

RW-Turb kick off meeting

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Universal Multifractal framework



- Wind power (available and produced) shows a highly non-Gaussian distribution, where intensity depends upon the scale of measurement.
- The framework of Universal Multifractals (UM) enables to characterize the extreme variability of geophysical fields across scales, with the help of a limited number of parameters with physical meaning (Schertzer and Lovejoy (1987)).

$$p(\boldsymbol{\varepsilon}_{\boldsymbol{\lambda}} \geq \boldsymbol{\lambda}^{\boldsymbol{\gamma}}) pprox \boldsymbol{\lambda}^{-c(\boldsymbol{\gamma})}$$

 $egin{aligned} &\langle m{arepsilon}_{\lambda}{}^{q}
angle pprox \lambda^{K(q)} \ &K_{c}(q) = rac{C_{1}}{lpha - 1}ig(q^{lpha} - qig) \end{aligned}$

Statistical moments (q) of a scale invariant field (ϵ) scale with resolution ($\lambda)$

Moment scaling function (K(q)) characterized using only two parameters,

multi-fractality index α and mean intermittency co-dimension C₁.

• Using UM, we try to characterize variability of wind power and its correlation with other highly variable geophysical fields such as wind velocity, rainfall rate and air density.

Data analysis: inputs and methods



Available wind power

 $P_w = \frac{1}{2}\rho A v^3 C_p$

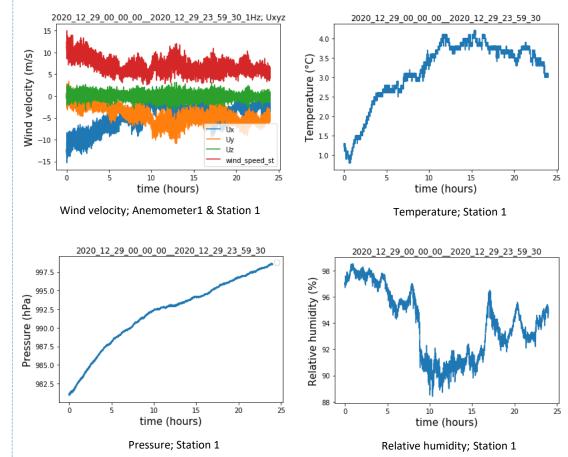
where, ρ - air density (kg/m^3) A - rotor swept area (6,362 m^2 for Vestas-90) V - wind velocity (m/s; from Anemometers) C_p - Betz coefficient

Air density of moist air, CIPM-2007

$$\rho(t,p,h) = \frac{pM_a}{Z(t,p,h)RT(t)} \left\{ 1 - x_v(t,p,h) \left[1 - \frac{M_v}{M_a} \right] \right\}$$

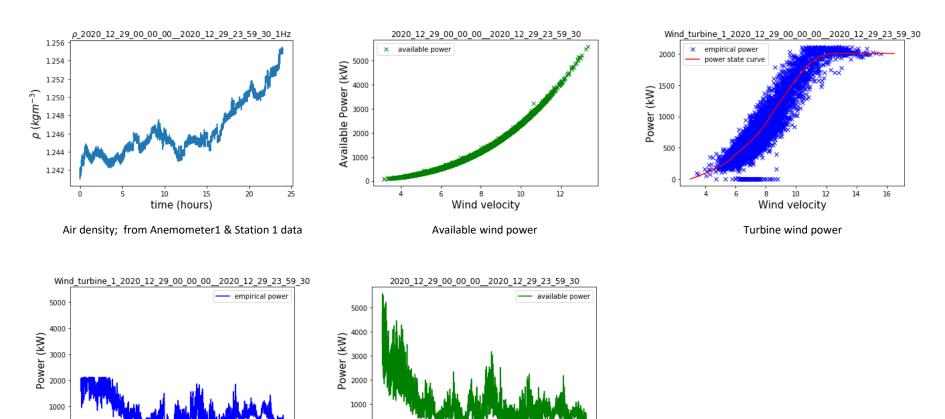
t is temperature (°*C*; from Meteorological station) *p* is pressure (*Pa*; from Meteorological station) *h* is humidity ($0 \le h \le 1$; from Meteorological station) *T*(*t*) is air temperature (K; from *t*)

