

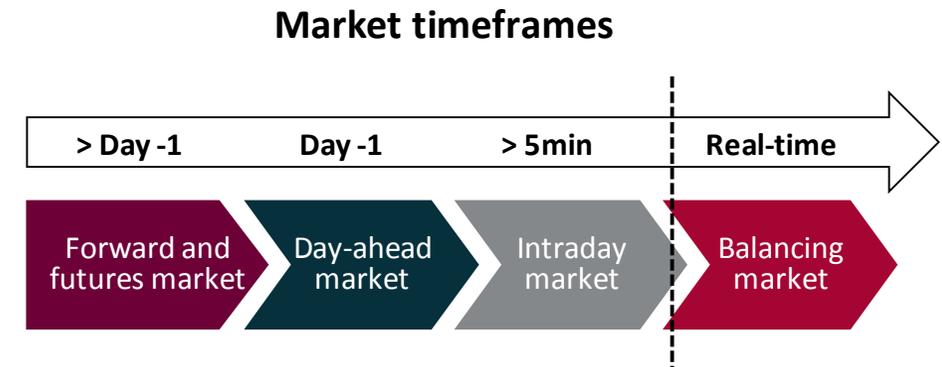
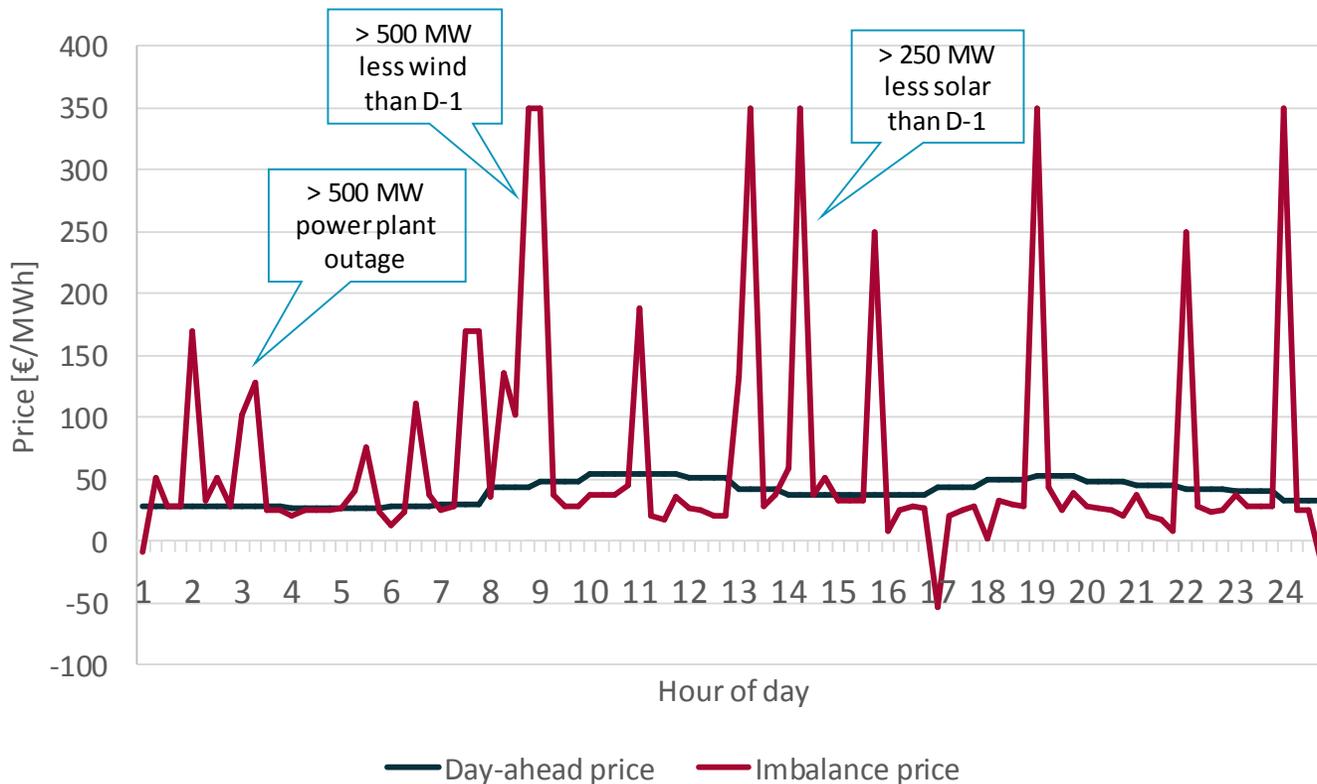
# Benefits of Probabilistic Forecasting in Power Trading

