

# Effects of price-making and portfolio size in stochastic optimization of trading in sequential electricity markets

Emil Kraft, Kim K. Miskiw, Stein-Erik Fleten  
16.02.2023, IEWT2023, TU WIEN



# Contents



**Motivation**



**Stochastic model**

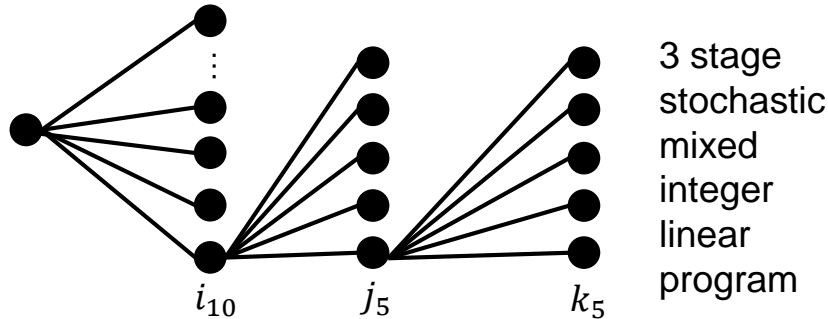
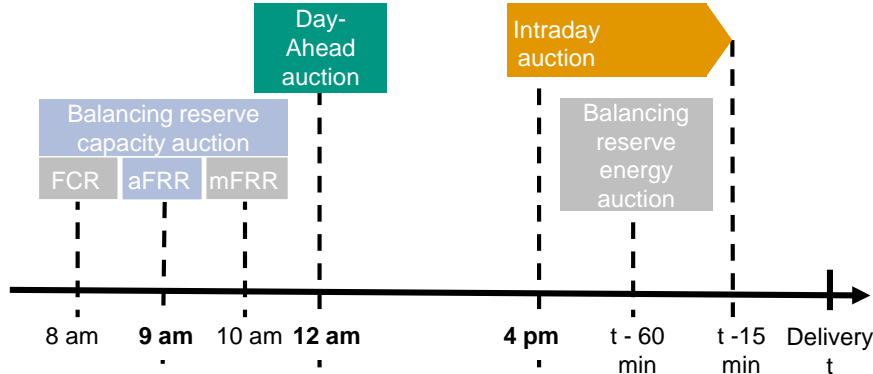


**Extensions and results**



**Summary and further steps**

# Motivation



Uncertainty

Prices aFRR

Prices DAM, RE forecast

Prices IDM, Update RE forecast

## Research Gap

- Increasing importance of products with short lead time [1, 2]
- Currently trading decisions often made deterministic [3, 4]
  - Large value of coordination of bids?
  - Dependency on portfolio composition?
  - Effect of price making?

Flexibility [3]

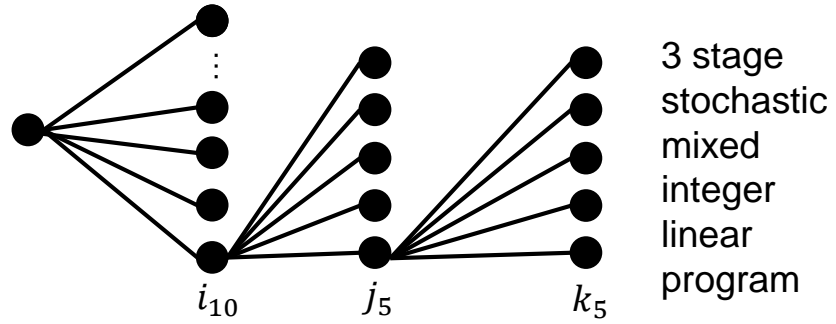
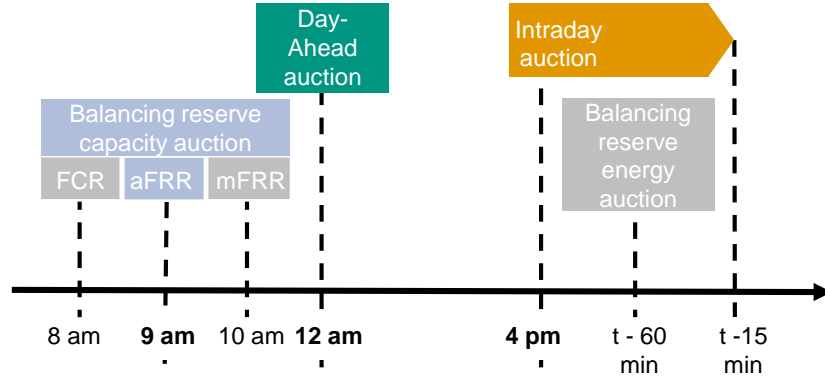
Price impact [3]