Z is compressibility factor R is molar gas constant $(Jmol^{-1}K^{-1})$ x_v is mole fraction of water vapour M_a is molar mass of dry air $(gmol^{-1})$ M_v is molar mass of water $(gmol^{-1})$



Data analysis: outputs





Available wind power, time series

time (hours)

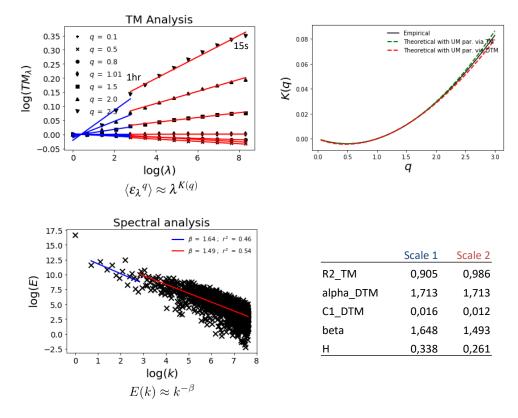
time (h) Power produced by Turbine, time series

Initial MF analysis



Ensemble analysis of Power produced:

- Ensemble analysis using the longest available time data to have an idea on average behaviour of field
- Negative wind power values were set to 0
- Wind turbine power time series shows 2 scaling regimes, whether its needed should be investigated
- Effect of upper threshold (2000kW) might create some bias (needs investigation)
- Analysis need to be refined according to rainfall conditions, will be repeated based on rainfall

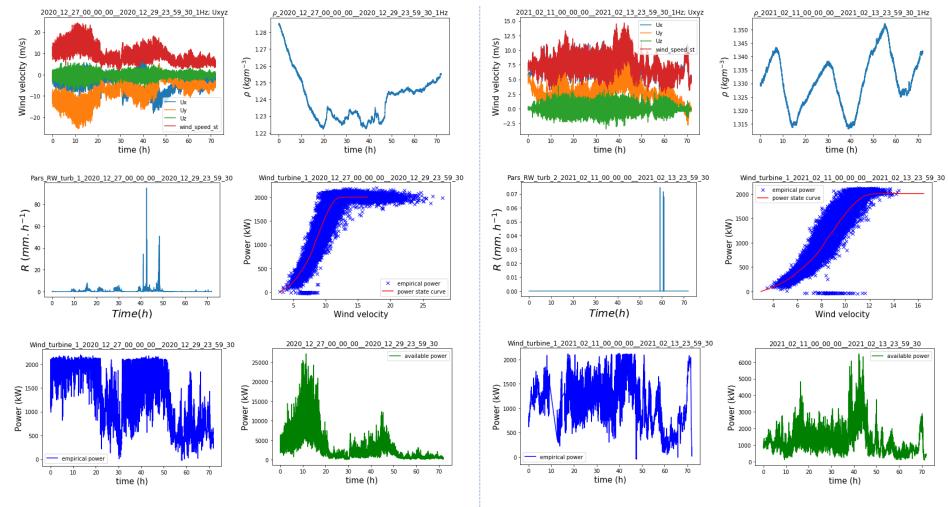


Time period – 12 Dec 2020 to 14 Feb 2021 Data – Wind turbine #1

Sample length – 4096 (close to 1 day) No. of samples – 67

Rainy days

Dry days



Rainy days: 27 to 29 Dec 2020;

