# A study of histograms of wavelet coefficients of the Mars topography to determine its scaling properties

KLEYNTSSENS Thomas Institute of Mathematics, University of Liège (Belgium) tkleyntssens@ulg.ac.be

joint work with Samuel NICOLAY and Laura REYNAERTS



Introduction and Short Review on the Scaling Properties of Mars Topography

Previous works about the scaling properties of Mars topography show two distinct scaling regimes. The scale break and the scaling exponents H vary from one paper to another:

Method	Small scales	Large scales
Power spectral density [1]	$H \approx 1.2 \; (< 10 \; \mathrm{km})$	$H \approx 0.2 - 0.5$

Variance of a wavelet transform [4]	H	$\approx 1.25$	(< 24  km)	H	$\thickapprox 0.5$
Statistical moments [5]	H	pprox 0.76	(< 10  km)	H	$\thickapprox 0.5$
Wavelet leaders method $[2]$	H	$\approx 1.1$ (-	< 15  km)	H	$\approx 0.7$

Let us notice that the three first methods are based on along-track measurements, which implies that the 2D part of the topography field has not been taken into account. The wavelet leaders method has allowed to perform the first complete 2D study of the surface roughness of Mars. The presence of specifics regions such as craters can disrupt this last method and thus the interpretation of the scalings of some signals can be more difficult. In this work, a recent method based on the use of histograms of wavelet coefficients is used to avoid these scaling problems [3]. This poster presents the first results for the small scales. We use the MOLA data, using the 128 pix/deg map (pds-geosciences.wustl.egu).

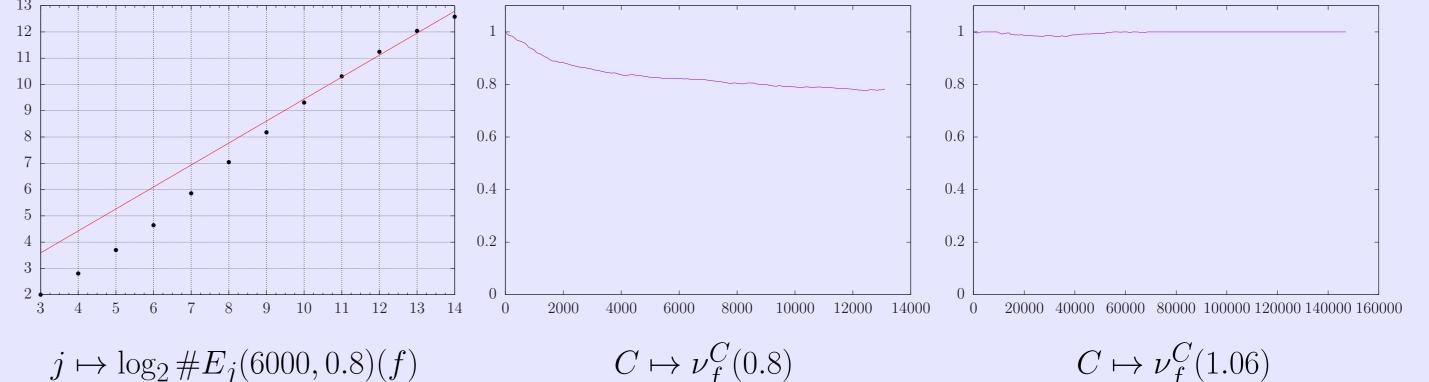
### The Wavelet Profile Method (WPM)

Let us denote by  $c_{\lambda}$  the wavelet coefficient of a function  $f : \mathbb{R}^n \to \mathbb{R}$ , associated to the dyadic cube  $\lambda = [k2^{-j}, (k+1)2^{-j})^n \ (j \in \mathbb{N}$  and  $k \in \{0, \ldots, 2^j - 1\}^n$ . One has that  $c_{\lambda_j(x)} \sim 2^{-H_f(x)j}$  as  $j \to +\infty$ , where  $\lambda_j(x)$  is the unique dyadic cube of size  $2^{-j}$  that contains x and  $H_f(x)$  is the scaling exponent of f at the point x. Let us set

$$\nu_f(h) = \lim_{\epsilon \to 0^+} \limsup \frac{\log \# E_j(1, h + \epsilon)(f)}{\log 2j},\tag{1}$$

### 1D Analysis

These figures represent the classical steps of the WPM. The scale j corresponds to  $0.463 \times 2^{15-j}$  km (1 pixel corresponds to 0.463 km).



