

Original Research Paper

Study of Sunspot Time Series Using Wavelet-Based Multiracial Analysis during Solar Cycle 23 and Ascending Phase of Cycle 24

ABSTRACT

Wavelet based Multifractal analysis techniques provides a sophisticated statistical characterization of many complex dynamical phenomena related with Sun and its environment. In this work multifractal property of the Sunspot number time series, during the period of Solar cycle 23 and ascending phase of Solar cycle 24 are examined using the Wavelet transform and wavelet based multifractal approach. Present analysis has been performed using the software FRACLAB, developed at INRIA and available online at <http://www-rocq.inria.fr>. The results from the wavelet based multifractal analysis show that multifractality exists in Solar Cycle 23 and ascending phase of cycle 24. It is also found singularities spectrum for sunspot time series was well Gaussian in shape suggesting the multifractal characteristics of time series. It was concluded that this multifractal based approach provide the local and adaptive description of dynamics processes occurring in the Sun and can be applied effectively to the study of solar activity.

Keywords: Sunspot number, Magnetic field, Multifractal analysis and Wavelet transform technique.

1. INTRODUCTION

The significant feature of the Sun's outer regions is the existence of a reasonably strong magnetic field which governs all the solar activities inside the Sun and its atmosphere. According to the lowest order of approximation, the sun's magnetic field is dipolar and axisymmetric in nature and at a typical point on the solar surface its value is few gauss. However significant variation in this value and there are localized regions (sunspots) in which the field can much higher by Bray [1]. In fact, sometimes the magnetic field value is so strong which result solar flares. The symmetry of the twisted magnetic lines as the origin of sunspots, it is generally seen in pairs or in groups of pairs at both sides of the solar equator. As the sunspot cycle progresses, spots appear closer to sun's equator giving rise to the so called "butterfly diagram" in the time latitude distribution by Petrovaye [2]. The twisted magnetic fields above sunspots are sites where solar flares are observed. It has been found that chromospheric flares show a very close statistical relationship with sunspots by Bray [1]. The number of sunspots continuously changing in time in a random fashion and constitutes a typically random time series Fanchiotti et al. [3].

Tarbell et al. [4] were the first, used the fractal analysis in the context of solar magnetic field to find a fractal dimension. They have found high resolution (0.3 – 0.5 arcsec) magnetograms of an active region (at the Swedish Solar Observatory on La Plama Conary Islands). Multifractal spectrum studied by the many authors for example Feder [5], Takayasu [6], Frisch [7], Schroeder [8] and Ireland et al. [9]. Gallagher et al. [10] applied the fractal analysis to structure of the solar corona found in extreme ultraviolet (EUV) spectral range. Tarbell et al. [4] and Balke et al. [11] used the fractal technique to study the photospheric magnetic structure. Tao et al. [12] attempted the multifractal analysis of the surface magnetic field as numerically simulate distribution. Later, numerical simulations of multifractality and intermittency of the solar structure were performed by so many researchers Cadavid et al., Georgioulis et al., Dmitruk et al., Janseen et al., and Nigro et al., [13-17]. Schroeder [8] used the multifractal and intermittency to study the social and biological systems and throughout the broad range of scientific studies and interstellar medium by Falgarone et al.; Cho et al.; Krotsuk and Norman; Balsara and Kim [18-21].

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Sunspots and related activities have been analyzed by various methods, including correlation analysis by Temmer et al. [22], Chaos analysis by Veronig and Jevtic [23, 24] and multifractal analysis by Abramenko, Movahed and McAteer [25-27]. Multifractal theory provides an elegant statistical characterization of many complex dynamical variations in nature and engineering by Gao et al. [28, 29]. It is conceivable that it may enrich characterization of the sun's magnetic activity and its dynamical modeling by Howe et al. [30]. In recent times, a particularly noteworthy work has been reported by Movahed [32]. The relative fraction of small scale fluctuation in the magnetic field contributes significantly more prior to flaring by Abramenko et al. [31]. It was found that active regions reach a critical state of intermittency prior to flaring by Abramenko [25]. The multifractal scaling behaviors reported by Movahed [26] are valid for timescales up to more than 50 years. McAteer [32] found that analytical connection between multifractal formalism and the set of 3D equations that govern the small-scale and large-scale magnetic structure on the Sun and also based on fractal geometry and turbulence in magnetized plasmas by McAteer et al. [33,34]. Recently, Georgoulis [35] and McAteer [32] achieved a contrary conclusion that studies in multifractals do not provide a predictive ability for the onset of solar flares. In this paper we have used the multifractal techniques and noticed the presence of multifractality in sunspot number time series during the Solar Cycle 23 and ascending phase of Solar Cycle 24.