Portfolio size [4]

Risk aversion [1]

Price maker

# Stochastic optimization model






<b>Uncertainty</b>	Prices aFRR	Prices DAM, RE forecast	Prices IDM, Update RE forecast
--------------------	-------------	-------------------------	--------------------------------

## Profit maximisation with risk consideration

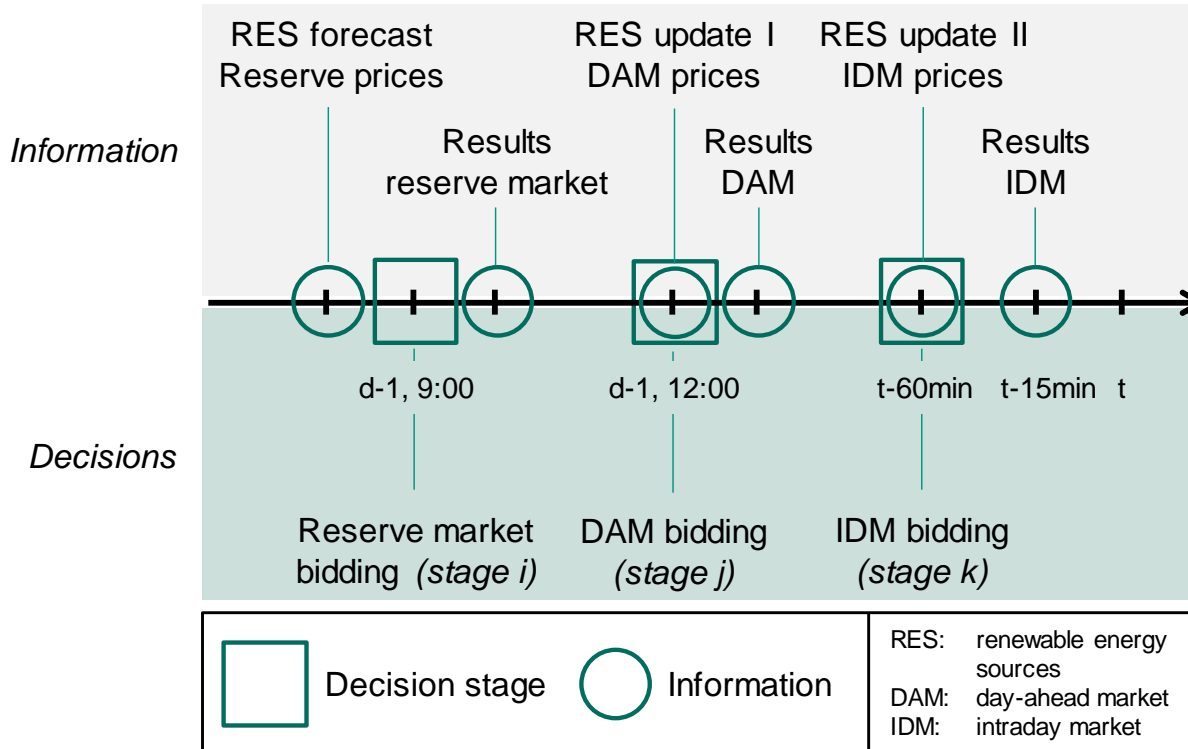
$$\max (1 - \lambda) \cdot \mathbb{E}(\pi_{i,j,k}) + \lambda \cdot \text{CVaR}_\alpha(\pi_{i,j,k})$$

$$\forall \alpha \in (0,1), \lambda \in [0,1]$$

$\lambda = 0$  risk neutral  
 $\lambda = 1$  very risk averse

-  100 MW Biomass  
100 MW PV
-  German market data and setting  
July 2019 – March 2020  
1 day = 96 quarter hours
-  18 type days  
Work day, season,  
low/medium/high residual load

