

## Motivation

We can help tackle **climate change** by a more effective use of renewable energy sources. The goal of my research is to improve existing methods of stochastic modelling and statistical inference to quantify risk and uncertainty of **renewable energy sources** in a more reliable way.



Figure: Wind farm.

## Electricity features

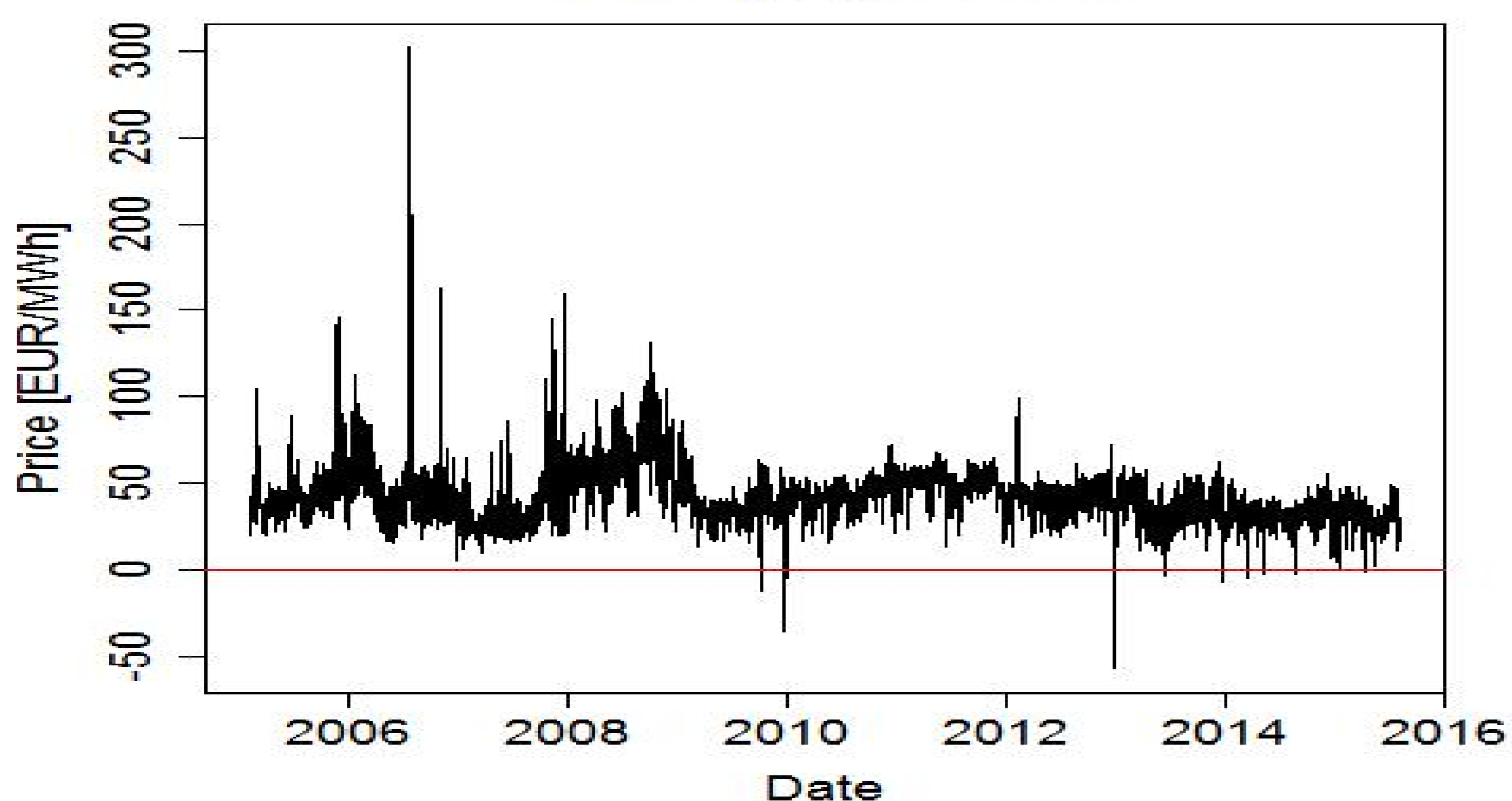
Since electricity is generally traded for consumption, it is considered a commodity. However, in contrast to other types of commodities, it has some unique features ([1]).

