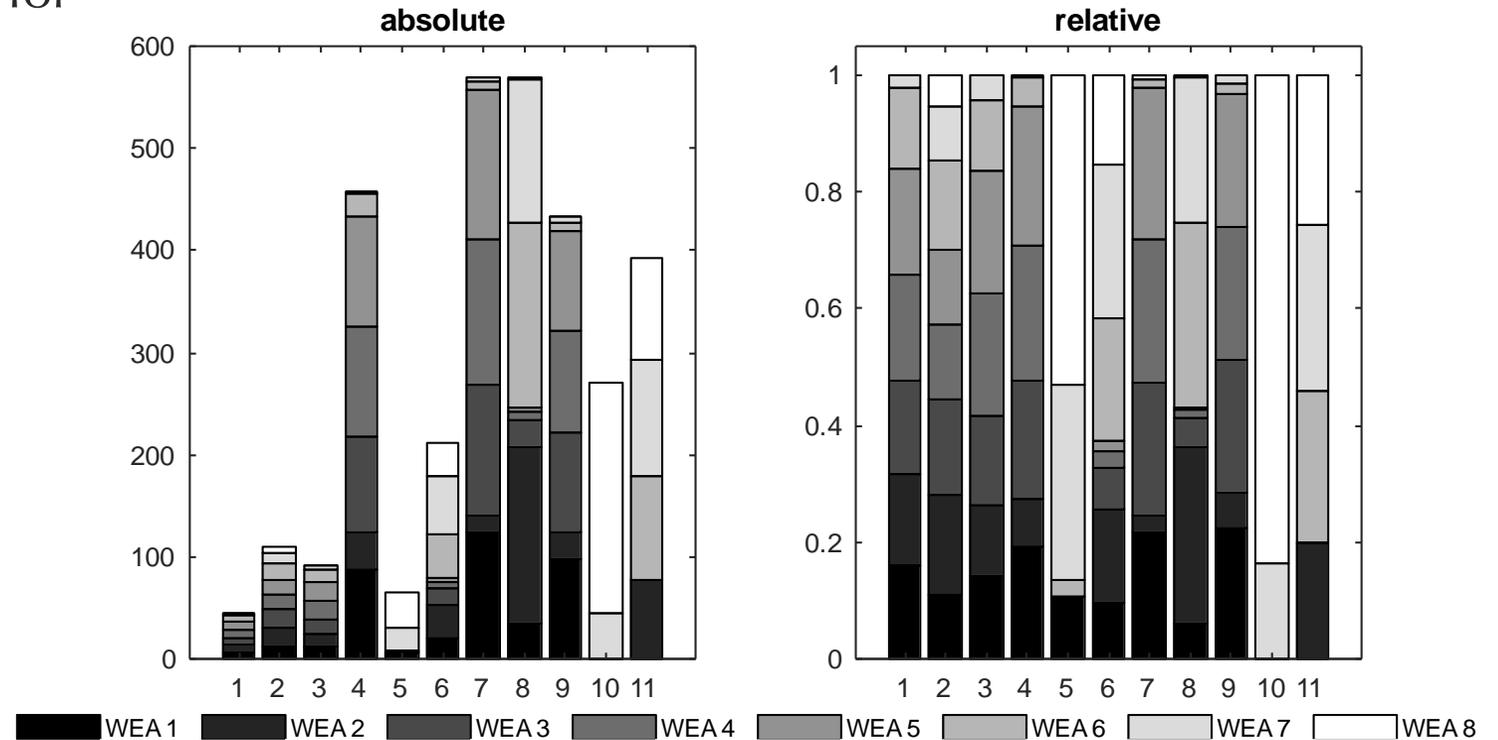
- Separation in two types of clusters
  - small, generally profitable clusters
  - large less profitable clusters
- Most profitable centroids (black) in North Germany or at mountain sites
- No obvious split between East and West clusters
  - spatial heterogeneity in that dimension not too important



