

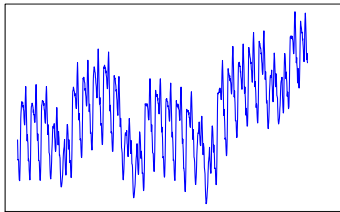
Modelling, Forecasting for Renewable Energy Production and Statistical Inference

Statistical and machine learning methods
to model and forecast Energy

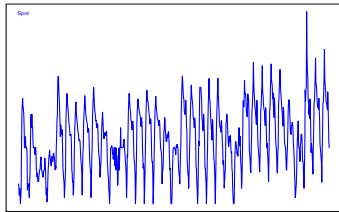
Mathilde Mougeot

ENSIIE & LPSM

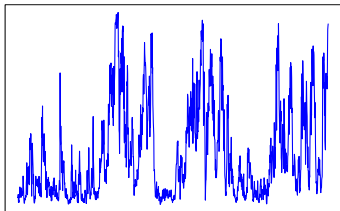
Thematic semester on Statistics for Energy markets
Dourdan, June 2018



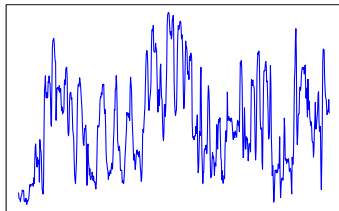
French electrical consumption



Energy Spot prices



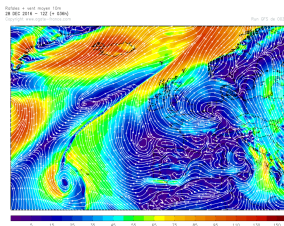
Wind turbine power



Industrial equipment

Physical Modelling vs Statistical Modelling

- Physical Modelling mostly based on physical equations :
 - Simple Direct equation
 $PV = nRT$
 - Partial differential equations
→ Need of a numerical model to study the dynamic and the evolution of a model. Ex : Navier-Stokes equation



→ Need to well understand the underlying phenomena

Statistical Modelling

- In a supervised setting (Y, X) :

$$Y \in \mathbb{R}, X \in \mathbb{R}^p$$

$$Y = \mathcal{M}(X) + \epsilon$$

- Parametric models.

$$\mathcal{M}_1(X) : Y = \beta_0 + \beta_1 X + \beta_2 X^2 + \epsilon$$

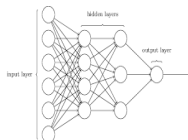
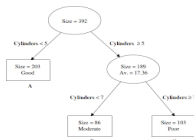
$$\mathcal{M}_2(X) : Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_p X_p + \epsilon$$

→ Probabilistic or not, depending on the law assumption on ϵ .

Data Set : n observations : (y_i, x_i)

→ to estimate (to compute) the parameters $\hat{\beta}$ of the model

- Non Parametric models No analytical (explicit) expression

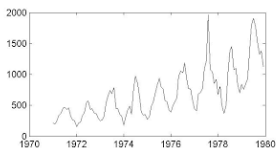


→ Models : Kernels, decision trees, neural networks(black boxes),...

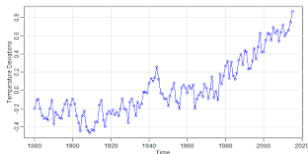
Crucial question : How to find the correct model ?

Time Series Modelling

- Time dependency and Patterns



- Without (iid?) :



$$Y_t = \mathcal{M}(X_t) + \epsilon_t$$

How to find the best model ?

- Bunch of p unitary, calibrated models

$\mathcal{M}_1, \dots, \mathcal{M}_p$: Parametric, Non Parametric models...

$\rightarrow \hat{\mathcal{M}}_1, \dots, \hat{\mathcal{M}}_p$: estimated model using Train data
 $\{(y_i, x_i) | 1 \leq i \leq n\}$.

- Selection vs Aggregation

1 Model selection

$\mathcal{M}_{opt} = \arg \min_{k \in \mathcal{P}} E_{\mathcal{M}_k}$, $E_{\mathcal{M}_k}$ is a cost function to evaluate \mathcal{M}_k

$\hat{\mathcal{M}}_{opt} = \arg \min_{k \in \mathcal{P}} E_{\hat{\mathcal{M}}_k}$

2 Aggregation

$\hat{\mathcal{M}}_{opt}(X) = \sum_{k=1}^p \hat{w}_k \hat{\mathcal{M}}_k(X)$ Exponential weight aggregation

$\hat{\mathcal{M}}_{opt}(X) = \sum_{i=1}^n \hat{w}_{n,i}^{\hat{\mathcal{M}}_{1..p}(X_i)}(X) Y_i^{Train}$ Consensual aggregation

Electricity Framework



Production

Nuclear power plants
coal fired power plants
wind farms
photovoltaic farms

Consumption

industrial plants
home and heating
building heating
...

→ Electricity can hardly be stored. There is a need to :
Balance between electrical production and consumption
Forecast consumption and production

Statistical & ML models for Energy in several directions.

① Consumption

- High dimensional regression models to Forecast the French National Consumption with D. Picard, K. Tribouley, V. Lefieux, JRSSB, AADA.
- Energy Savings. Over consumption monitoring and diagnostic for industrial equipments with O. Cadet, C. Harper.

② Production

- Modeling the Wind Farm production using machine learning tools with A. Fischer, L. Montuelle, D. Picard, Wind Energy 2017.
- From Numerical Weather Prediction outputs to accurate local surface Wind speed : statistical modelling and forecasts with B. Alonzo, R. Plougonven, A. Fischer, A. Dupre, & P. Drobinski.

③ Meteorological inputs

Homogeneous climate regions using learning algorithms.
with D. Picard, V. Lefieux, M. Marchand.

Outline of the lessons

- ① Functional models to forecast electrical consumption.
Statistical models.
Exponential weight Aggregation
- ② Machine learning models to model wind turbine production.
Machine learning models.
Consensual Aggregation.
- ③ Segmentation of multidimensional functional data.
Unsupervised learning.
Spectral clustering.
Aggregation of Similar clustering instances.

The never ending story...to answer to real data problems

