# A multifractal-based climate analysis

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# Hölder regularity

- Hölder exponent
- Spectrum of singularities
- Wavelet leaders method (WLM)

# Application to surface air temperature signals

- Data description and first results
- Hölder spaces-based classification and blind test
- Discussion and conclusions

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# Definition

Let f be a signal and  $x_0$  a real number. Then f belongs to the Hölder space  $C^{\alpha}(x_0)$  if there exists a polynomial  $P_{x_0,\alpha}$  of degree at most  $\alpha$ , a positive constant *C* and a neighborhood  $V_{x_0}$  of  $x_0$  satisfying

$$|f(x)- extsf{P}_{x_0,lpha}(x)|\leq C|x-x_0|^lpha$$

for all  $x \in V_{x_0}$ .

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# Definition

The Hölder exponent  $h(x_0)$  of f at  $x_0$  is defined as the supremum of the exponents  $\alpha$  such that f belongs to  $C^{\alpha}(x_0)$ :

$$h(x_0) = \sup\{\alpha : f \in C^{\alpha}(x_0)\}.$$

- Hölder exponent changes from point to point : f multifractal
- Constant Hölder exponent : f monofractal, i.e. f is regularly irregular
- Example of a monofractal function : fractional Brownian motion



Fractional Brownian motions with Hölder exponents 0.2, 0.4, 0.6 almost surely.

How to characterize the global regularity of a signal ?

## Definition

The spectrum of singularities of f is the Hausdorff dimension of the set of points sharing the same Hölder exponent :

$$d_f: h \mapsto \dim_{\mathcal{H}}(\{x_0 \in \mathbb{R}: h(x_0) = h\}),$$

where dim<sub> $\mathcal{H}$ </sub>(*X*) denotes the Hausdorff dimension of the set *X*.

Corollary : *f* is monofractal if and only if its spectrum of singularities is reduced to a single point.

1) Wavelet decomposition of the signal :

$$f(x) = \sum_{j,k \in \mathbb{Z}} c_{j,k} \psi(2^j x - k) = \sum_{\lambda \in \Lambda} c_{\lambda} \psi_{\lambda}$$

where  $\psi$  is a wavelet and  $c_{j,k}$  is the wavelet coefficient associated to the dyadic interval  $\lambda$  at scale *j* and position *k* :

$$\lambda = \lambda_{j,k} = [2^{-j}k, 2^{-j}(k+1)]$$

and

$$c_{j,k}=2^{j}\int_{\mathbb{R}}f(x)\psi(2^{j}x-k)dx.$$

2) For each  $\lambda$ , compute the wavelet leaders

$$d_{\lambda} = \sup_{\lambda' \subset \lambda} |c_{\lambda'}|$$

3) Remove the null wavelet leaders and compute

$$S(q,j) = rac{1}{2^j} \sum_{\lambda \in \Lambda_j} d_\lambda^q,$$

where  $\Lambda_i$  is the set of dyadic intervals at scale *j*.

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4) Compute the function  $\tau$  defined as

$$\tau(q) = \underline{\lim}_{j \to +\infty} \frac{\log(\mathcal{S}(q, j))}{\log 2^{-j}},$$

which is numerically obtained through the slopes of linear regressions at small scales of log(S(q, j)) seen as a function of *j*.

5) The spectrum of singularities is obtained as

$$d(h) = \inf_{q} \{qh - \tau(q)\} + 1.$$



log(S(q,j)) for a fractional Brownian motion with Hölder exponent 0.5 with q ranging from -1 to 1.



 $\tau$  function associated to the previous signal. Linear regression gives a slope of 0.494021.

 Remark : *f* is monofractal if and only if τ is a straight line, in which case the Hölder exponent of *f* is the slope of τ.

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- Remark : *f* is monofractal if and only if τ is a straight line, in which case the Hölder exponent of *f* is the slope of τ.
- If *f* is a monofractal signal with Hölder exponent *H*, then *f* belongs to the uniform Hölder space C<sup>H</sup>, and a norm in this space is defined by

$$\|f\|_{C^{H}} = \sup_{j,k} \{|c_{j,k}|/2^{jH}\} := N$$

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#### Analyzed data



- Daily mean temperature data from 1951 to 2003, calculated as average of minimum and maximum daily temperatures
- Weather stations located below 1000 meters of altitude
- 115 stations selected
- Missing data up to 7%, less than 1% for 97 stations
- Values integrated for more stable numerical results (i.e.  $x_n$  replaced by  $\sum_{j=1}^{n} x_j$ .)

#### Monofractal nature of the signals



τ functions associated to Aachen (green), Шепетивка (blue) and Milano (orange), with respective slopes 1.156, 1.218, 1.358.

#### Hölder exponents and norms

- $\tau$  linear  $\implies$  signals are monofractal
- Mean coefficient of determination :  $R^2 = 0.9975 \pm 0.0028$
- Hölder exponents ranging from 1.093 to 1.43
- Norms ranging from 8.23 to 30.45



Link with climate types ?

#### Köppen-Geiger climate classification

Classification based on maximum and minimum monthly mean temperatures (references fixed at 22°C and 0°C). Stations close to 0.5°C of another type of climate were also associated to this second category. Here, precipitations were not taken into account.



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### **Climate distribution**





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### Hölder spaces-based climate classification and results

Maximum matching with Köppen-Geiger classification if

$$H_1 = 1.186$$
  
 $H_2 = 1.275$   
 $N_1 = 14.81$   
 $N_2 = 16.18$ 

Result : 93.9% correctly associated

Remark : without the norm, 89.6% correctly associated.

#### Results on the map



Stations marked with a brown cross are the ones whose type of climate was erroneously predicted. The others were correctly predicted ; green discs stand for Oceanic climate, blue triangles correspond to continental stations and orange orange diamonds are the Mediterranean ones.

#### Blind test



- 69 other stations
- 40 years of data between 1951 and 2003
- Mean temperatures sometimes computed in a different way

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#### Blind test



Result : 88.4% correctly associated

Remark : without the norm, 84.1% correctly associated.

### Discussion of the results

# <u>Results</u>

Oceanic stations	$\longleftrightarrow$	Lowest Hölder exponents
Continental stations	$\longleftrightarrow$	Intermediate Hölder exponents
Mediterranean stations	$\longleftrightarrow$	Largest Hölder exponents

### Discussion

- On a daily basis, Oceanic climate is more irregular than the Continental weather, which is less regular than Mediterranean climate.
- Explanation could be the North Atlantic Oscillation (NAO), anticyclonic conditions in Southern Europe, ...

### Conclusions and future work

# Conclusions

- WLM shows surface air temperatures signals are monofractal signals
- Their belonging to functional spaces reflects their temperature-based Köppen-Geiger climate type
- Algorithm and results confirmed through blind tests

### Future work

- Checking of the validity of current climatic models
- Analysis of other climate indices (pressure, precipitation,...)
- Generalization to global temperatures

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