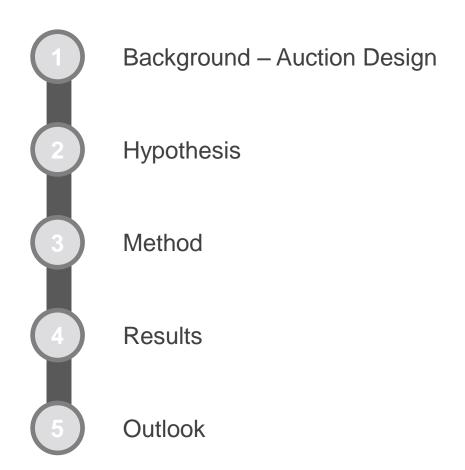
Auktionen für EE in kleinen Märkten

Möglichkeiten und Grenzen

Die Fallstudie, die zitiert wird, stammt aus dem Horizon2020 project AURES, grant number 646172.

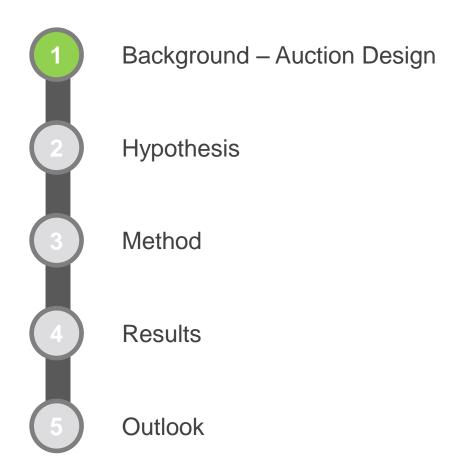








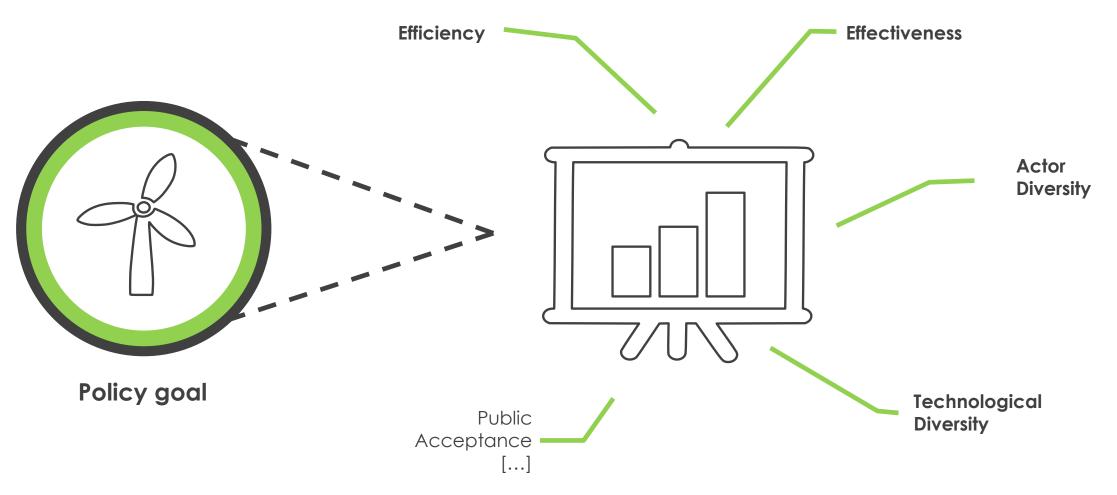






Auction Design 1/3

What to consider?

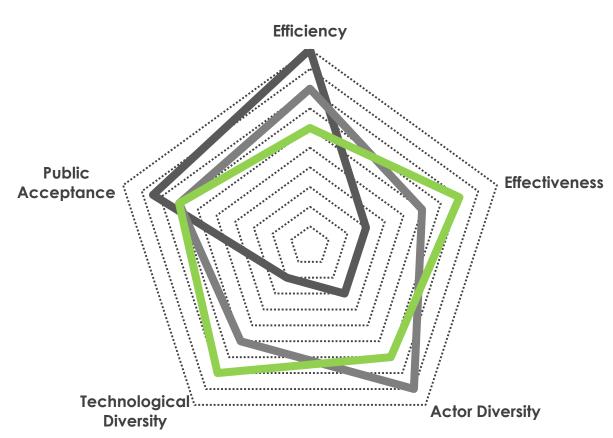




Auction Design 2/3

How do the goals line up?

Policy Goal Metrics



- When designing auctions "the devil lies in the details".
- Different policy goals, aside of (cost) efficiency and effectiveness can be envisaged.
- These policy goals sometimes diverge and can not all be combined equally as illustrated in the metrics (left hand side).
- Designing an auction carefully is therefore crucial to achieve the planned result.