21.01.2020

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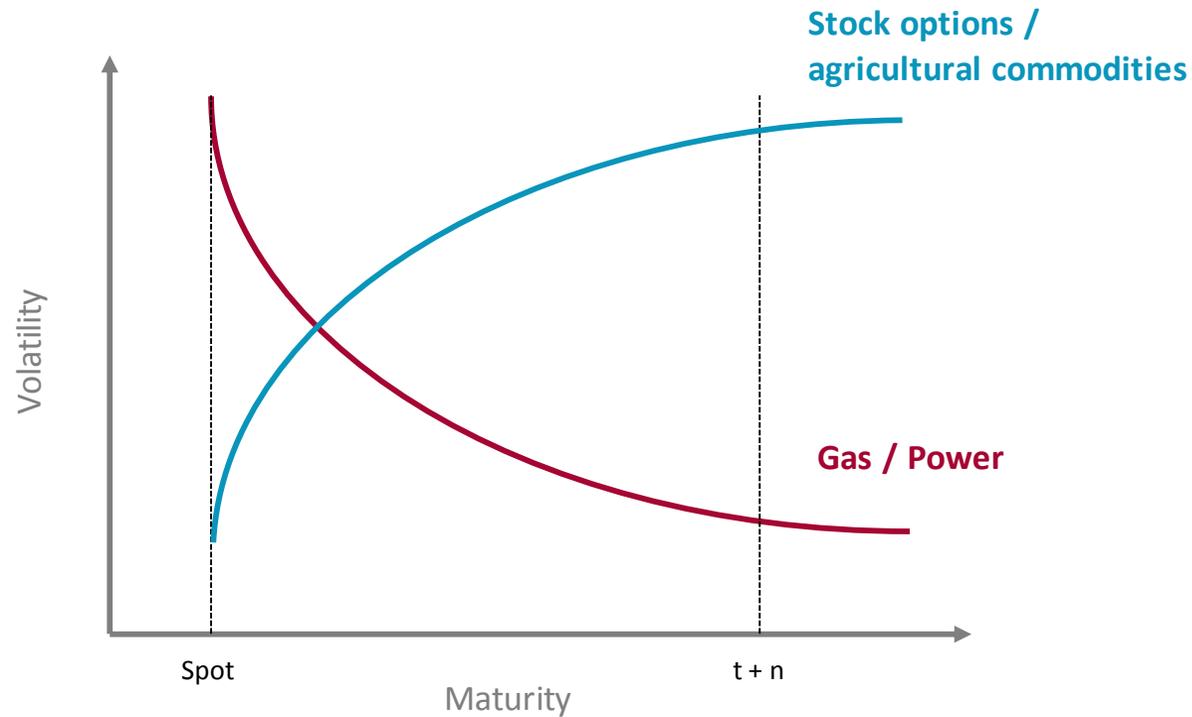
# Day-ahead price vs. imbalance price

