

# A multifractal analysis of air temperature signals based on the wavelet leaders method

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### 1 Hölder regularity

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

### 2 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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## Hölder exponent

## 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

$$|f(x) - P_{x_0,\alpha}(x)| \leq C|x - x_0|^\alpha$$

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

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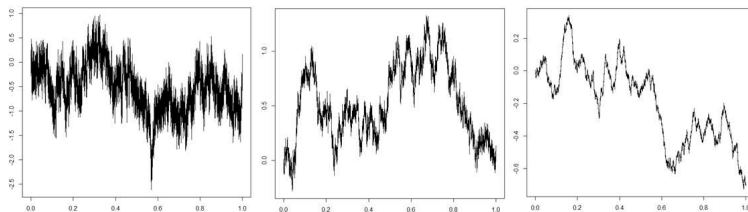
## Definition

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  $C^\alpha(x_0)$  :

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

# Monofractality

- 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.

# Spectrum of singularities

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  $\dim_{\mathcal{H}}(X)$  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.

## Wavelet leaders method (WLM)

### 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

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



## Wavelet leaders method (WLM)

- 3) Remove the null wavelet leaders and compute

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

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

- 4) Compute the function  $\tau$  defined as

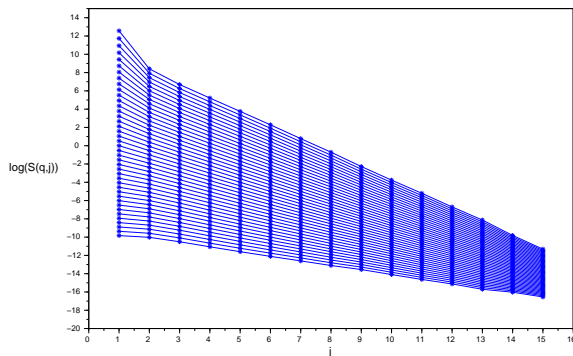
$$\tau(q) = \lim_{j \rightarrow +\infty} \frac{\log(S(q, j))}{\log 2^{-j}},$$

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

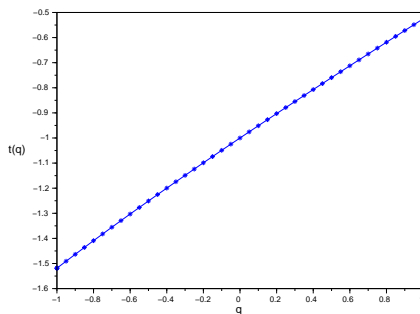
$$d(h) = \inf_q \{qh - \tau(q)\} + 1.$$

## Wavelet leaders method (WLM)



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

## Wavelet leaders method (WLM)



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

- 6) Remark : if  $\tau$  is a straight line, then  $f$  is monofractal, in which case the Hölder exponent of  $f$  is the slope of  $\tau$ .

# Wavelet leaders method (WLM)

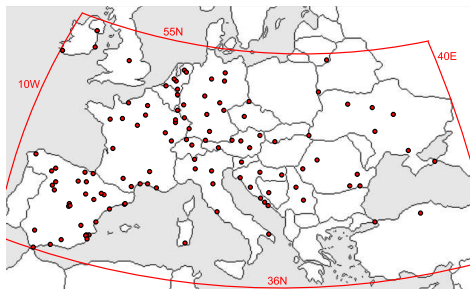
- Remark : if  $\tau$  is a straight line, then  $f$  is monofractal, in which case the Hölder exponent of  $f$  is the slope of  $\tau$ .
- If  $f$  is a monofractal signal with Hölder exponent  $H$ , then  $f$  belongs to the uniform Hölder space  $C^H$ , and a norm in this space is defined by

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

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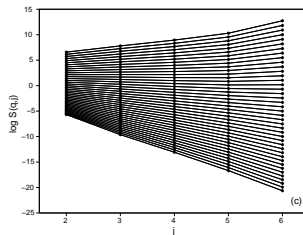
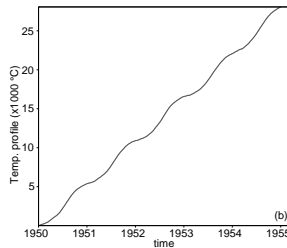
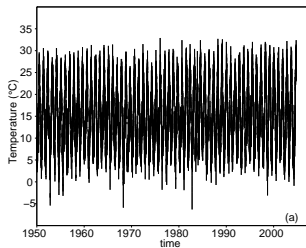
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## Analyzed data

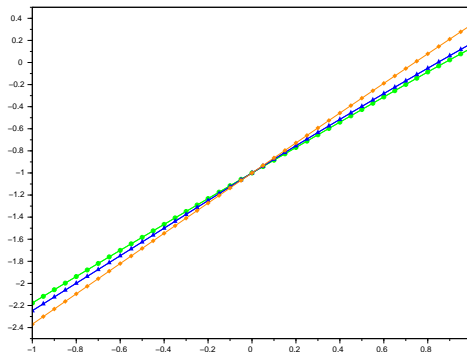


- Daily mean temperature data from 1951 to 2003, calculated as average of minimum and maximum daily temperatures
- Weather stations located below 1000 meters of altitude
- 115 stations selected
- Missing data up to 7%, less than 1% for 97 stations
- Values integrated for more stable numerical results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