- Non-storability (supply and demand must always match).
- Seasonality (higher demand in winter months due to the need of heating and longer use of lights).
- Periodic behaviour (higher demand in the peak time, i.e., Monday to Friday between 8 am and 8 pm).
- Mean reversion (over time the electricity prices will tend to their average).
- Large and heteroscedastic volatility.

## Empirical data

I work with a set of German data consisting of daily prices of spot and futures contracts over about 10 years.

Electricity spot prices



Front month futures

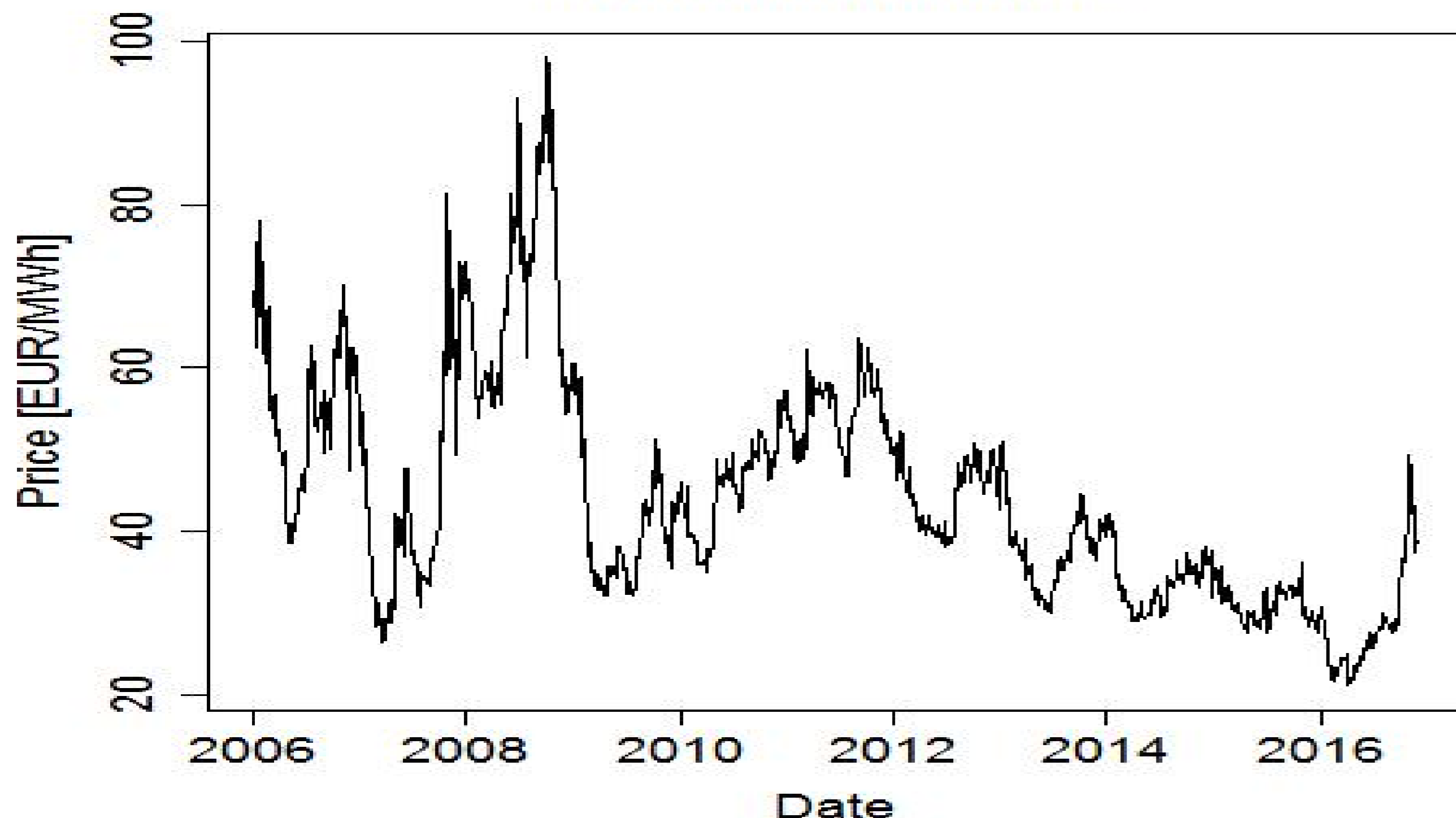


Figure: Data for German electricity spot and futures prices.