In The Netherlands on the 10<sup>th</sup> of January, 2020



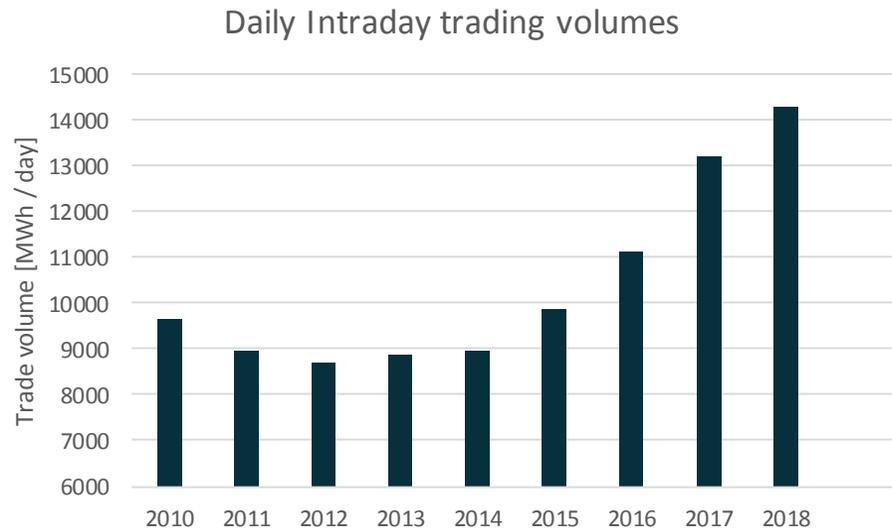
# Price uncertainty in power markets

## Volatility evolution



# Developments in short-term energy markets

## Intraday power market trends



### Main findings (3/3)

Larger imbalance volumes and higher balancing capacity prices.

### Intraday prices

In 2018: larger differences between DA and ID prices

### Intraday trading volumes

Renewables growth and XBID go-live contributed to the further increase of intraday market volumes