# Analyzed data (Granada)



# Monofractal nature of the signals



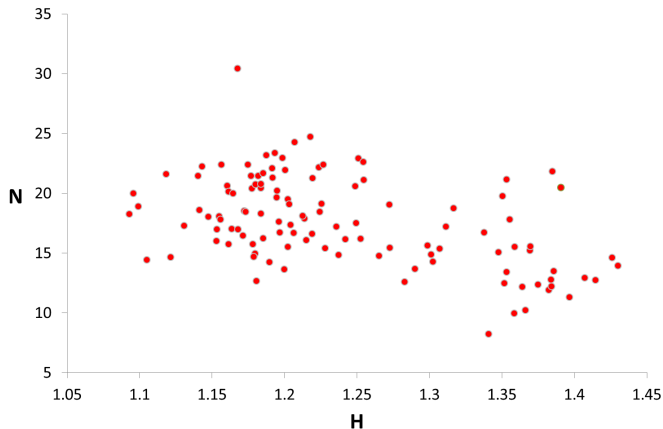
$\tau$  functions associated to Aachen (green), Шепетивка (blue) and Granada (orange), with respective slopes 1.156, 1.218, 1.323.



# 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

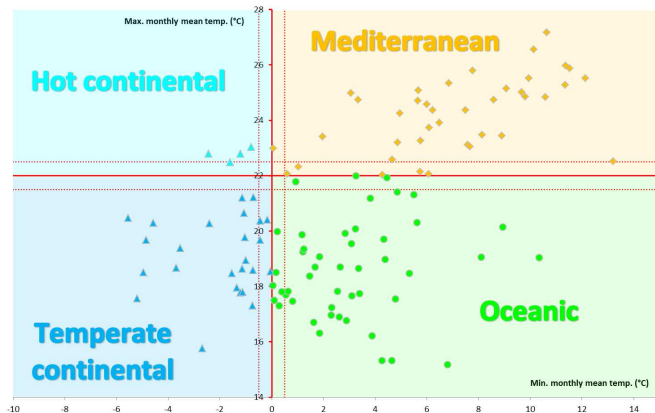
# Distribution of the exponents and norms



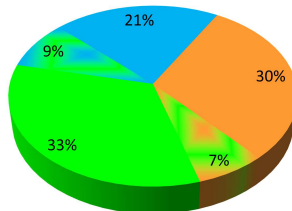
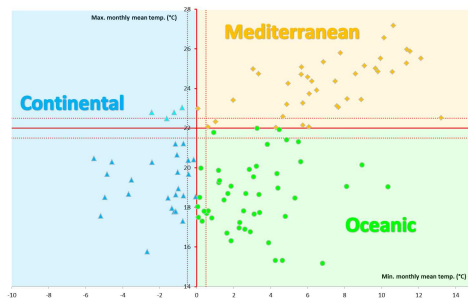
Link with climate types ?

## Köppen-Geiger climate classification

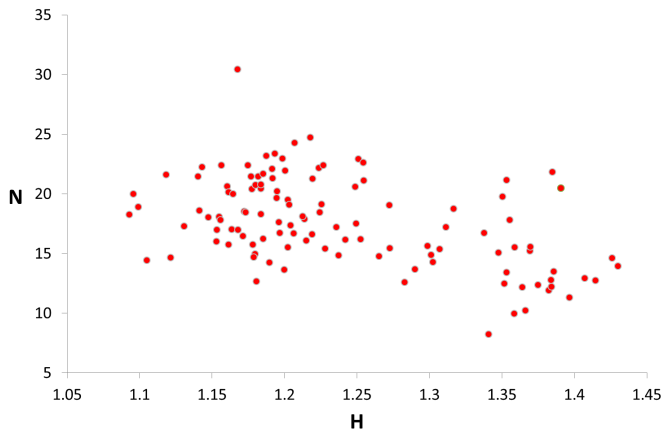
Classification based on maximum and minimum monthly mean temperatures (references fixed at  $22^{\circ}\text{C}$  and  $0^{\circ}\text{C}$ ). Stations close to  $0.5^{\circ}\text{C}$  of another type of climate were also associated to this second category. Here, precipitations were not taken into account.



