

Understanding bidding behavior in the electricity spot market:

an empirical study of European Markets in 2022

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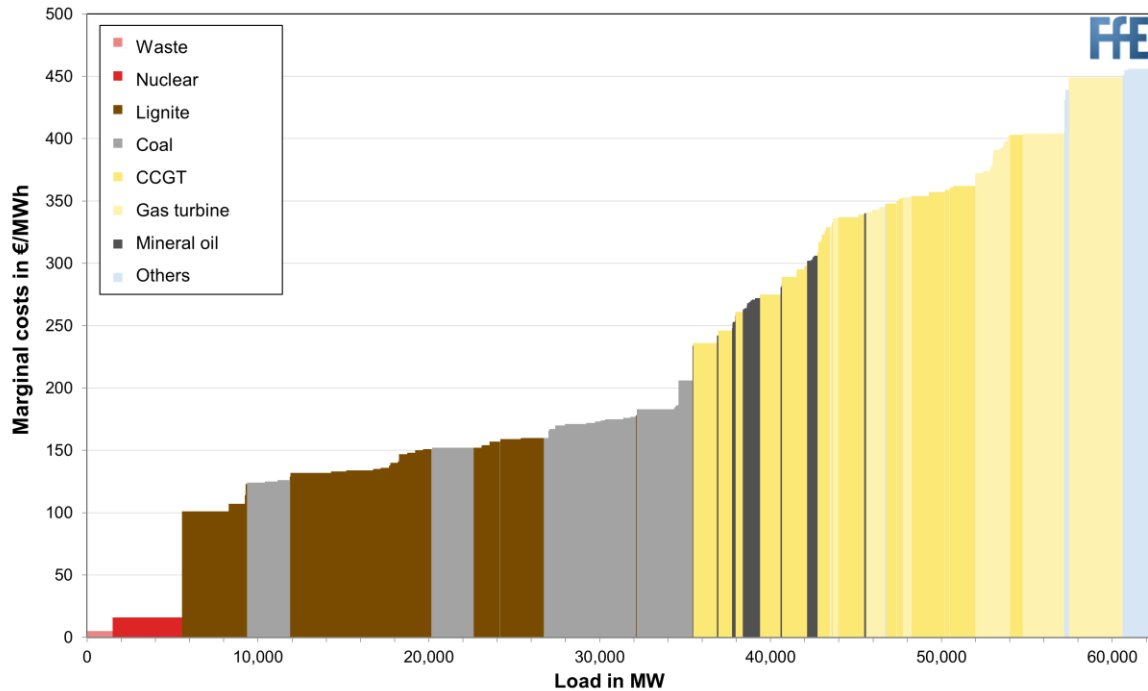
ABM4Energy Conference 2025

Introduction

Models assume bidding according to Marginal Cost

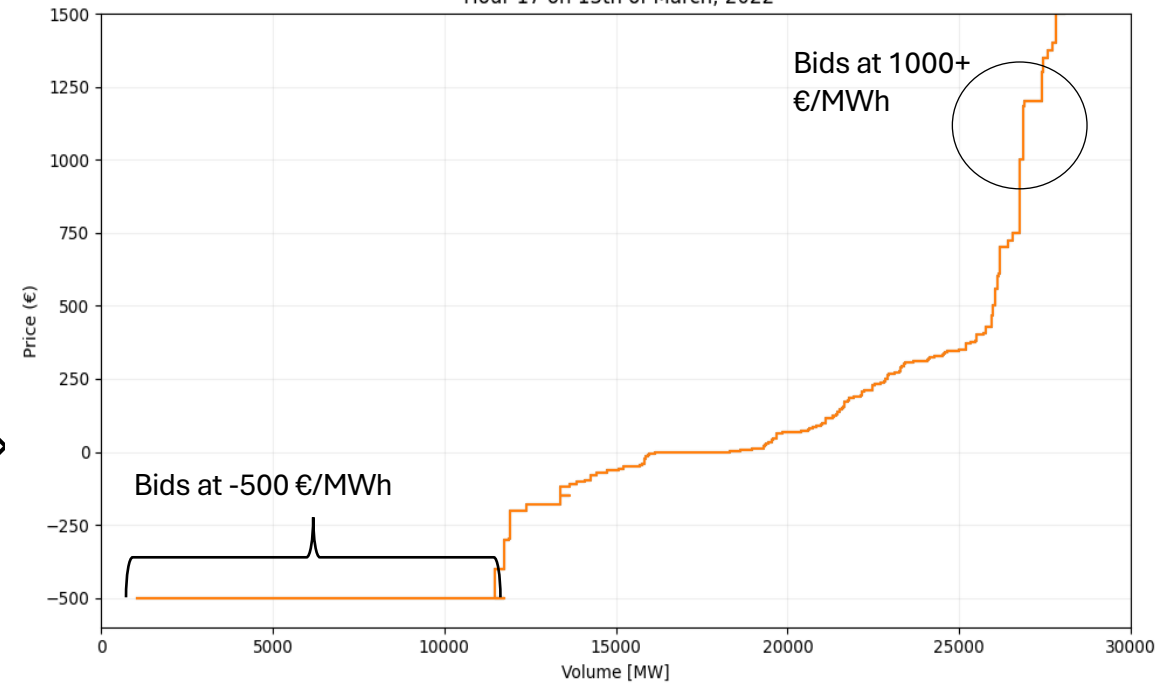
Empirical bidding is significantly different:

Merit Order of thermal power plants in Germany
 2022 (90 € CO₂-costs, 123.3 €/MWh gas price)



Theoretical MC-based supply curve from [1]

Bidding Curves for Germany:
 Hour 17 on 13th of March, 2022



Empirical Bidding Curve for Hour 17 for the 13th of March 2022, Germany

=> Can we understand **Empirical Bidding** better?