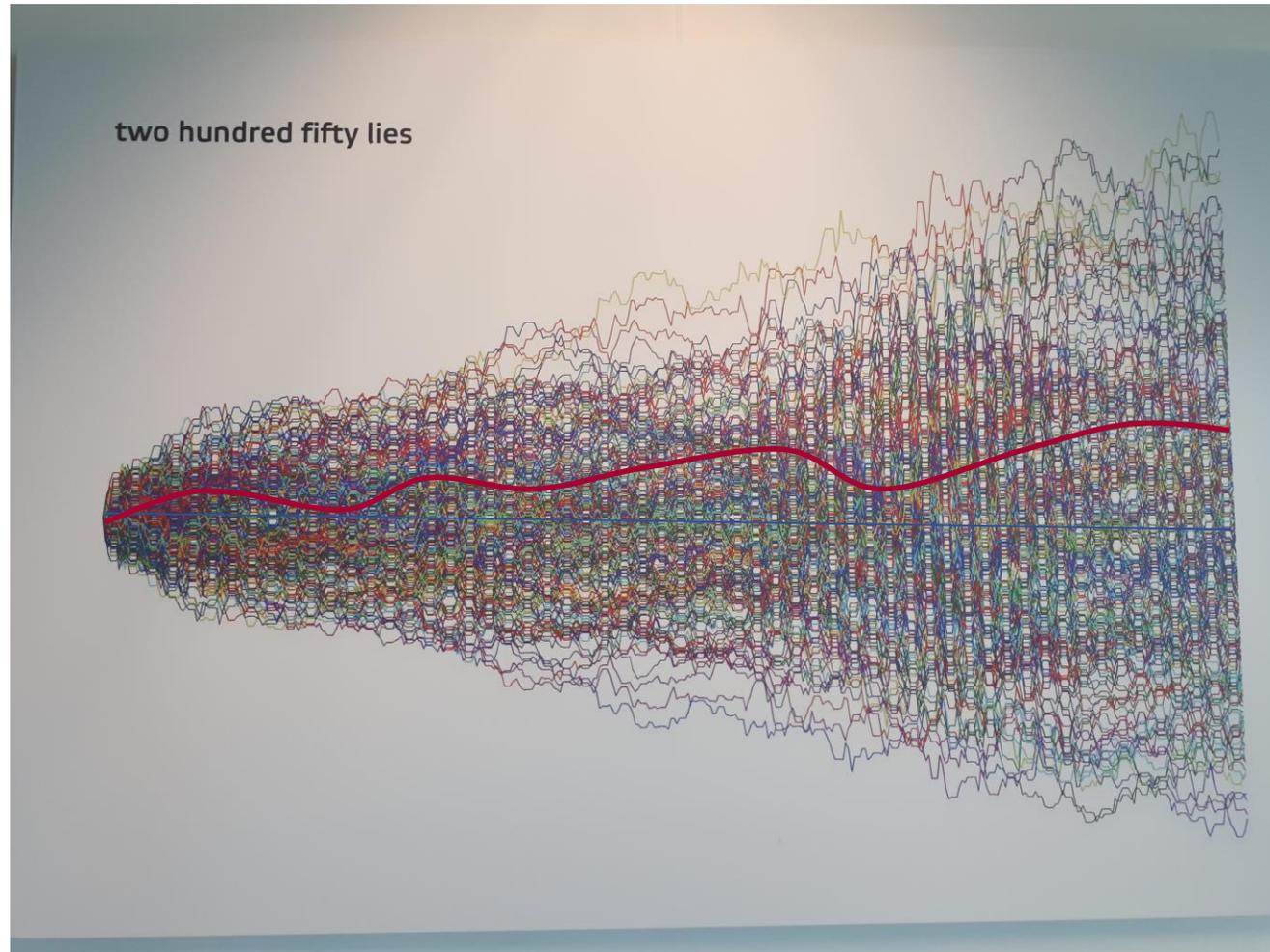
### Net imbalance volumes NL

Trend of larger imbalance volumes continues



The short-term power market in **the Netherlands** reached a combined 38.7 TWh (2018: 39.5), with a **55.7% growth** rate on the Intraday market which reached 3.3 TWh (2018: 2.1 TWh).

# How to manage the uncertainty?



**Common approach:**  
Mean absolute or mean squared error optimized point-forecast

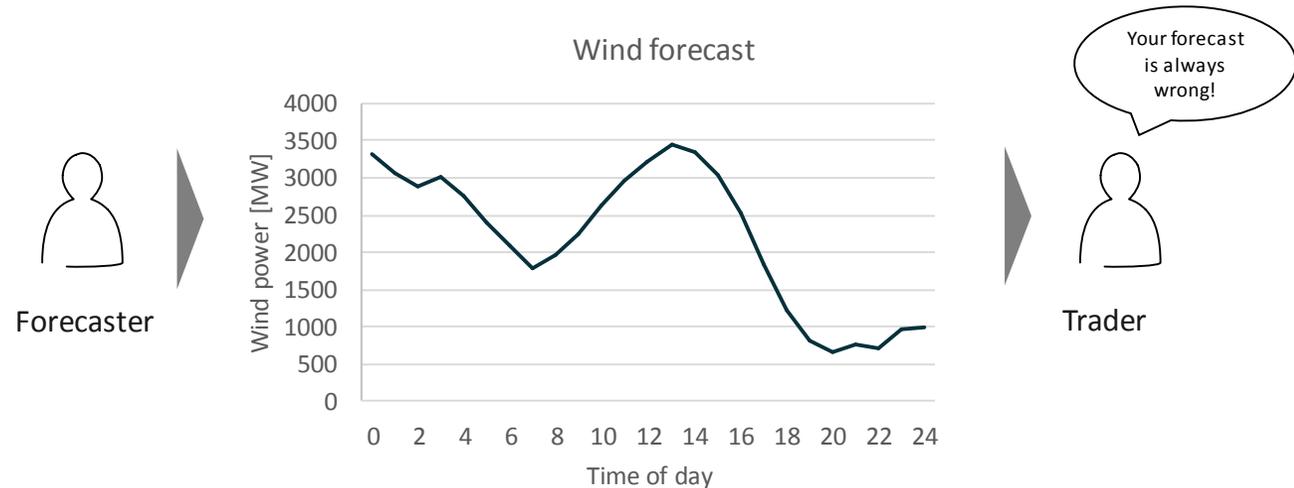
# How to manage the uncertainty?