# Decision structure



Kraft et al. (2022)

# Extensions

## Start Point

Multi-stage stochastic model & derived scenarios  
Kraft et al. (2022)

### 1. Implement Price Impact

$$\text{Revenue} = (\hat{y}_{j,k,qh}^{ID} - b_{qh} * x_{i,j,k,qh}^{ID,trade}) * x_{i,j,k,qh}^{ID,trade}$$

- IDM with limited liquidity
- Continuous trading  $\approx$  one uniform auction with ID3 price
- Impact derived from historical data following [3,6,7]
- Accepted bids from last 3 h
- Ordered and linear regression fitted  $\rightarrow b_{qh} = \text{slope}$
- Linearized following [6]

### 2. Formulate Bidding Heuristic

### 3. Sensitivity Analysis

# Extensions

Start Point

Multi-stage stochastic model & derived scenarios  
 Kraft et al. (2021)

## 1. Implement Price Impact

$$\text{Revenue} = (\hat{y}_{j,k,qh}^{ID} - b_{qh} * x_{i,j,k,qh}^{ID,trade}) * x_{i,j,k,qh}^{ID,trade}$$

- IDM chosen ← limited liquidity
- Continuous trading  $\approx$  one uniform auction with ID3 price
- Impact derived from historical data following [3,6,7]
- Accepted bids from last 3 h
- Ordered and linear regression fitted  $\rightarrow b_{qh} = \text{slope}$
- Linearized following [6]

## 2. Formulate Bidding Heuristic

Optimizing market stages separately [3,6,8]  $\rightarrow$  Myopic bidding heuristic

### Stage 1:

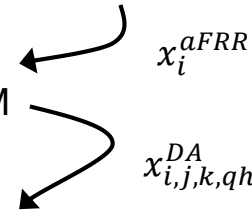
Maximise profit aFRR & DAM assuming average realizations of DAM scenarios