# Technological diversification

The case of wind energy in Germany - Aggregation of investment choices

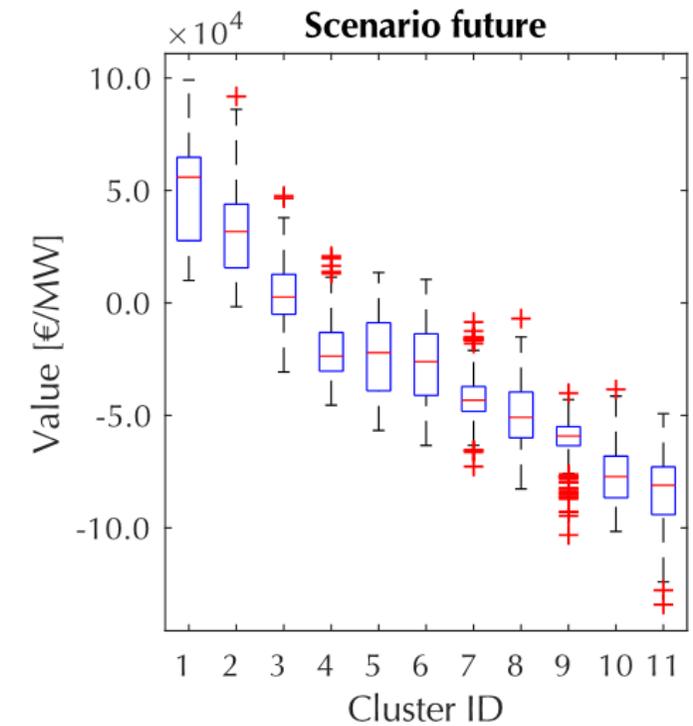
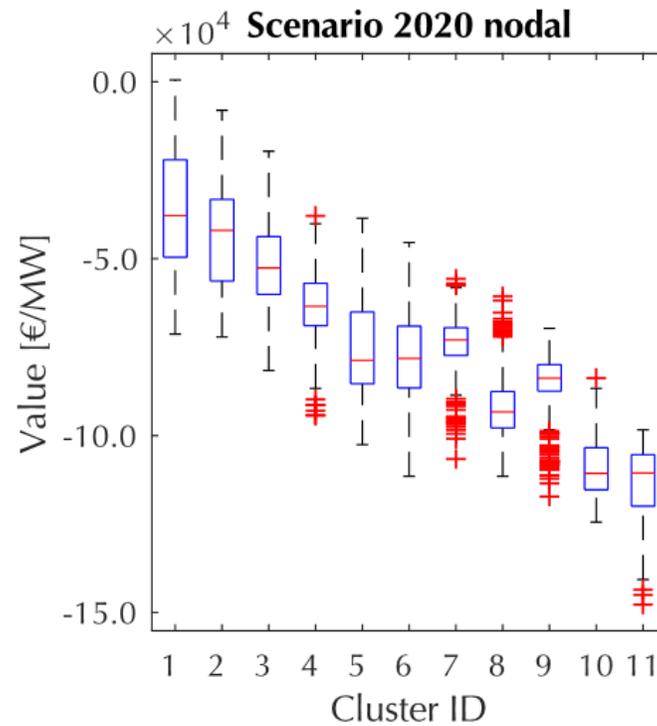
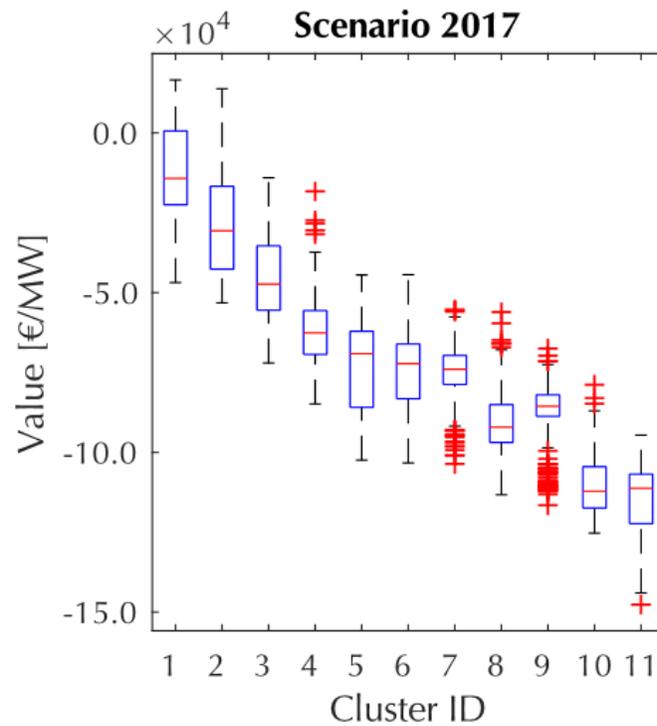
- Turbine type is similarly distributed for the four most profitable clusters
- Clusters five and ten are quite specialized on high speed wind turbines