## Misconceptions:

- Point-forecasts are often given too much confidence
  - “Your forecast is always wrong” implies the expectation that the forecast needs to be “right”
- The measures of forecast error, whether it be MAPE, WMAPE, MAE, RMSE or any similar metric, all assume that the perfect forecast can be expressed as a single number

## How can we do better?

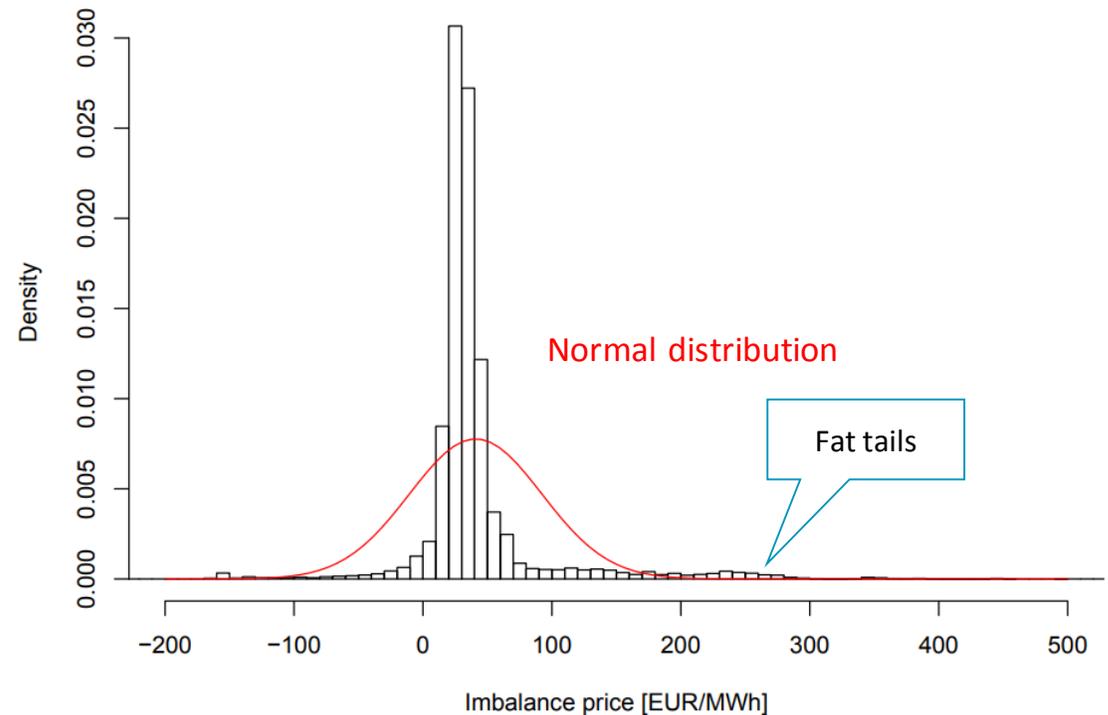
- Transparency
  - No “black box” models
  - What is the Input/output data?
- Uncertainty
  - While we can’t predict wind power production perfectly due to its inherent weather variability, we can predict the weather variability
  - Maximizing the likelihood of a correct prediction rather than minimizing the imbalance / variance



# Volume $\neq$ Value

- Forecasts optimized on error metrics like MAE, RMSE aim to reduce the imbalance volume / the variance
- When selling 1 MW throughout whole 2019 in the Dutch Imbalance Market, **50%** of the profit is generated **in just 20% of the time**
- Imbalance price distribution is asymmetrical
  - Mean = 40.87 EUR
  - Median = 31.46 EUR
- Imbalance price distribution has fat tails -> the extreme values have the biggest impact

Dutch imbalance prices in 2019



# Volume $\neq$ Value

“...‘If you go out in the evening with a shotgun and see two flocks of ducks, and you aim for the least-squares location, you will go home hungry’ Miller says. The maximum-likelihood approach is to aim for one of the flocks instead...”