2. DATA SET

In this analysis we used the monthly counts of sunspot number for the multifractal analysis for the time span of 1996 to 2013. This period include complete Solar Cycles of 23 and ascending phase of current Solar Cycle 24. The data set available online and downloaded from <http://www.ngdc.noaa.gov/stp/space-weather/solar-data/solar-indices/>.

3. THEORETICAL BACKGROUND

3.1. WAVELET TRANSFORMS

Wavelet transform is ideal for the analysis of real world signals that contain sharp changes and localized discontinuities by Quiroz [36]. The wavelet transform use different window sizes, which are able to compress and stretch wavelets in different scales or widths; these are then used to decompose a time series by Santos et al. [37] and decompose a one-dimensional signal into two-dimensional time-frequency domains at the same time by Adamowski et al. [38]. Wavelet transform can be performed using two approaches: Continuous Wavelet Transform (CWT) and Discrete Wavelet Transform (DWT). The CWT introduces a very redundant and finely detailed description of time series in terms of both time and frequency. It is particularly helpful in resolving problems involving signal detection and identification of hidden transients such as hard to detect, short – lived elements of a time series. The scales and locations used in DWT are normally based on a dyadic arrangement (i.e. integer powers of two) by Chou [39]. DWT is especially useful for time series containing sharp jumps or shifts by Partal et al. [40].

3.2 TIME SERIES DECOMPOSITION VIA THE DISCRETE WAVELET TRANSFORM (DWT)

The DWT is usually based on the dyadic calculation of position and scale of a signal by Chou [39]. The DWT is excellent for denoising the signals by Fugal [41]. The DWT of a vector is the outcome of a linear transformation resulting in a new vector that has equal dimensions to those of the initial vector by Chou [42]. This transformation is the decomposition process. In this study all the time series (for the Solar Cycles 23 and ascending phase of current Cycle 24) were decomposed using the Daubechies (db2) and Coifman (coif5) wavelets, splitting the signal into its approximation and detail components and displayed in the figures (Fig. 1 to 4). The discretization of wavelet functions is accomplished using a logarithmic uniform spacing that has a coarser resolution at higher scales by Mallat and Daubechies [43, 44]. In this DWT- based study of sunspot data through the Daubechies (db) and Coifman (coif) wavelets were used. Daubechies wavelets provide compact support by Vonesch [45], indicating that the wavelets have non-

not clear.

zero basis functions over a finite interval, as well as full scaling and translational orthonormality properties Popivanov [46]. These features are very important for localizing events in the time-dependent signals Popivanov [46].

3.3 WAVELET BASED MULTIFRACTAL FORMALISM

The wavelet transform not only locates isolated anomalous events, but can also characterize more complex multifractal sunspot data having non isolated singularities. Multifractal objects cannot be completely described using a single fractal dimension (mono fractals). They have an infinite number of dimension measures associated with them. The wavelet transform takes advantage of multifractal self-similarities, in order to compute the distribution of their singularities. This singularity spectrum is used to analyze multifractal properties. The concepts of fractals and multifractals and their relevance to the real world data were introduced by the Mandelbrot [47]. The time series of sunspot numbers usually depict fractal or multifractal features. Time series are commonly called self-affine functions as their graphs are self-affine sets that are similar to themselves when transformed by anisotropic dilations.

Mathematically, if $f(x)$ is a self-affine function representing the sunspot number then

For $x_0 \in R, \exists H \in R$ such that for any $\lambda > 0$,

$$f(x_0 + \lambda x) - f(x_0) \cong \lambda^H (f(x_0 + x) - f(x_0)). \quad (1) \text{ rewrite}$$

The exponent H here is called roughness or Hurst exponent. Note that if $H < 1$, then f is not differentiable and smaller the exponent H , the more singular is f .