Research Questions & Goals

1. Define *Expected* bidding behavior

- What is regular empirical bidding like?
- How is this influenced by market drivers?

PART 1: Clustering of Curves

2. Investigate bidding *Anomalies*

- What deviations from this expected bidding do we observe?
- What drives such deviations?

PART 2: Anomaly Detection

1. Understand **Expected bidding** and **Deviations from empirical data**

Our Study

2. Define the empirical bidding of realistic actors

Contribute to ABM?

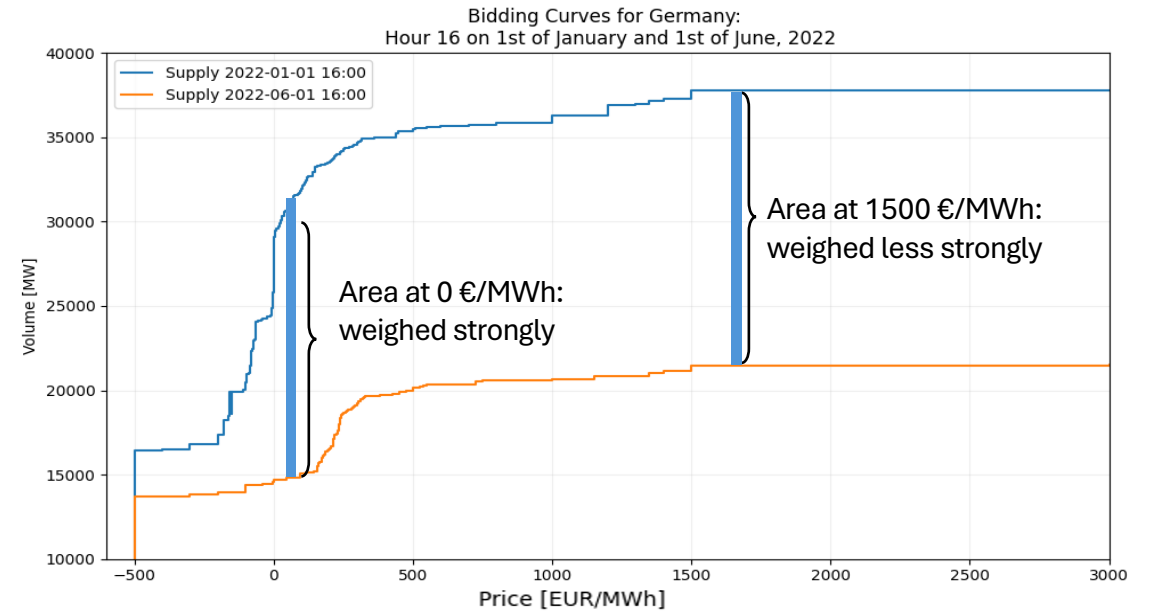
3. Implement in spot market models (ABM/...?)

Methodology (1) – Clustering

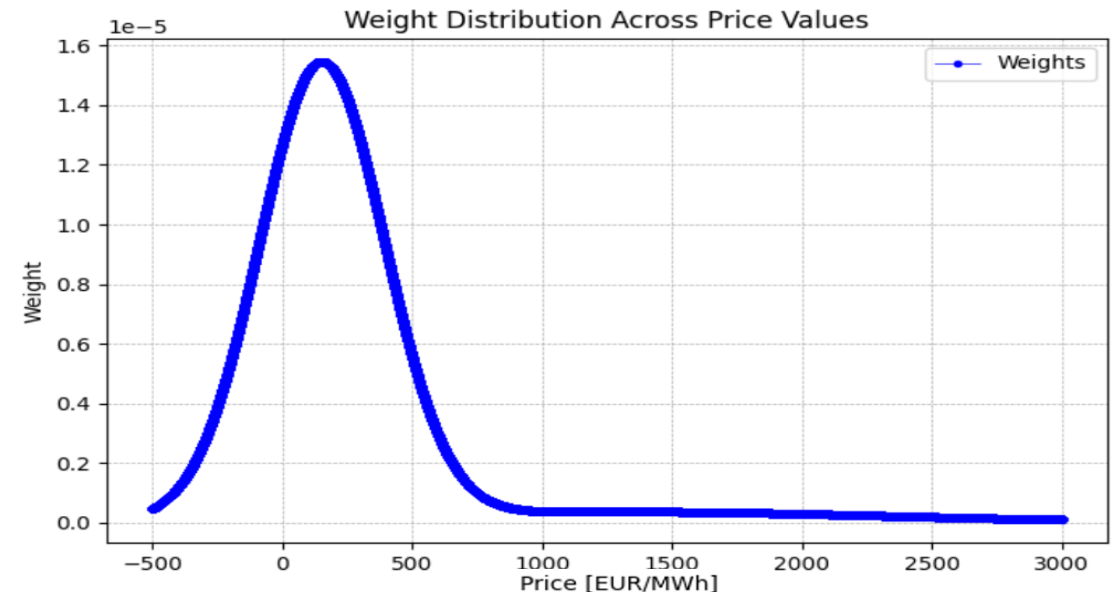
- Cluster similar curves => based on **shape**, following methodology of [2]
 - Area * weights at each price
- => **1 distance value** per pair of N curves
- Weights ~ Frequency of bids at each price
 - NxN Distance matrix => *Hierarchical Clustering*
 - Calculate mean driver data per cluster
- => **Relate *Expected* bidding curves per cluster to the driver data per cluster**