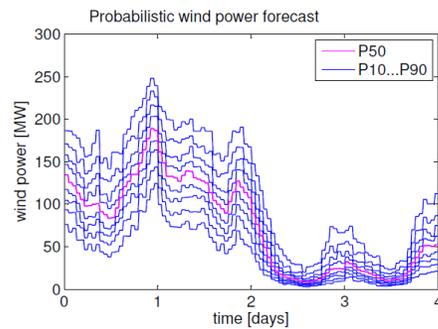
Source: “Ensemble Kalman Filters Bring Weather Models Up to Date”, Dana Mackenzie, SIAM News, Volume 36, Number 8, October 2003

# Probabilistic forecasting methods

## Physical

- Numerical weather prediction ensembles

e.g. ECMWF ensemble forecast:



- Uncalibrated
- Not valid for very short-term

## Statistical

### Estimation of quantiles

- Quantile regression
  - No assumptions about the distribution of the residuals

- Preventing quantile crossing can become difficult

### Prediction intervals

- Range of coverage probabilities assuming normally distributed forecast errors

- Uncertainty expressed in one number
- Symmetric distribution

### Full density functions

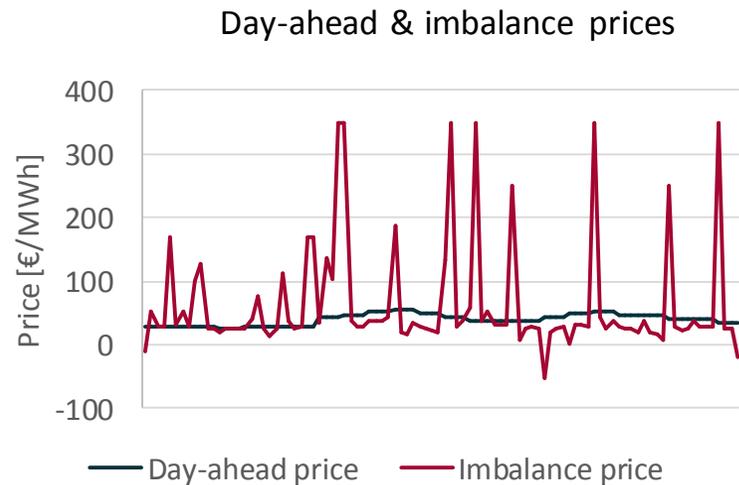
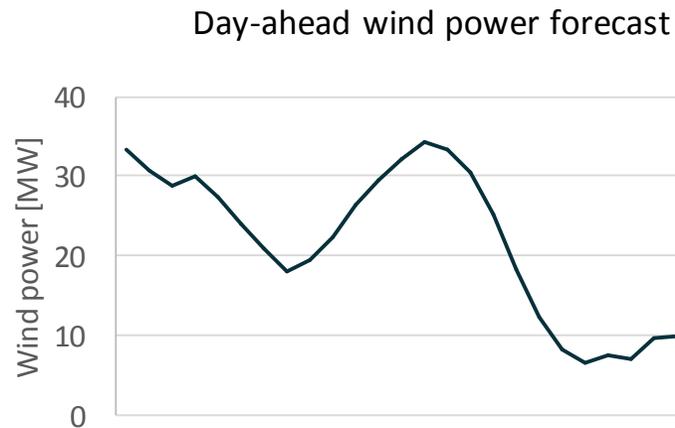
- Parametric
  - E.g. process is assumed to be Gaussian
- Non-parametric
  - Kernel density estimation
  - Analog Ensembles
  - Bayesian methods

- Can be computationally expensive

# Probabilistic forecasting methods

## Optimized bids of a Belgium Wind farm using stochastic optimization models

Input data:

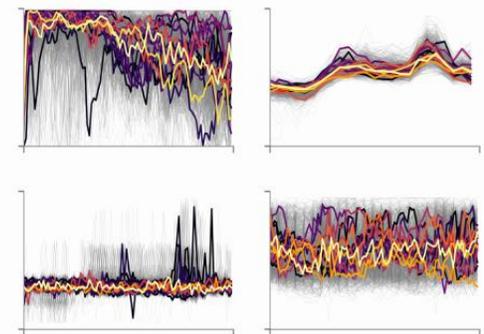


Gerrit Deen  
Researcher at Whiffle

Increasing the Market Value of Wind Power Using Improved Stochastic Process Modeling and Optimization