# Obtained clusters: Profitability across scenarios

Evaluation of cluster statistics

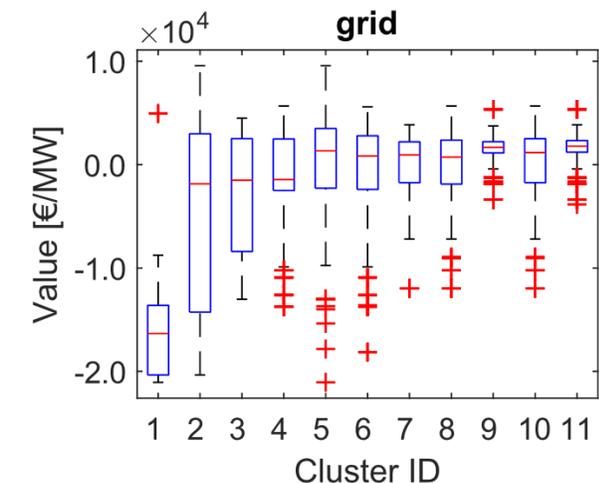
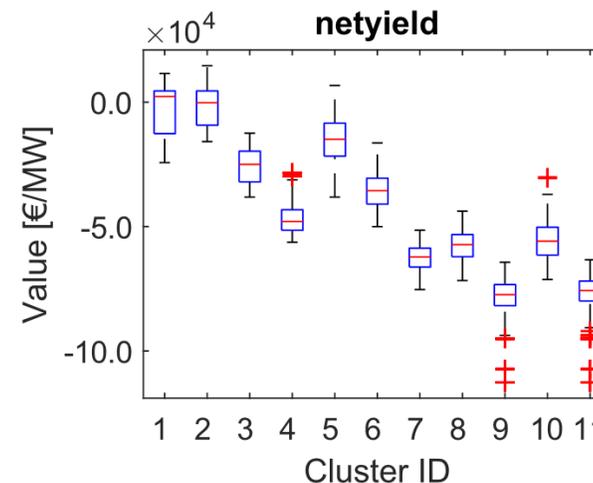
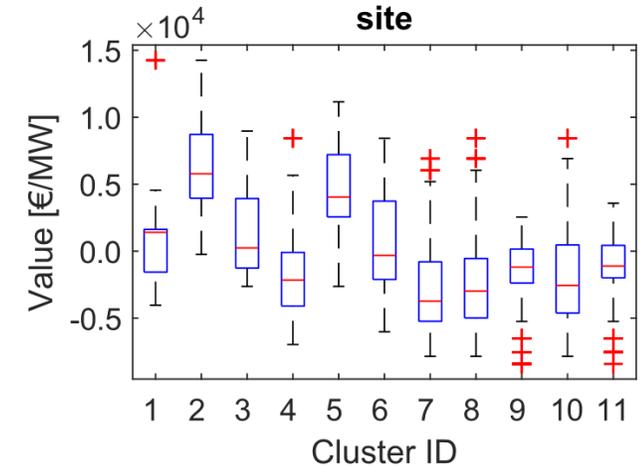
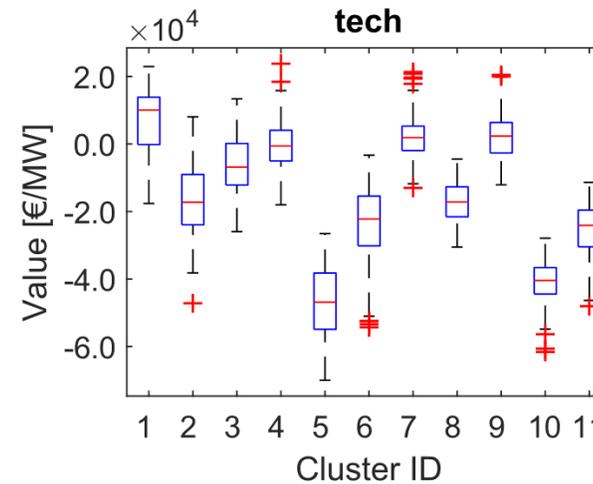
- Similar ranking of clusters in all scenarios
- Indication of limited changes in relative market value



# Obtained clusters: The four value components in comparison

## Evaluation of cluster statistics

- Technology and net-yield component
  - in absolute terms the highest
  - considerable divergences between clusters
- Net-yield component
  - Most important driver for the overall profit
- Site component
  - smallest influence
  - lower variability
- Grid component
  - negative due to scarce transfer capacities in times of high renewable infeed



- ✓ Methodology enables robust clustering of investment choices in view of use in aggregate models
  - ✓ Net-yield component is a key influencing factor for cluster formation
  - ✓ Diversity in sites does not impact clustering as much
  - ✓ The profitability ranking of the obtained centroids is rather robust against changes in portfolio mix and price
- 
- Test robustness of clusters against further scenarios (e.g. varying carbon caps or solar penetration)
  - Test of developed aggregation in an optimization environment

# Thank you for your attention!

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