Use Case:

- 2022: Energy Crisis/War
- => Interesting year to test Method
- 8760 Aggregated bidding curves from EPEX Spot [3]



Method to determine inter-curve distances

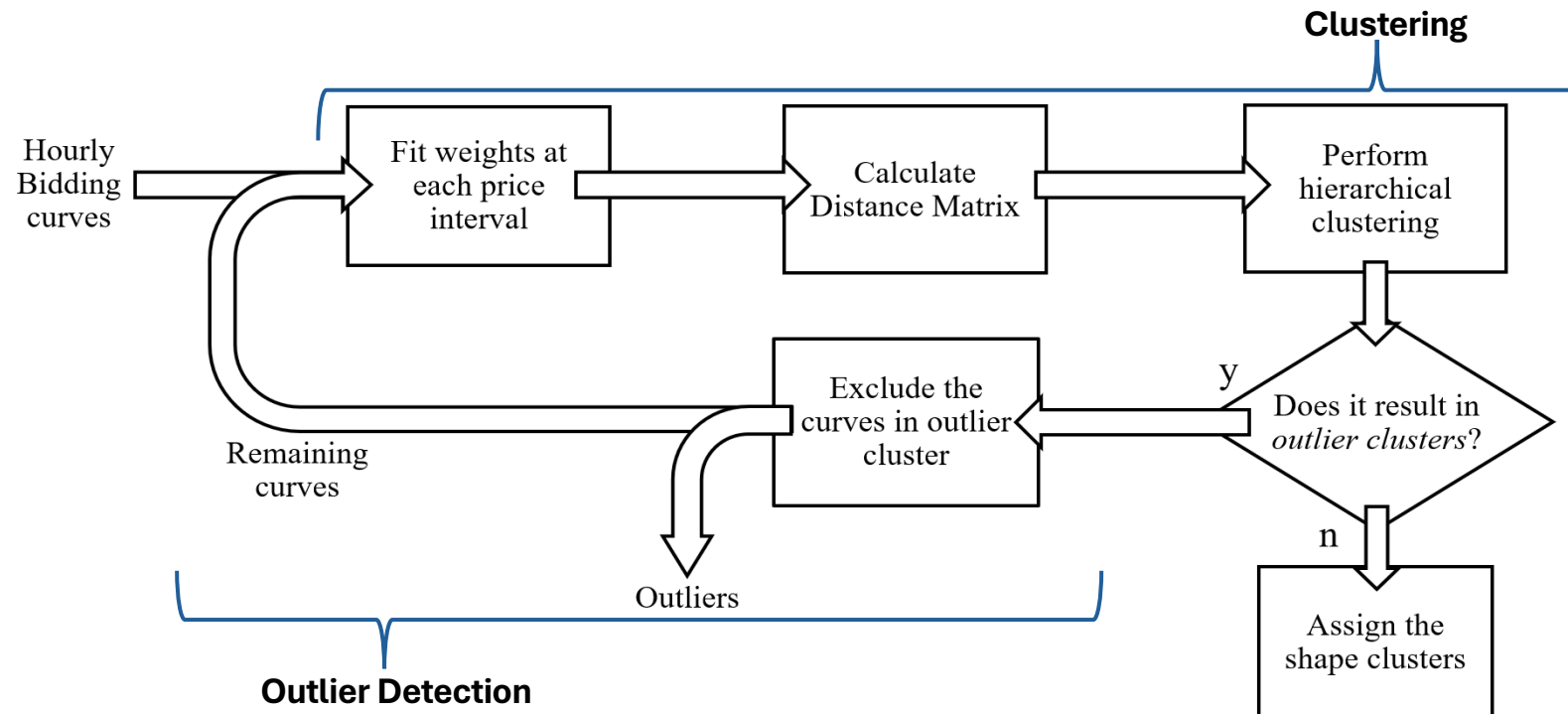


Clustering weights applied at each price

Methodology Anomaly Detection (1) - Outliers

1. *Hierarchical Clustering*: groups alike curves
2. Creates extremely small cluster? => Indicates **curves deviate significantly from all other curves**
3. *Outliers* are excluded, clustering reiterated