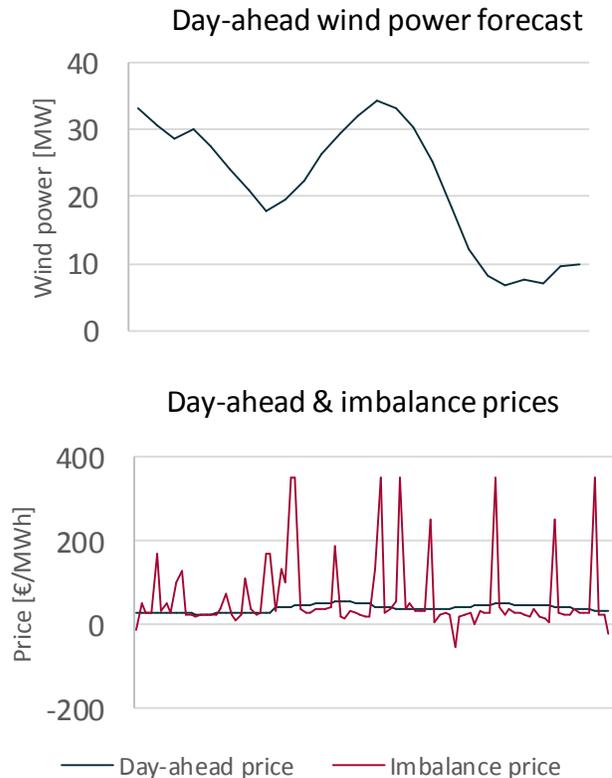
A Case Study of a Belgian Wind Power Producer

G.J. Deen

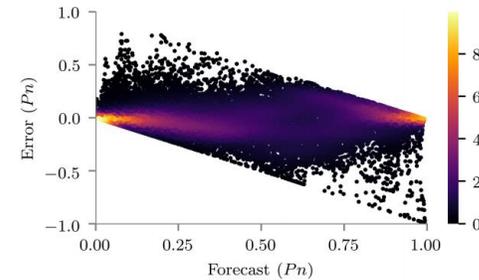


# Stochastic Process Modeling

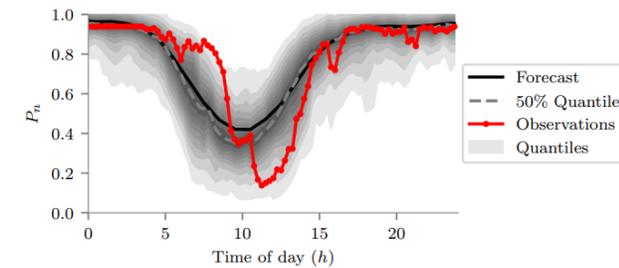
**Approach:** model forecast errors as stochastic processes using conditional Kernel Density (CKD) estimation



