Uncertainty in Wind Power production

Reliable wind power generation forecasting is crucial to:

- ▶ Meet energy demand through renewable power sources.
- Energy trading of future excess power.
- Design of Investment strategies.

We would like to:

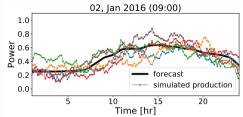
- Simulate forecast error and quantify its uncertainty.
- ► Calibrate a forecast model for optimal dispatch of electric power.

Our model is based on:

- Parametric Stochastic Differential Equations.
- ▶ Continuous optimization of approximate Maximum Likelihood.

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Prediction error of Wind Power production can be modeled by a parametric SDE,



$$dV_t = - heta_t V_t \ dt + \sqrt{2 heta_t lpha(V_t + p_t)(1 - V_t - p_t)} dW_t$$
 $V_0 = v_0$ 1.0 0.8 where V_t is an error stochastic pro-

where V_t is an error stochastic process, p_t is a given wind power forecast and (α, θ) are parameters to the model

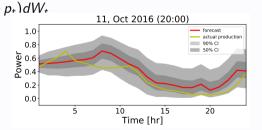


Figure 1: Example of Wind power production simulation and confidence intervals.

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Parameters of the model are inferred using optimization techniques of approximate transition likelihood functions.

$$\mathcal{L}(\boldsymbol{\theta}; V) = \prod_{j=1}^{M} \prod_{i=1}^{N} \rho(V_{j,i+1}|V_{j,i}, \boldsymbol{\theta}) \rho(V_{j,0})$$

Where $V^{M,N} = \{V_{t_1^{M,N}}, V_{t_2^{M,N}}, \dots, V_{t_N^{M,N}}\}$ is a set of M paths with N observations observed in intervals of time Δ_N .

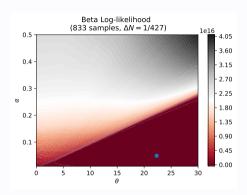


Figure 2: Likelihood of sample paths. Point of optimality shown in blue.