=> Until **No more Outliers found, large bidding curve clusters remain**



Methodology to generate clusters and detect outlier clusters

Methodology Anomaly Detection (2) – Cluster Prediction

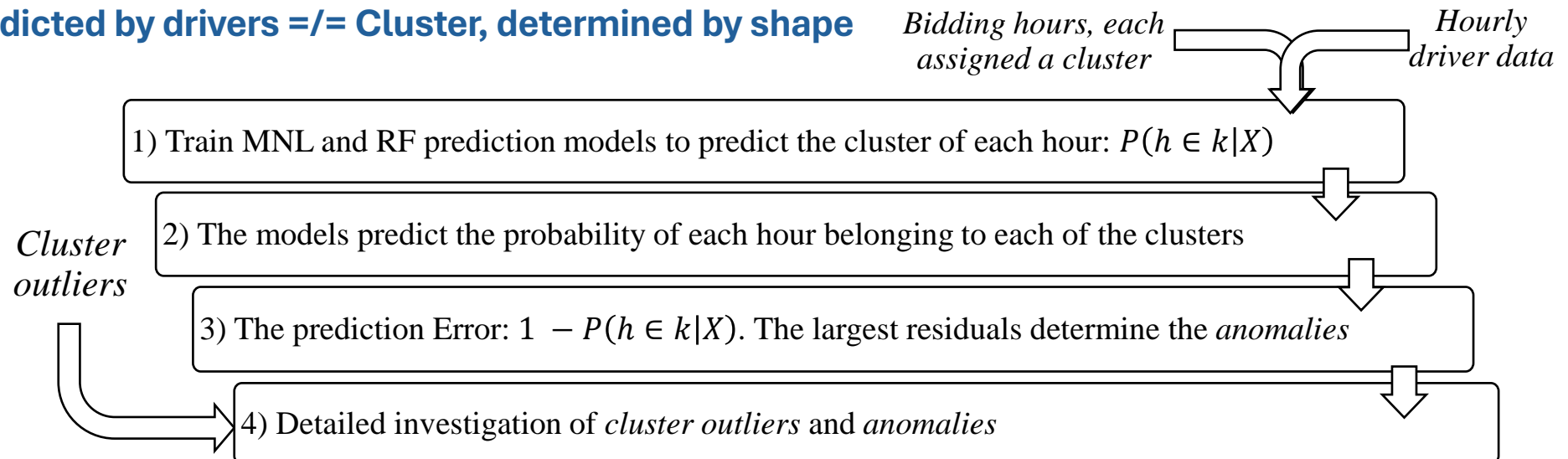
- Every hour is assigned a Cluster
- **What if driver data X_h deviates from other hours in cluster?**
- Predict Probability P that hour $h \in$ Cluster (k) based on X_h
 1. MultiNomal Logit (MNL) model
 2. Random Forest model
- X_h = Hourly driver data: °C, Load/Generation forecast, VRE, Fuel prices¹
- Prediction Error defines *Anomalies*:

$$P(\hat{Y} = k | X) = \frac{e^{-\beta_k^T X}}{\sum_{j=1}^K e^{-\beta_j^T X}},$$

for $k = 1, 2, \dots, K$

MNL Prediction Model

Cluster, predicted by drivers \neq Cluster, determined by shape



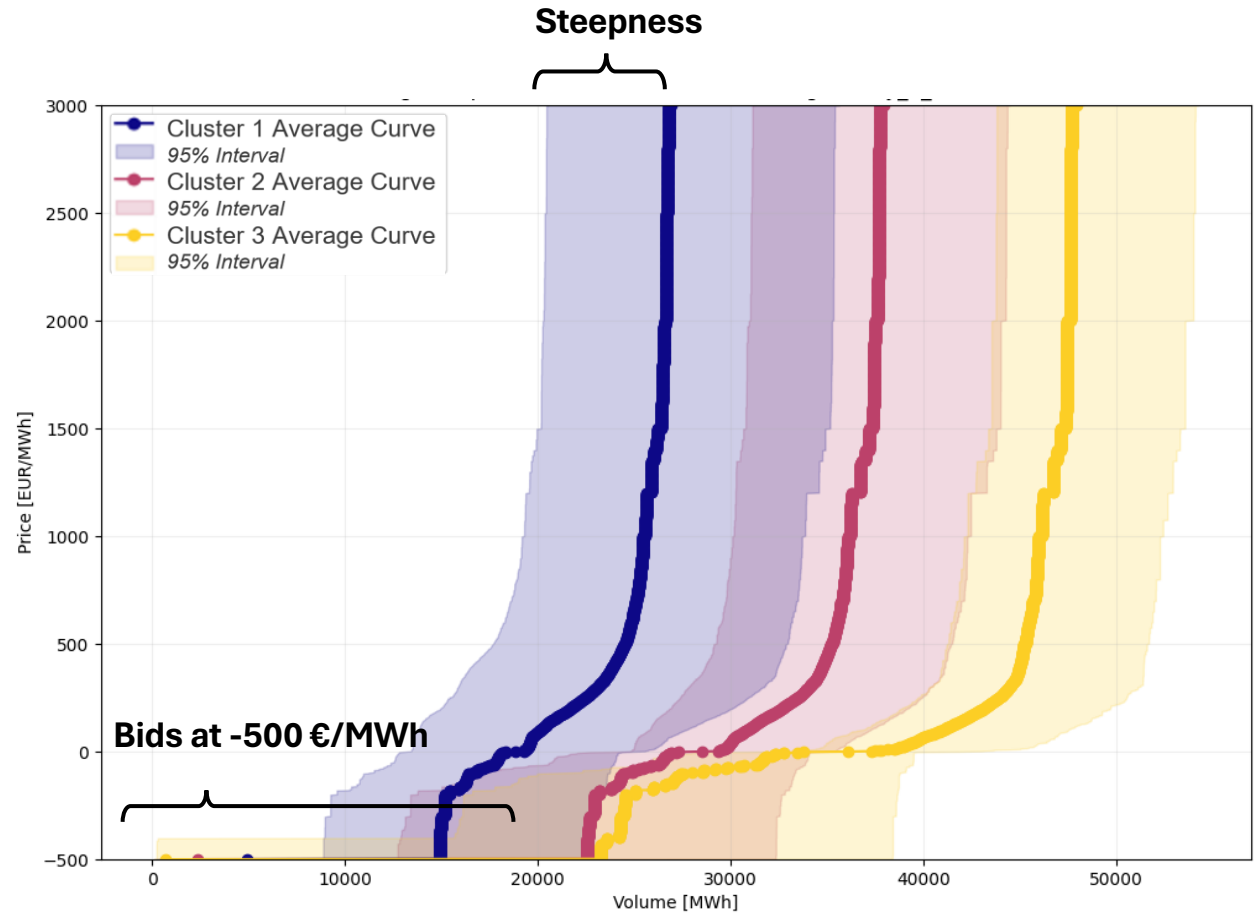
Methodology to determine Anomalies

¹ Data from Entso-E [4]