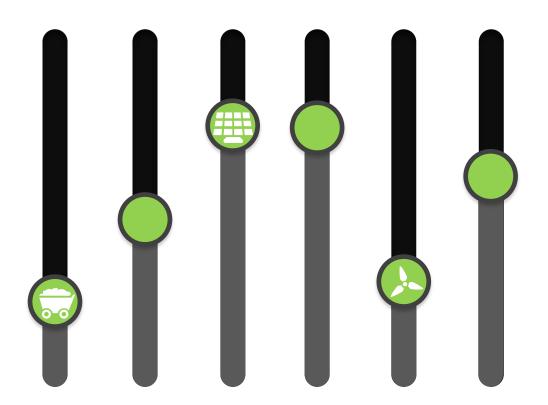
Figure: Performance of three possible auction policy designs (source: own elaboration)



Auction Design 3/3

How to achieve the goals?

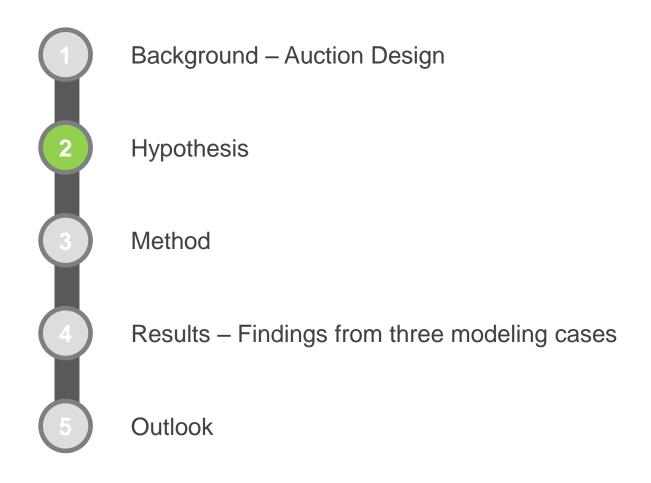
Auction design set screws



- Penalties/pre-qualifications
- Volume/schedule
- Level of technological diversity
- Budget flexibility
- Measures to influence actor diversity









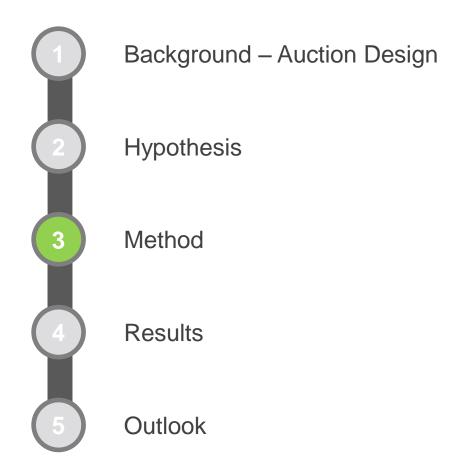


Questions concerning small markets

- Small markets with a limited number of potential auction participants are quite frequent in the European Union
- Implementing renewables auctions there can be challenging due to potential lack of competition, relatively small auctioned capacities and other factors.
- The main research question is how setting the auction schedule and the auctioned volume per round impacts the outcomes of the auction.
- Furthermore, a flexibility mechanism is tested, that allows budget shifts between rounds and can potentially increase deployment rates.









Method 1/2

Combining Auction Theory and ABM for the model-based analysis.

Modeling of new auction design schemes

- Insights from policy makers on auction design process and/or detailled auction outcomes.
- Using a novel approach to analyse a new support scheme.
- Auction theoretic concepts implemented in an agent-based model (ABM).
- Use of all available data to model the respective market and their participants close to reality.
- Statistical analysis of auction outcomes and technological data to complement the analysis.



Method 2/2

How does learning take place over auction rounds?

Bidder's calculus before and during the

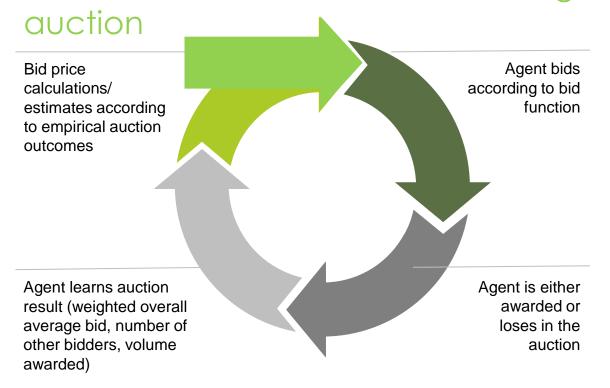
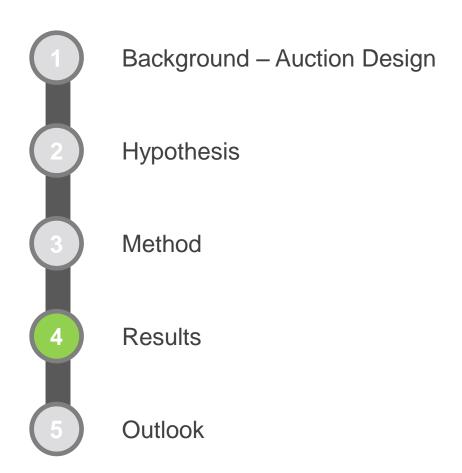


Figure: Simplified depiction of the agent's learning process in the model (source: Anatolitis and Welisch, 2017)

- Mean value (μ) is the central configuration parameter.
- Agents' learning algorithm consists of adapting μ to new information generated in the auction rounds.
- First round: the assumptions on μ of F() are based on each agent's own signal (her cost).
- In each round, new information is incorporated: agents adjust cumulative distribution function by updating μ with the last round's overall mean bid.