### Stage 2:

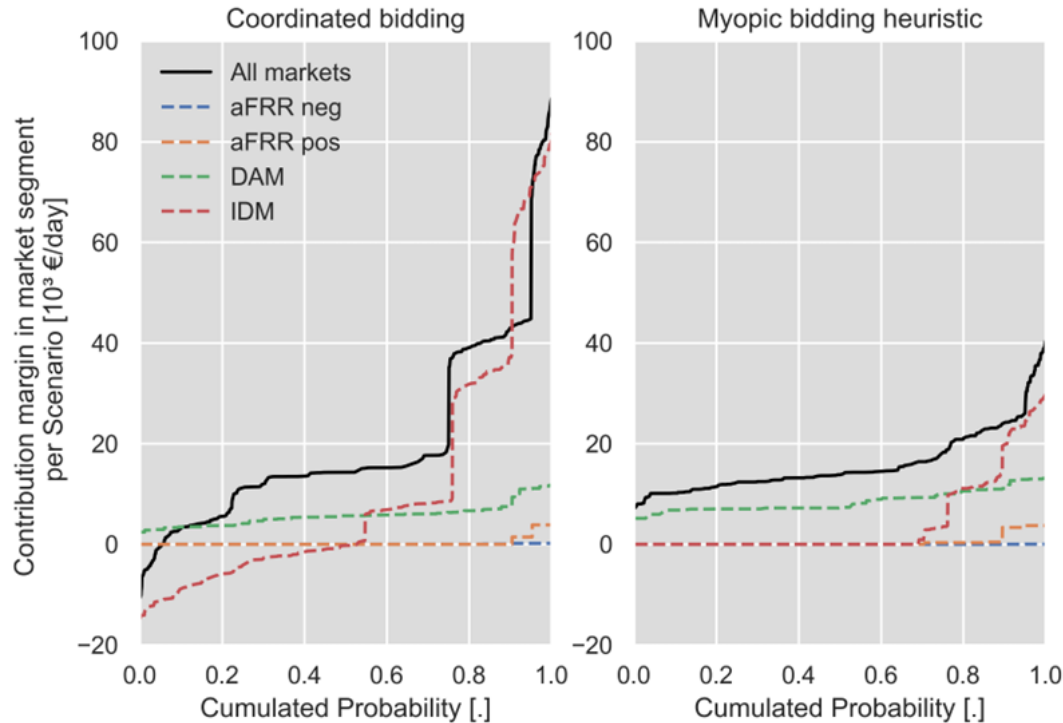
Maximize profit DAM

### Stage 3:

Maximize profit IDM



## 3. Sensitivity Analysis



## Comparison

- Maximise immediate revenue – greedy approach
- Shift volume to earlier market stages
- Less volume left for Intra-day
- No negative revenue in myopic bidding

➤ Significant difference in head and tail



# Extensions

Start Point

Multi-stage stochastic model & derived scenarios  
 Kraft et al. (2021)

## 1. Implement Price Impact

$$\text{Revenue} = (\hat{y}_{j,k,qh}^{ID} - b_{qh} * x_{i,j,k,qh}^{ID,trade}) * x_{i,j,k,qh}^{ID,trade}$$

- ID chosen ← limited liquidity
- Continuous trading  $\approx$  one uniform auction with ID3 price
- Impact derived from historical data following [3,6,7]
- Closed bids from last 3 h
- Ordered and linear regression fitted  $\rightarrow b_{qh} = \text{slope}$
- Linearized following [6]

## 2. Formulate Bidding Heuristic

Optimizing market stages separately [3,6,8]  $\rightarrow$  Myopic bidding heuristic

### Stage 1:

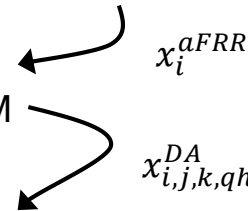
Maximise profit aFRR & DAM assuming average realizations of DAM scenarios

### Stage 2:

Maximize profit DAM

### Stage 3:

Maximize profit IDM



## 3. Sensitivity Analysis

- Follow up on interdependencies in literature
- Analyse spread of average contribution margin = value of coordination

# Extensions

## 1. Price Impact

## 2. Bidding Heuristic

## 3. Sensitivity Analysis

Base Case:



100 MW Biomass  
100 MW PV

- Risk neutral trading
- Technical parameters taken from [1]
- One type day (transition, medium load, weekday)

### Risk aversion [1]

1.  $\lambda = 0.1$
  2.  $\lambda = 0.5$
- $$(1 - \lambda) \cdot \mathbb{E}(\pi_{i,j,k}) + \lambda \cdot \text{CVaR}_\alpha(\pi_{i,j,k})$$

### Portfolio size [4]

1. 500 MW Biomass, 500 MW PV
2. 1GW Biomass, 1GW PV → similar size to other portfolios analysed