## The arithmetic model for spot prices

Let  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \in \mathbb{R}}, \mathbb{P})$  be a probability space, where the filtration  $\{\mathcal{F}_t\}_{t \in \mathbb{R}}$  satisfies the 'usual conditions'. Let  $S(t)$  be the spot price. Following [2], I proposed the arithmetic model

$$S(t) = \Lambda(t) + Z(t) + Y(t),$$

where  $\Lambda(t) + Z(t)$  is the long-term factor, while  $Y(t)$  describes the short-term behaviour, which includes the **impact of renewables**.

## Model terms

- $\Lambda(t)$  – a deterministic seasonality/trend function.
- $Z(t)$  – a Lévy process with zero mean (under the physical measure).
- $Y(t) = \int_{-\infty}^t g(t-s)\sigma_s dL_s$  with a kernel  $g(t-s)$  such that  $\lim_{t \rightarrow \infty} g(t-s) = 0$ .

Here  $\sigma_t$  is a càdlàg stochastic process describing the volatility of  $Y(t)$ . Similarly to [3], I defined it as

$$\sigma_t = \int_{-\infty}^t j(t-s)dV_s,$$

where  $j$  is a deterministic, positive function and  $V(t)$  – a Lévy subordinator. I assumed that  $\sigma_t$  is independent from the driving Lévy process  $L(t)$ .

## The arithmetic model for futures prices

By the no-arbitrage arguments, one can define the price of a futures contract with maturity  $T$  as

$$f(t, T) = \mathbb{E}_{\mathbb{Q}}[S(T)|\mathcal{F}_t],$$

where  $0 \leq t \leq T < \infty$  and  $\mathbb{Q}$  is a risk neutral probability measure (see eg. [2]).

In the arithmetic case the forward price at the time  $t$  equals

$$f_t(T) = \Lambda(T) + Z(t) + (T-t)\mathbb{E}_{\mathbb{Q}}[Z(1)] + \int_{-\infty}^t g(T-s)\sigma_s dL_s + \mathbb{E}_{\mathbb{Q}}[L_1] \int_t^T g(T-s)\mathbb{E}_{\mathbb{Q}}[\sigma_s|\mathcal{F}_t] ds.$$

In the long run futures prices can be approximated by

$$f_t(T) \approx \Lambda(T) + Z(t) + (T-t)\mathbb{E}_{\mathbb{Q}}[Z(1)] + \mathbb{E}_{\mathbb{Q}}[L_1] \frac{\mathbb{E}_{\mathbb{Q}}[V_1]}{\delta} \int_0^{\infty} g(y)dy.$$

## Current and future work

The most important step is to fit the proposed model to the empirical data in order to find the most appropriate form of  $\Lambda(t)$ ,  $Z(t)$  and  $Y(t)$ . I am especially interested in the **impact of renewables** on electricity prices. Furthermore, I am going to compare the results to an alternative approach, in which one models futures prices directly using ambit processes (see eg. [4]).

## References

- R. ter Haar, "On modelling the electricity futures curve," 2010.
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- O. E. Barndorff-Nielsen, F. E. Benth, and A. E. D. Veraart, "Modelling energy spot prices by volatility modulated Lévy-driven Volterra processes," *Bernoulli*, vol. 19, no. 3, pp. 803–845, 2013.
- F. Barndorff-Nielsen, O. E. Benth and A. Veraart, "Modelling Electricity Futures by Ambit Fields," *Advances in Applied Probability*, vol. 745, no. 46(3), pp. 719–745, 2014.