For sunspot number $f(x)$, a function $h(x)$, the Holder function of f , which measures the regularity of f at which point t is associated.

The point wise Holder h of f at point x_0 is defined as:

$$h(x_0) = \limsup \{h: \exists c > 0, |f(x) - f(x_0)| \leq c|x - x_0|^h, |x - x_0| < \rho\} \quad (2)$$

(Here h is an integer and f is non-differential).

One may also define local exponent $h_1(x_0)$ as:

$$h_1(x_0) = \limsup \{h: \exists c > 0, |f(x) - f(y)| \leq c|x - y|^h, |x - x_0| < \rho, |y - x_0| < \rho\} \quad (3)$$

Where h and h_1 are different in general. For example for

$$f(x) = |x|^h \sin \frac{1}{|x|^\beta}, h(0) = 0, \text{ while } h_1(0) = \frac{h}{1+\beta}. \quad (4)$$

They have quite different properties. For instance h_1 is stable through differentiation ($h_1(f', x_0) = h_1(f, x_0) - 1$), whereas h is not. The smaller $h(x)$ is, the more irregular the function f is at t .

3.4 PARTITION FUNCTION

Calculation of pointwise Lipschitz (Holder) regularity of multifractal is not possible because its singularities are not isolated and the finite numerical resolution sufficient to discriminate them. To overcome this difficulty by Muzy [48] have introduced the concept of wavelet transform modulus maxima using a global partition function by Arneodo [49]. Let Ψ be a wavelet with n vanishing moments. Mallat [43] has shown that if f has a pointwise Holder (Lipschitz) regularity $\alpha_0 \leq n$ at v then the wavelet transform $T_\Psi f(a, b)$ has a sequence of modulus maxima that converges towards v at fine scales. The set of maxima at the scale a can thus be interpreted as a covering of the singular support of f with wavelets of scale a . At these maxima locations

$$|T_\Psi f(a, b)| \approx a^{\alpha_0 + 1/2} \quad (5)$$

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148 Let $\{u_p(a)\}_{p \in \mathbb{Z}}$ be the position of all local maxima of $|T_\psi f(a, b)| \approx a^{a_0+1/2}$ at a fixed scale a . The partition
 149 function Z measures the sum at a power q of all these wavelet modulus maxima:

$$Z(q, a) = \sum_p |T_\psi f(a, u_p)|^q \quad (6)$$

150 For each $q \in \mathbb{R}$, the scaling exponent $\tau(q)$ measures the asymptotic decay of $Z(q, a)$ at fine scale a :