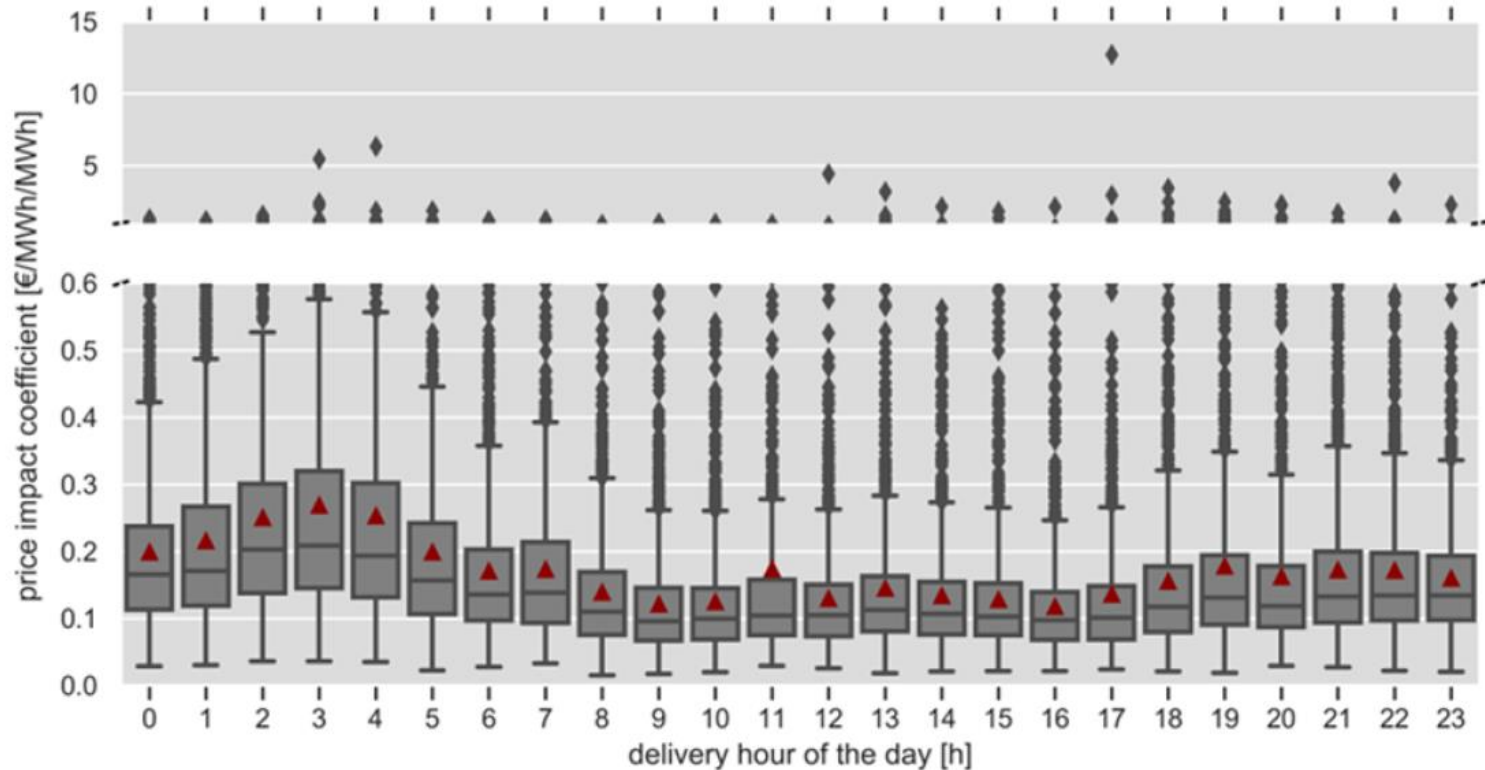
### Price impact [3]

1. Average price impact in each hour
2. 0.25 quartile of price impact factor
3. 0.75 quartile of price impact factor

Flexibility parameter variation [3]