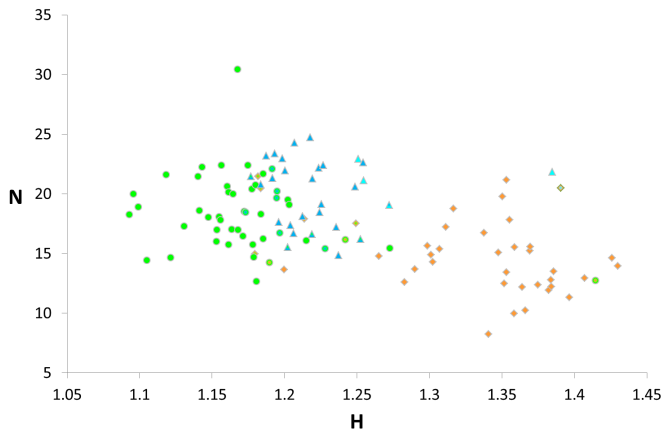
# Climate distribution



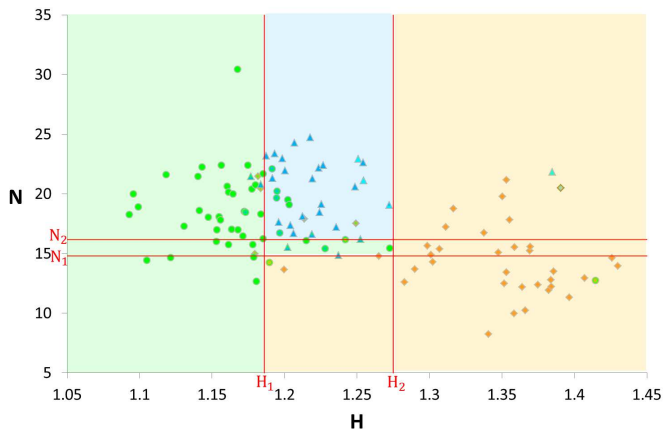
# Distribution of the exponents and norms



# Distribution of the exponents and norms



# Distribution of the exponents and norms



## Hölder spaces-based climate classification and results

Maximum matching with Köppen-Geiger classification if

$$H_1 = 1.186$$

$$H_2 = 1.275$$

$$N_1 = 14.81$$

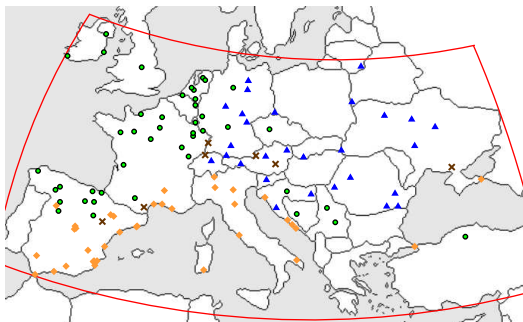
$$N_2 = 16.18$$

**Result : 93.9% correctly associated**

Remark : without the norm, 89.6% correctly associated.

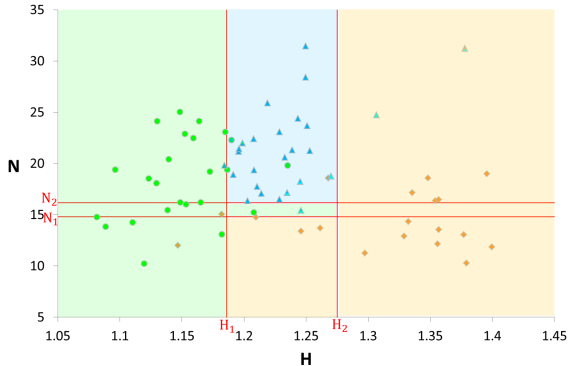


## Results on the map



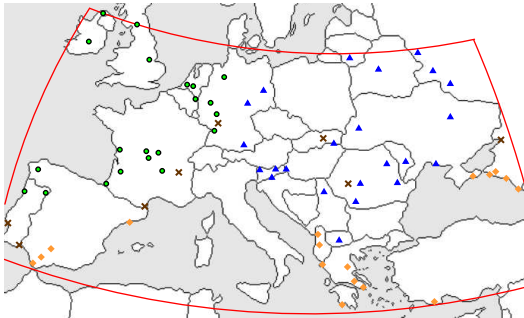
Stations marked with a brown cross are the ones whose type of climate was erroneously predicted. The others were correctly predicted ; green discs stand for Oceanic climate, blue triangles correspond to continental stations and orange orange diamonds are the Mediterranean ones.

# Blind test



- 69 other stations
- 40 years of data between 1951 and 2003

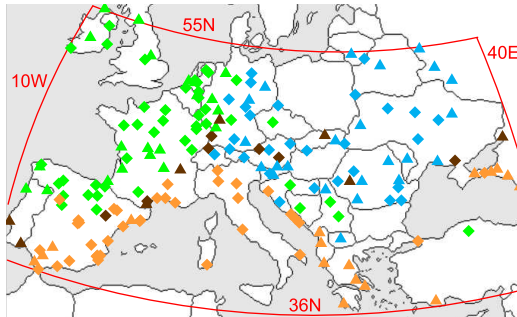
## Blind test



Result : 88.4% correctly associated

Remark : without the norm, 84.1% correctly associated.

## All the stations



115 stations of reference (diamonds) + 69 stations of the blind test (triangles)

Overall result: 91.8% correctly associated

## Discussion of the results

### Results

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

## Some references



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