Case Study: Germany 2022

- **3 Clusters**
- Cluster 1 – Cluster 3 differ in:
 - Bids at minimum price
 - Steepness of curves
- Driver Data:
 - Cluster 1: Low VRE
 - Cluster 2: Moderate VRE
 - Cluster 3: High VRE
- VRE determines the *expected bidding curves* in Germany
- Belgium: very similar findings.
- Denmark1: Wind is the only differentiator between Cluster:

=> Significance of Wind vs Solar in driving bids depends on capacity in zone



	Temp. °C	Solar MW	Wind MW	Generation MW	Load MW	Price €/MWh	Gas €/MWh	Oil €/kBarrels	Coal USD/t	Carbon €/tCO2
Cluster 1	8.86	3096	12100	53541	54992	264.5	136.0	46.7	290.8	80.7
Cluster 2	14.39	14758	17713	63808	60205	181.4	126.5	46.0	288.7	82.5
Cluster 3	13.09	14322	29260	66954	56812	44.3	100.6	45.0	272.7	83.7

Average Curves and driver data for the three clusters of the German Market, 2022

Results (2) – Outlier Anomalies Germany

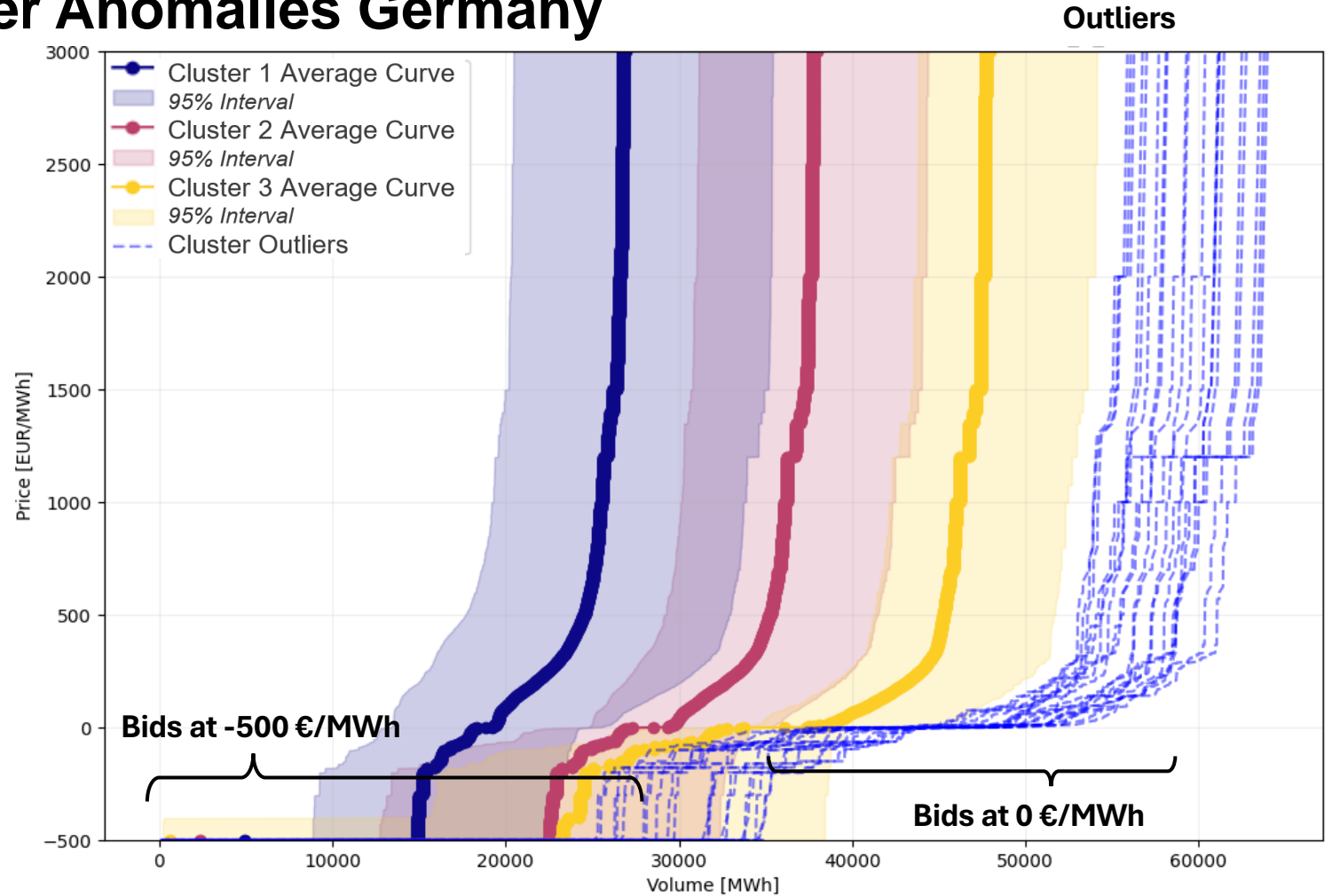
20 Outliers detected for Germany

- Very high volumes at -500€/MWh
- Very high volumes at 0 €/MWh

Driver Data:

- ⇒ VRE exceeds Cluster 3 means
- ⇒ Extreme Wind + Solar hours

Extremes in Renewables causes extremes in bidding



	Temp. °C	Solar MW	Wind MW	Generation MW	Load MW	Price €/MWh	Gas €/MWh	Oil €/kBarrels	Coal USD/t	Carbon €/tCO2
Cluster Outliers	11.79	21397	33325	70740	57135	-3.2	90.3	44.9	287.1	83.3
Cluster 3	13.09	14322	29260	66954	56812	44.3	100.6	45.0	272.7	83.7

Average Curves and Cluster Outliers for German Market, 2022

Results (2) – Outlier Anomalies Belgium

- (1) Some Outliers due to extreme VRE
 - (2) Outliers indicate **low Generation in market**
- => NOT indicated by domestic drivers:

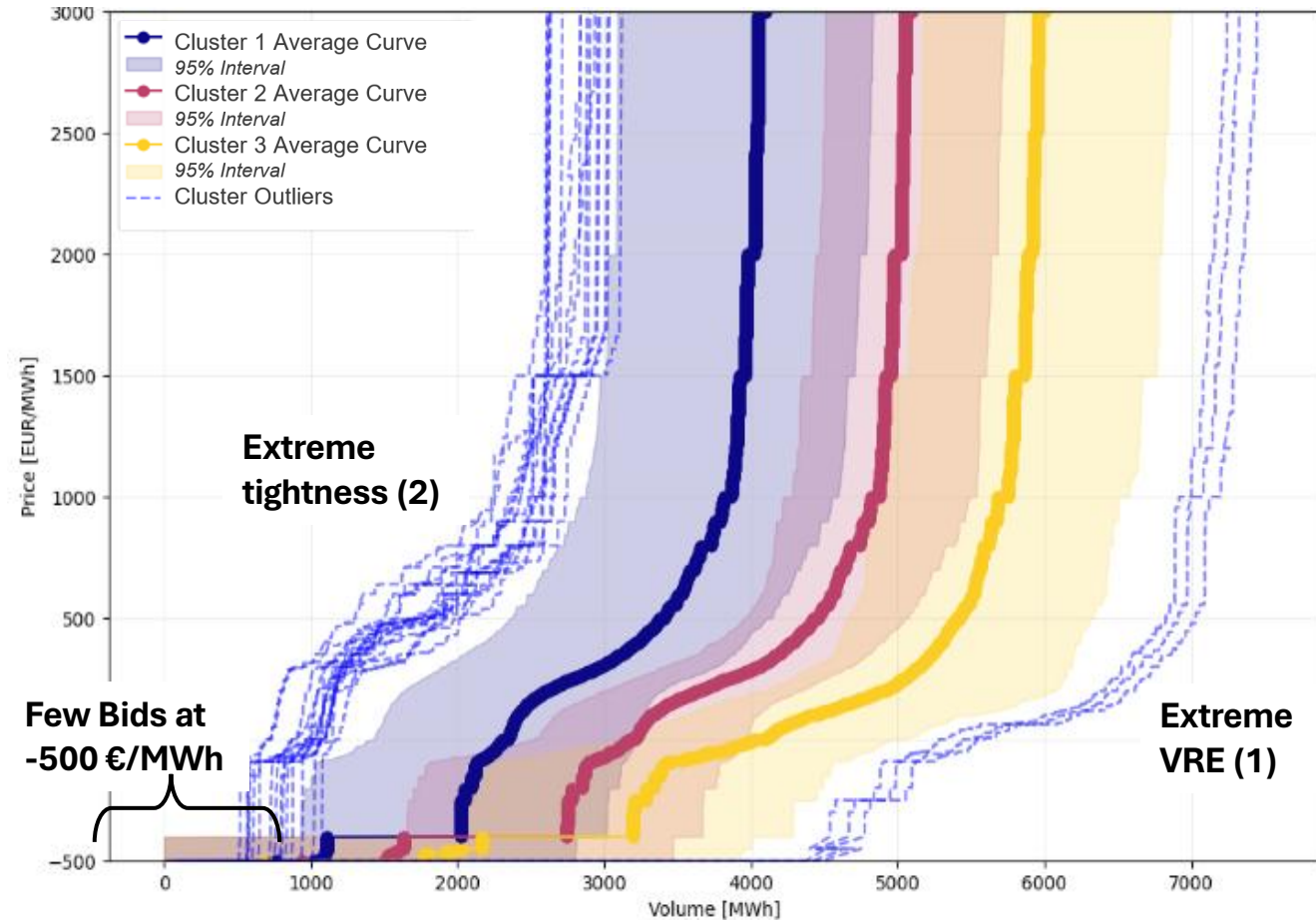
	Temp. °C	Solar MW	Wind MW	Generation MW	Load MW	Price €/MWh
Cluster Outliers (2)	5.68	14	1985	10032	9269	399.3
Cluster 1	12.13	769	1297	10121	9460	44.3

DE & FR drivers give explanation:

	FR Price €/MWh	FR Generation	FR Load	DE Price €/MWh	DE Generation	DE Load
Cluster Outliers (2)	402.27	56886	57833	388.85	51922	53079
Cluster 1	296.84	47574	51169	250.44	56946	56297