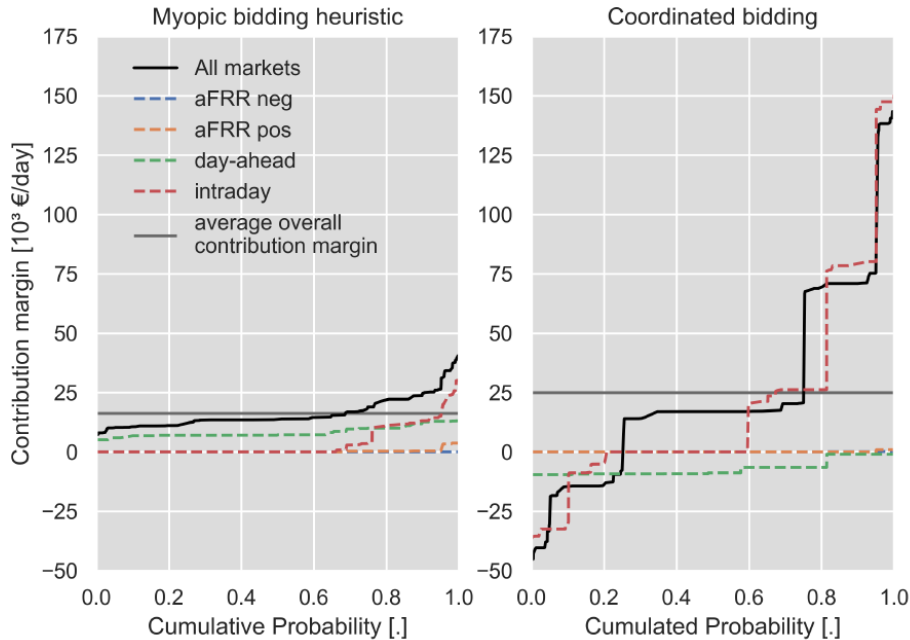
Share of renewables in portfolio [5]