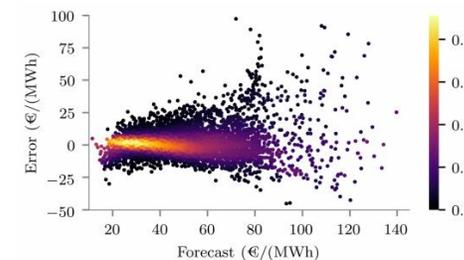
CKD wind forecast error vs. load factor



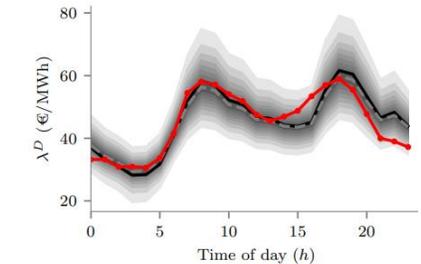
Quantile wind forecast



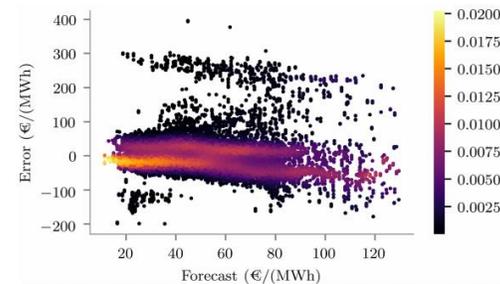
CKD D-1 price forecast error vs. price



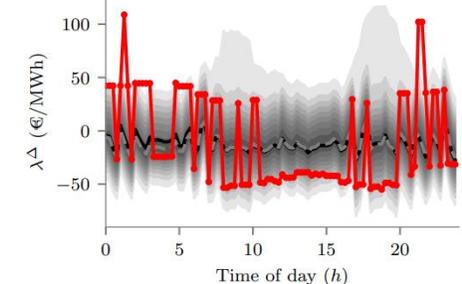
Quantile D-1 price forecast



CKD imbalance price forecast error vs. price



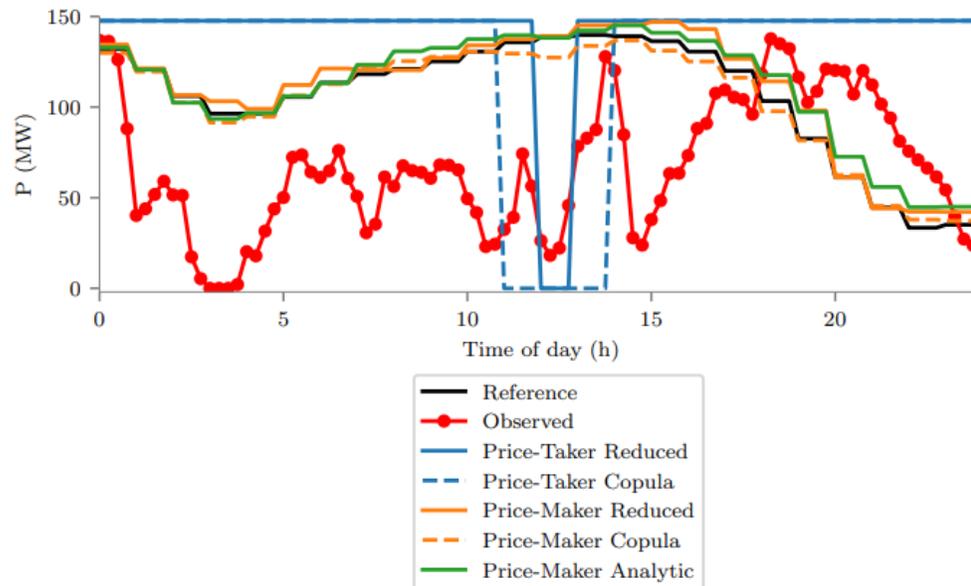
Quantile imbalance price forecast



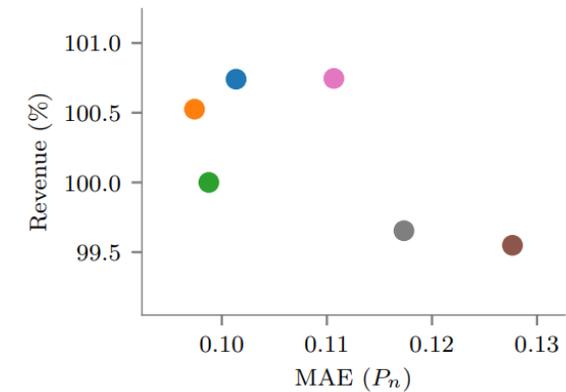
# Optimized bids of a Belgium Wind farm

## Results

Optimized bids



Forecast error vs. Revenue



### Conclusions:

- Higher MAE & higher revenue (-> Volume  $\neq$  Value)
- Reduced total system costs
- Reduced risk of extreme system imbalance

Thank you!

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