 $j \rightarrow +\infty$ 

where

#### $E_j(C,h)(f) = \{\lambda \in \Lambda_j : |c_\lambda| \ge C2^{-hj}\}$

and  $\Lambda_j$  is the set of dyadic cube of size  $2^{-j}$ . If a signal f has the same scaling exponent H at each point, one has that  $\nu_f(h) = -\infty$  if h < H and  $\nu_f(h) = n$  if  $h \ge H$ . More generally, the smallest h such that  $\nu_f(h) = n$ represents the *dominant scaling exponent* of f, noted  $H_f$ . Moreover, theoretically, the constant 1 in Equality (1) is arbitrary but in practice, it is not the case because we have only a finite number of wavelet coefficients. Here is a algorithm to approximate  $H_f$  [3]:

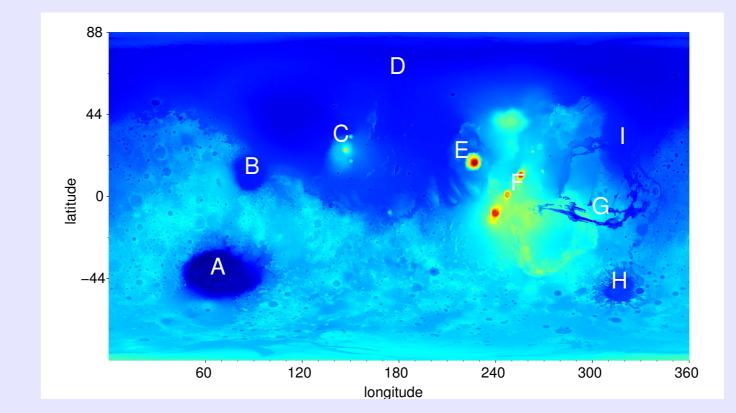
• For a fixed h > 0, we first do a linear regression, noted  $\nu_f^C(h)$ , on the function

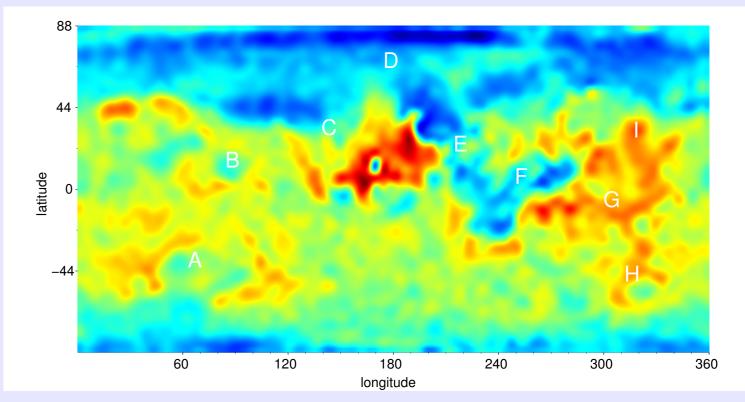
 $j \mapsto \log_2 \# E_j(C, h)(f). \tag{2}$ 

• Secondly, we construct the function  $C \mapsto \nu_f^C(h)$ . If h is larger than  $H_f$ , then it should exist an interval I for which the values  $\nu_f^C(h)$  with  $C \in I$ are close to n. The typical length of I is the median of  $(|c_{\lambda}|/2^{-hj})_{j,k}$  (see [3]). The smaller h verifying this property is an approximation of  $H_f$ . In this poster, we do a linear regression of the function (2) on the small scales j (< 15 km). Results obtained for the longitudes (l) and latitudes (L) :

## 2D Analysis

The map is gridded into tiles of  $1024 \times 1024$  pixels, corresponding to windows of  $8^{\circ} \times 8^{\circ}$  on Mars. The spatial distribution of the scaling exponents at small scales shows that the most distinctive features of Mars can be recovered.





#### Conclusion

This works shows that the Wavelet Profile Method is well-suited for studying the scaling properties of planetary surfaces. We still have to refine the method, study the multifractality and the large scales, and compare the results with results of some classical methods.

#### Acknowledgement

- This research used resources of the "Plateforme Technologique de Calcul Intensif (PTCI)" (http://www.ptci.unamur.be) located at the University of Namur, Belgium, which is supported by the F.R.S.-FNRS under the convention No. 2.5020.11. The PTCI is member of the "Consortium des Équipements de Calcul Intensif (CÉCI)" (http://www.ceci-hpc.be).
- The authors would like to thank Adrien Deliège from the University of Liege for the use of some Scilab's code to generate the maps of this poster.

topographic map spatial distribution of *H* (WPM with wavelet leaders) Regions of interest : Hellas Planitia (A), Isidis Planitia (B), Elysium Mons (C), Vestitas Borealis-Northern plains (D), Olympus Mons (E), Tharsis (F), Valles Marineris (G), Argyre Planitia (H) and Acidalia Planitia (I).

#### **References :**

- [1] O. Aharonson, M. Zuber, and D. Rothman, Statistics of Mars' Topography from the Mars Orbiter Laser Altimeter: Slopes, Correlations, and Physical Models, Journal of Geophysical Research, 106(E10), 23723-23735, 2001.
- [2] A. Deliège, T. Kleyntssens, and S. Nicolay, Mars Topography Investigated Through the Wavelet Leaders Method: a Multidimensional Study of its Fractal Structure, Planetary and Space Science, 136C, 46-58, 2017.
- [3] T. Kleyntssens, C. Esser, and S. Nicolay, An algorithm for computing non-concave multifractal spectra using the  $S^{\nu}$  spaces, Communications in Nonlinear Science and Numerical Simulation, 56, 526-543, 2018.
- [4] B. Malamud, D. Turcotte, Wavelet analyses of Mars polar topography, Journal of Geophysical Research, 106(E8), 17497-17504, 2001.

 [5] F. Landais, F. Schmidt, S. Lovejoy, Universal multifractal Martian topography, Nonlinear Processes in Geophysics Discussions 2, 1007-1031, 2015.