Dry days: 11 to 13 Feb 2021

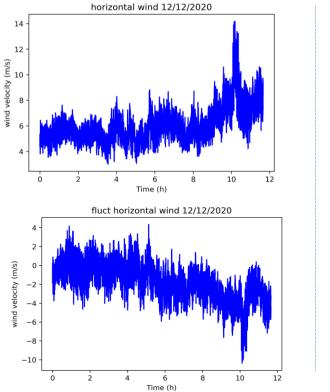


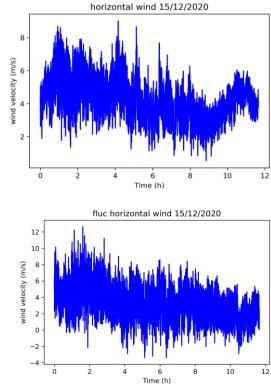
Outline

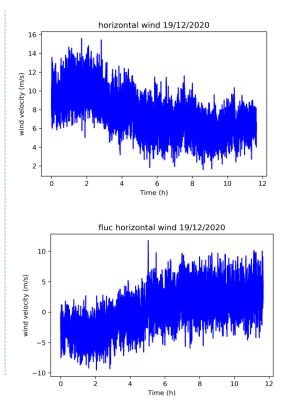
- Main purpose, develop modelling chain
- Bibliography to understand universal multifractal (UM) framework
- Analysis of atmospheric fields
- Compute simulations



Data 100hz Anemometer



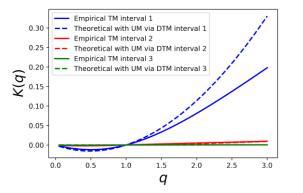






Preliminary Analysis

- Ensemble of 7 rainy days
- Spectral Analysis in horizontal wind
- High values of H
- UM analysis in fluctuations



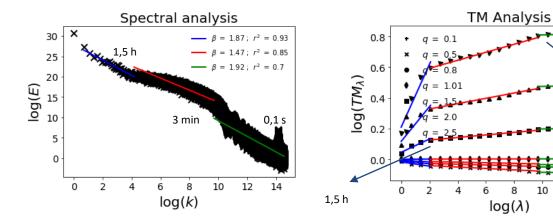
3 min

12

14

16

	Blue interval	red interval	green interval
alpha	1.30	1.75	1.84
c1	0.003	0.02	0.03
beta	2.05	1.33	1.51
Н	0.52	0.19	0.28



	r2		
q	blue interval	red interval	green interval
1,5	0.57	0.98	0.98



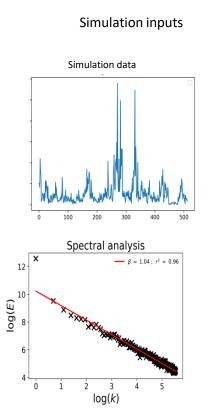
Fractional Integration/Differentiation

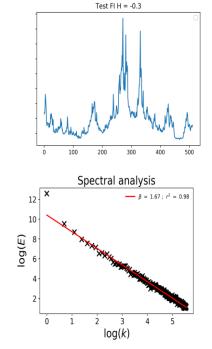
alpha=1.89

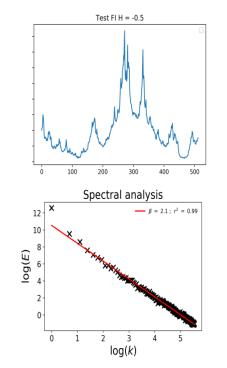
c1=0.07

Samples=100

n=9





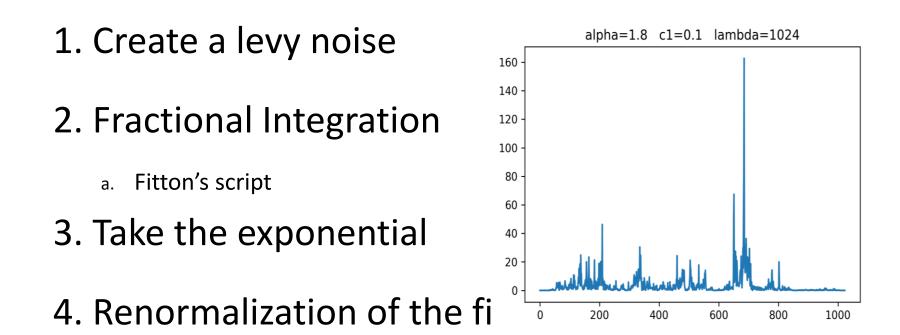


 $E(k)\approx k^{-\beta}$

 $\beta = 1 + 2H - K(2)$



Simulation, UM continuous cascades



Simulations

How to manage negative fields $Data = RE(e^{\log X + (\log Y)j})$

Replicate analysis results

Using parameters recovered in analysis

Improve the simulations

Vector fields

Different rainfall conditions