**Lack of Generation in France and Germany,
 + high French Load**

- ⇒ Sets high price across zones
- ⇒ Belgium traders bid cross-border



Average Curves and Cluster Outliers for Belgian Market, 2022

Results – Prediction Anomalies

- Cluster mispredictions:
- **Cluster, predicted by drivers \neq Cluster, determined by shape**

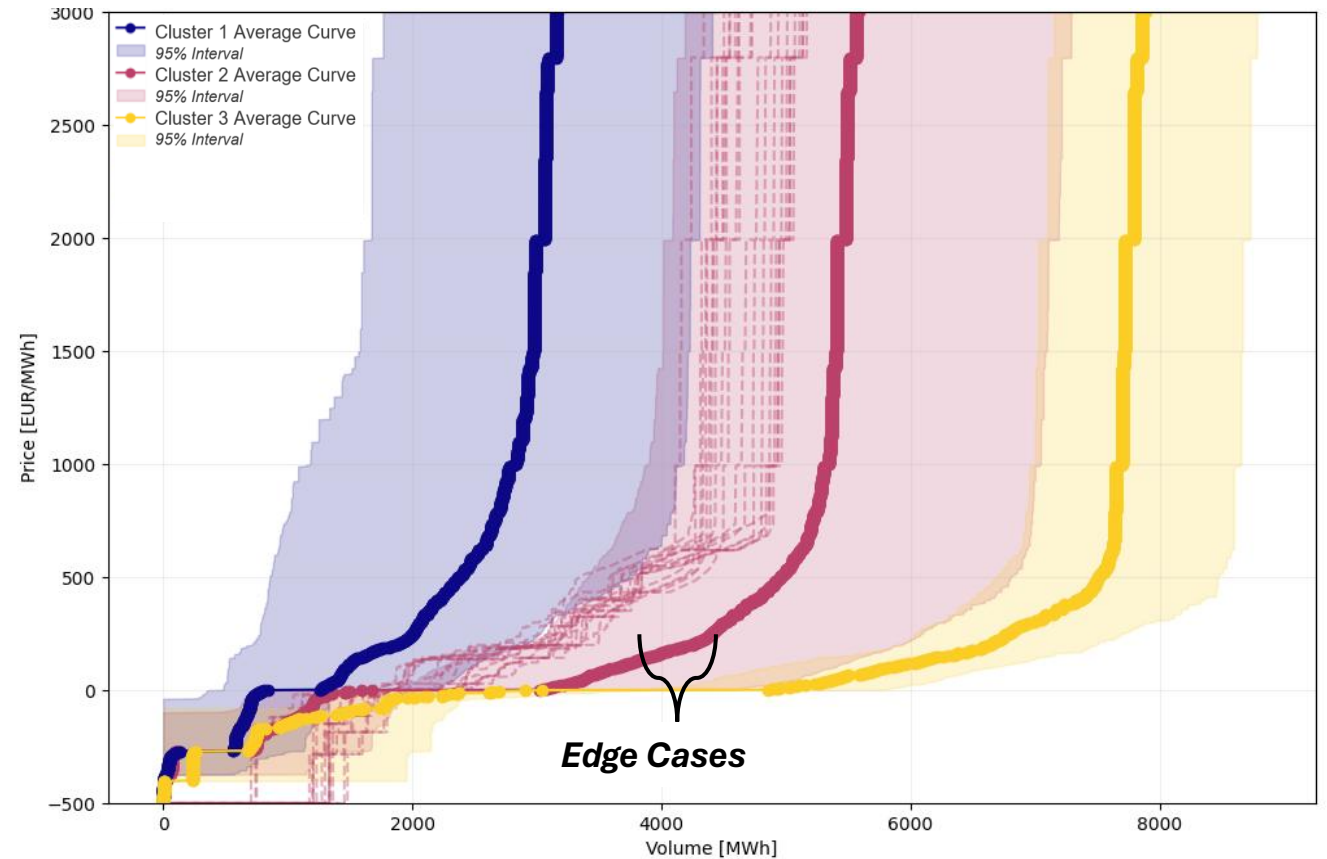
Some can be explained more quickly:

- **Edge Cases:** Cluster 2 Anomalies:

=> Curves assigned **cluster 2 based on shape, predicted cluster 1 based on drivers**

⇒ Curves on the *Edge* between Cluster1 and 2:

Difficult to predict



Edge Case Anomalies for Denmark 1 zone

Results – Prediction Anomalies

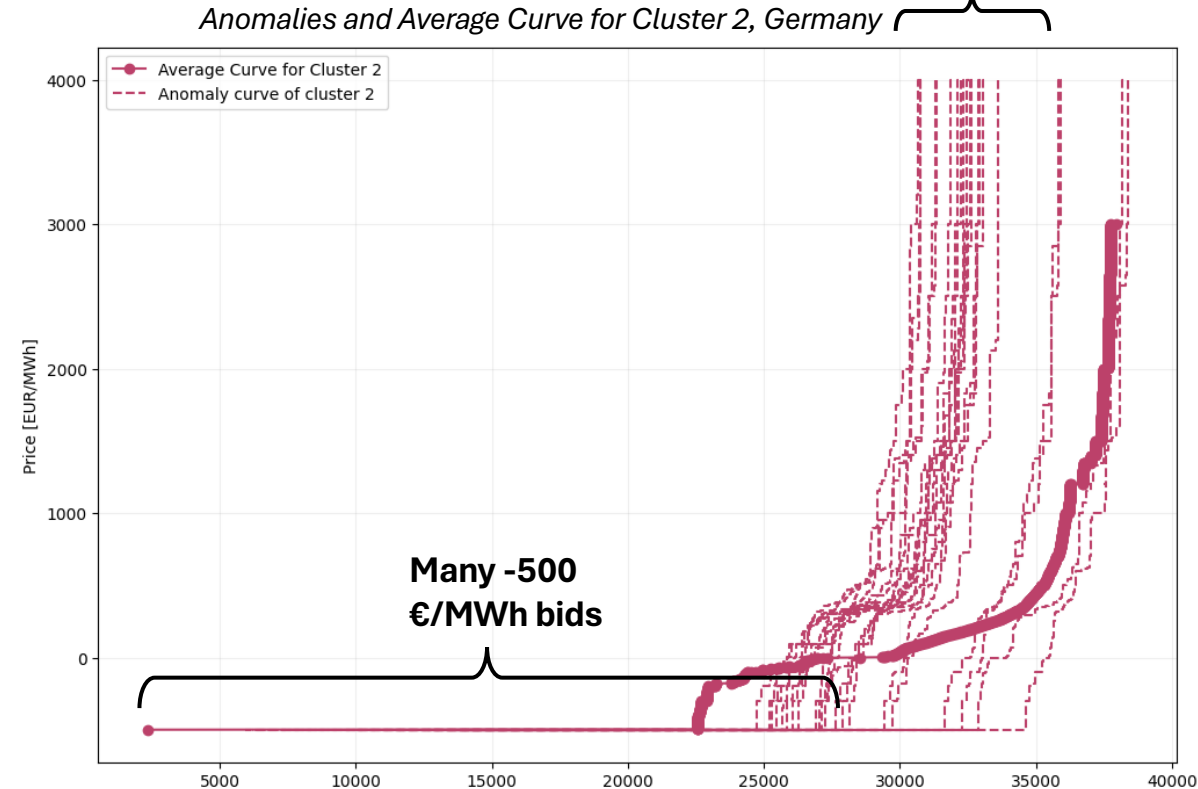
- **Other Anomalies required hands-on Investigation: Germany 22/06**
- Cluster 2 Anomalies: Very different curves from Cluster 2 average

High offered volumes at minimum price \longleftrightarrow Very low flexibility offered in market

	Temp. °C	Solar MW	Wind MW	Generation MW	Load MW	Price €/MWh
Anomalies	18.88	3059	3647	47314	53998	359.1
Cluster 2	14.39	14758	17713	63808	60205	181.4

- Driver data **distinct from Cluster 2**, yet shape closest to Cluster 2 => causes *misprediction*
- **Hypothesis: Three factors coincide:**
 1. **Low flexibility:** High generation unavailability for this period
 2. **Steep curves:** Reduction of gas exports => Gas price increase
 3. **Changes in policy announced: Opening of Coal Plants**
=> Possibly, fulfilling of forward contracts by traders caused many -500 €/MWh Bids? [6]

Low Generation In market



Conclusions

Contributions

From empirical data, our method can:

1. Define *expected spot market bidding*, based on market drivers
2. Detect anomalies from this expected bidding behavior

Applied to 2022: Results

1. Renewables determine the *expected bidding*:
 - Bidding curves differ mostly in bids at -500€/MWh and steepness of curves
 - Significance of Wind vs Solar determined by installed capacity of each
2. Empirical bidding also shows significant anomalies, with varying causes:
 - Extremes in VRE Generation => Extreme bidding curves (Outliers)
 - Unforeseen EU-wide events (Energy Crisis)
 - Cross-Border trading
 - ...

Future Outlook

Expand the Method

- Apply method to other time period/bidding zones
- Implement Intra-day/Forward Market Data to improve analysis of anomalies

Use in ABM?

1. As exogenous input
 - ABMs often use *theorized* bidding curves => **Use Empirical Curves instead:**
 - Depending on driver data
 - The mean Bidding Curve for Cluster 1, 2, or 3, can be used as exogenous input
 2. Empirical Trading strategies:
 - Anomalies might show *empirical trading strategies*, to be implemented in Trading-Agents
- => Our Study can contribute to more accurate spot market modelling

References

- [1] Forschungsstelle für Energiewirtschaft e.V. (FfE). (2022, September 14). *Merit order shifts and their impact on the electricity price*. Forschungsstelle für Energiewirtschaft e.V. Retrieved from <https://www.ffe.de/en/publications/merit-order-shifts-and-their-impact-on-the-electricity-price/>
- [2] Li, Z., Alonso, A. M., Elías, A., & Morales, J. M. (2024). *Clustering and forecasting of day-ahead electricity supply curves using a market-based distance*. International Journal of Electrical Power & Energy Systems, 158, 109977. Retrieved from <https://www.sciencedirect.com/science/article/pii/S0142061524001984>
- [3] EPEX SPOT. (2024). EPEX SPOT - A part of EEX Group. EPEX SPOT. Retrieved from <https://www.epexspot.com/en>
- [4] ENTSO-E. (2017). *ENTSO-E Transparency Platform*. ENTSO-E. Retrieved from <https://transparency.entsoe.eu/>
- [5] Energy Brainpool. (n.d.). *No more woes on the gas market? A review of the current developments*. Energy Brainpool. Retrieved from <https://blog.energybrainpool.com/en/no-more-woes-on-the-gas-market-a-review-of-the-current-developments/>
- [6] EPEX SPOT. (2024) *Future-to-Spot (FTS) Service*. EPEX SPOT. Retrieved from <https://www.epexspot.com/en/glossary#l6>

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