How does setting the schedule and volume influence auction outcomes?

Background

- Different criteria for the design of multiunit renewable energy (RES) auctions in small markets are assessed.
- The multi-technology RES auctions in Denmark beginning 2018 serve as an exemplary case for the assessment.
- Two auction rounds of 200 MW planned for 2018-2019, after that (up to 2025, frequency and volume to be auctioned are still up to discussion).
- Pay-as-bid, multiple technology "open-door" common tender scheme (onshore wind, solar PV and offshore wind).
- Setting the auction schedule and the auctioned volume per round impacts the auction outcomes.
- A flexibility mechanism that allows up to 50% of the auction volume to be shifted between auction rounds is also tested.



How does setting the schedule and volume influence auction outcomes?

Approach

- Number of and capacity bid by solar PV bidders estimated from recent joint solar PV auction between Denmark and Germany; for onshore wind bidders this information stems from the most recent analysis on the Danish market (ENS, 2017).
- Bidders' costs were calculated by estimating the LCOE with ENS data on technology costs and developments.
- Bidders were subdivided into multi and single project bidders; multi-project bidders submit three bids each for simplification reasons.
- Pay-as-bid pricing was modeled as explained earlier.



How does setting the schedule and volume influence auction outcomes?

Fixed Round 1 Round 2 Round 3 Round 4 Fixed Round 1 Round 2 Round 3 Round 4 Reduced Budget Reduction passed on to next round Fixed Budget Size Project Project Project Project Not supported

Figure: Comparison of flexible and fixed budget options (source: own elaboration)

Results

- First application of the model, was to look into the impacts of varying volume and schedule.
- Specifically, a flexibility mechanism (left hand side) was implemented and assessed.
- Too small volumes exclude large-scale/multiproject bidders and decrease capacity expansion; bid prices don't increase strongly with higher volumes/fewer rounds.
- A flexibility mechanism can help to ameliorate this.



How does setting the schedule and volume influence auction outcomes?

Auction round comparison (2018-2025) 1.6 1.4 1.2 40 (MW) 10 10 Annual auctions (100 MW) 10 10 Annual auctions (200 MW) Average non-realization

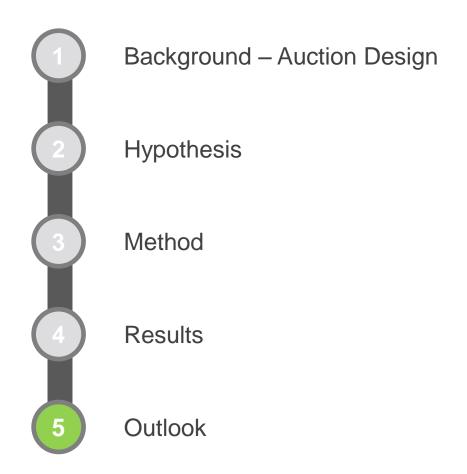
Figure: Comparison of prices and non-realization in annual and bi-annual DK auction schemes (from 2018 to 2025) (source: own elaboration)

Results

- The bi-annual scheme performs better in terms of bid prices, whereas the annual scheme is able to ensure a much higher realization rate.
- Average non-realization in the form of rejected marginal projects is almost double the amount in the bi-annual case.
- Agent distribution is not shown, as due to the large price differences and relatively low auction volumes, only onshore wind bidders were awarded.



Agenda The way to go & stops in between.







How does setting the schedule and volume influence auction outcomes?

Applicability for small markets

- An auction design for small markets should account carefully for the volume auctioned in each round
- It should ideally be flexible in allowing for the marginal bidder to exceed the auctioned volume.
- Furthermore, low auction volumes could lead to a concentration of the cheapest technology and largescale bidders.



Thank you!

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Back-Up: Pricing Mechanisms

Nomenclature

Auction		
range(T)	rounds per iteration	
t	auction round	
n^t	number of bidders in round t	
n_s^t	number of successful bidders in round t	
d^t	total demand in round t	MW
s^t	total supply in round t	MW
p_{lim}^t	price limit in round t	ct/kWh



Back-Up: Pricing Mechanisms

Nomenclature

 c_i^0 bidder i's initial costs in the first round ct/kWh

 q_i quantity offered by bidder i MW

 δ discount factor $\in (0,1)$

comp initial assumption on competition

succ initial assumption on successful bidders

 λ^t degression factor in round t

 β bidding function

x bidder's signal

δ uncertainty factor