# Largest price impact during night and smallest in the peak hours

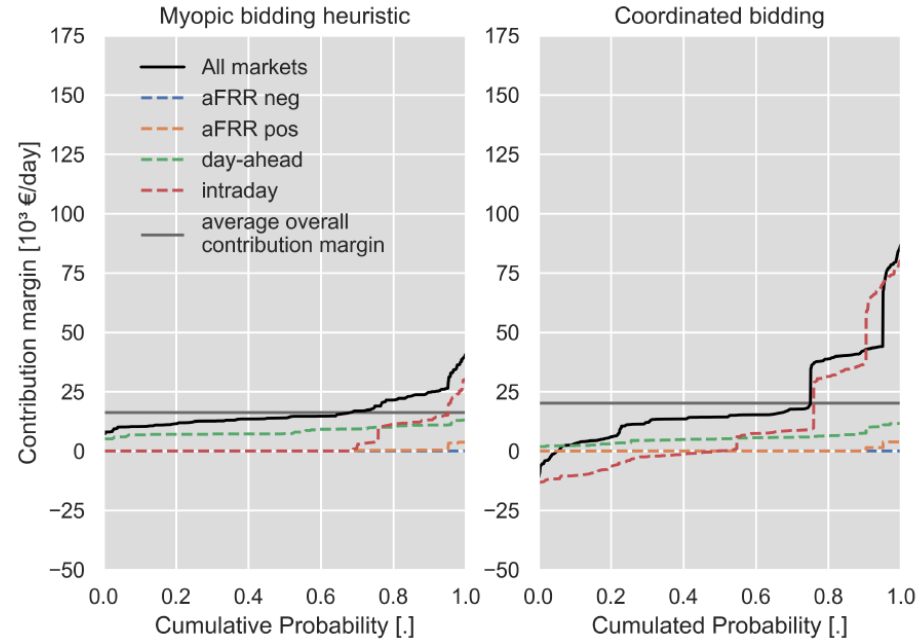


# Price making has little impact on heuristic, but large impact on coordinated trading strategy

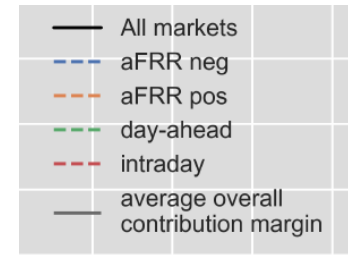
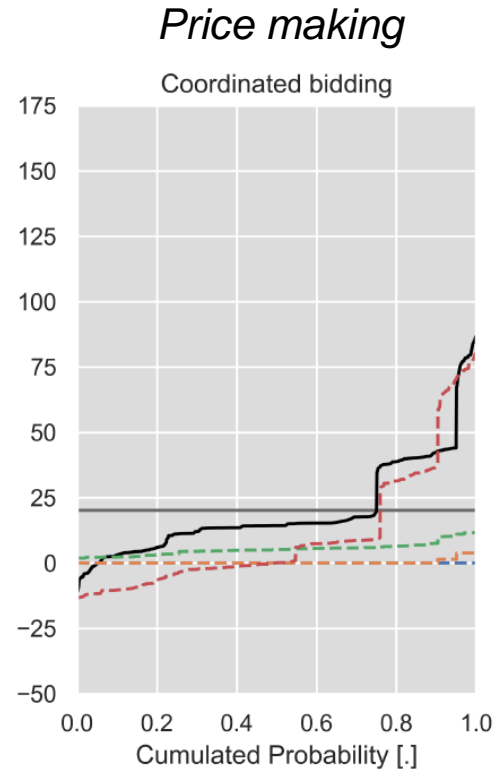
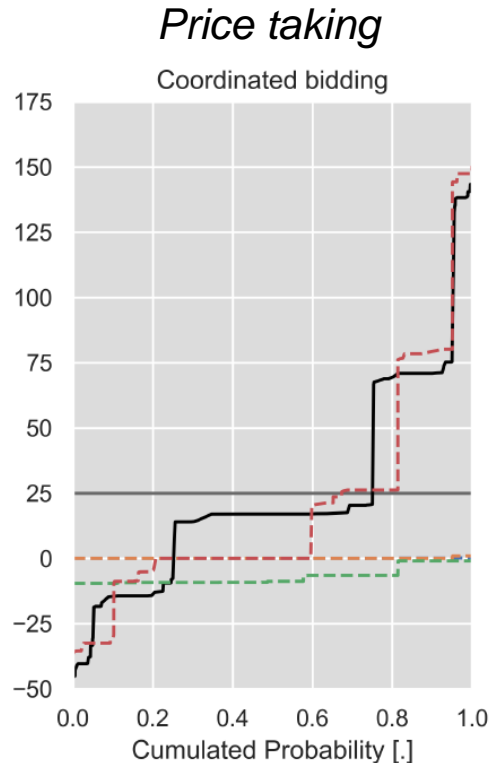
*Price taking*



*Price making*



# Less speculation on high intraday prices



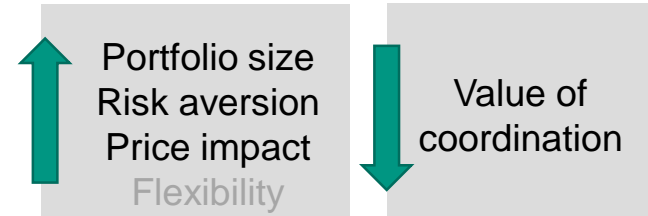
# High variance across type days complicates drawing general conclusions

- Gain of coordination for portfolio depends on:
  - Price relations and steepness of supply curve in market segments
  - Inframarginal / extramarginal power plant
  - Share of renewable generation
  - Flexibility of portfolio / degrees of freedom in dispatch
- For case study:
  - In summer lower than in transition and winter
  - Weighted relative value of coordination ~18%
  - Pay attention when comparing relative values!

# Conclusion

## Summary

- Price impact depiction
  - Price impact for IDM modelled
  - Price impact estimation through empirical data (2019-2020)
  - Extension of stochastic optimization model computationally expensive but feasible through piece-wise linearization
- Benefit of coordinated bidding
  - Overestimation of IDM profitability by neglecting price making (Kraft et al. 2022)
  - Lower but robust value of coordination across markets
  - Large impact of uncertainty modelling and portfolio configuration on value of coordination (no one-size-fits-it-all conclusion)
  - Given an increase in renewables and the importance of IDM, further increase in importance can be expected for coordinated bidding



## Outlook



# Conclusion

## Summary



## Outlook



- Updating of scenarios to more recent data
- Evaluation of advantages of stochastic approach regarding risk exposure
- Scalability of model
- Interdependencies of influencing factors
  - Which situations require which degree of coordination?
  - How do sensitivities interact with each other?
- Add further technologies, e.g. storages of different time scales
- Translate principles to long-term energy markets



**Thanks for your attention!**