$$\tau(q) = \liminf \frac{\log Z(q, a)}{\log a} \quad (7)$$

152 This typically means that $Z(q, a) \sim a^{\tau(q)}$.

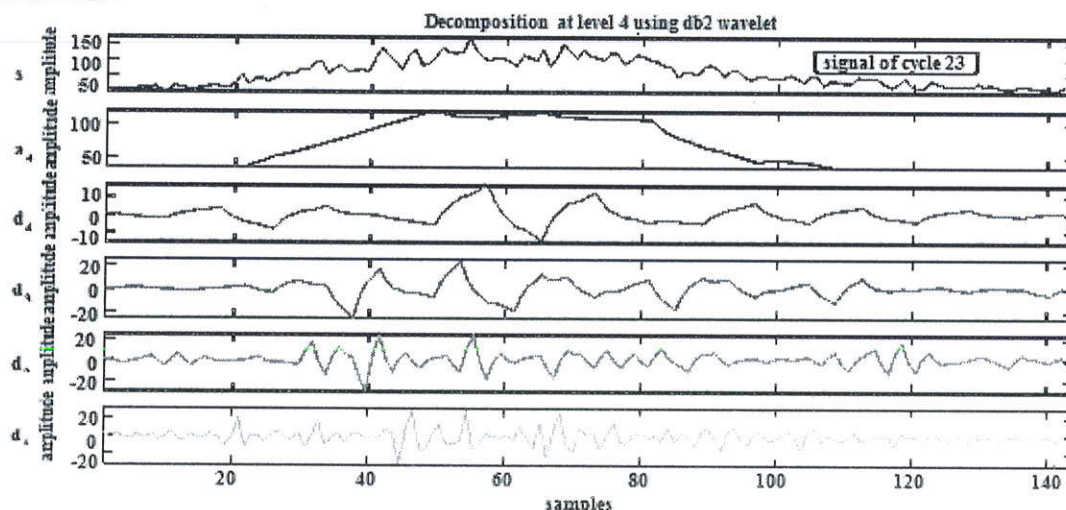
155 3.5 RESULTS AND ANALYSIS

156 In this study we have analyzed the multifractal characteristics of sunspot number during the Solar Cycle
 157 23 and current Cycle 24 using wavelet based multifractal techniques. Sunspot numbers (SNs), are widely
 158 and frequently used in astronomy and space science to reflect long term variations of solar activity, which
 159 has served as the primary time series to define solar activity by Hoyt and Schatten [50-53].

160 3.5.1 WAVELET ANALYSIS OF SUNSPOT NUMBERS

161 In this section a given data set is divided into components with different scales, which allows the
 162 investigations of each component with a resolution matched to its scale. This property is especially useful
 163 for time series that are non-stationary, short-lived transient components and have features at different
 164 scales or have singularities Kumar and Foufoula-Georgiou [54]. In the present analysis both Daubechies
 165 and Coifmann mother wavelets are used to decompose sunspot number and Morlet mother wavelet was
 166 used for Multifractal spectral analysis.

167 To study DWT, we have taken the 140 and 72 samples (Sunspot time series) of the Solar Cycle 23 and
 168 Solar Cycle 24 respectively. At the first decomposition level "d1" the sunspot time series samples passed
 169 through the high pass and low pass filters, followed by subsampling by 2. The output of the high pass
 170 filter has 60 and 35 samples i.e. half the time resolution and double the frequency resolution. These 60
 171 and 35 samples constitute the first level "d1" of DWT coefficients. The output of the low pass filter also has
 172 60 and 35 samples, but it spans half the frequency band. This samples is then passed through the same
 173 high pass and low pass filters for further decomposition. This process continues until two samples are left.
 174 The DWT composition (e.g. Fig. 1-4) shows that discrete form of Solar Cycle 23 and 24 sunspot time
 175 series due to this it was noticed that at lower and lower frequencies during Cycle 23 sunspot number
 176 were more frequent than current Cycle 24 therefore more multifractality. The horizontal axis is represent
 177 the number of samples of sunspot time series, whereas the vertical axis is the normalized to unit
 178 amplitude.



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Fig. 1. DWT decompositions of Solar Cycle 23 using Daubechies2 wavelet. Anomalous variation should be found between 45-60 no. of samples (i.e. at the maximum phase of cycle).

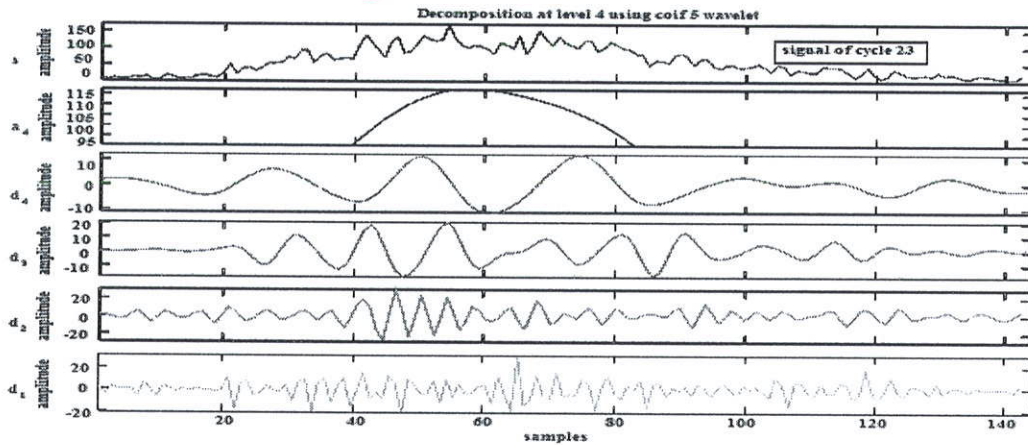


Fig. 2. DWT decompositions of Solar Cycle 23 using Coifman5 wavelet. Anomalous sharp variation should be found 40-60 no. of samples (i.e. at the maximum phase of cycle).

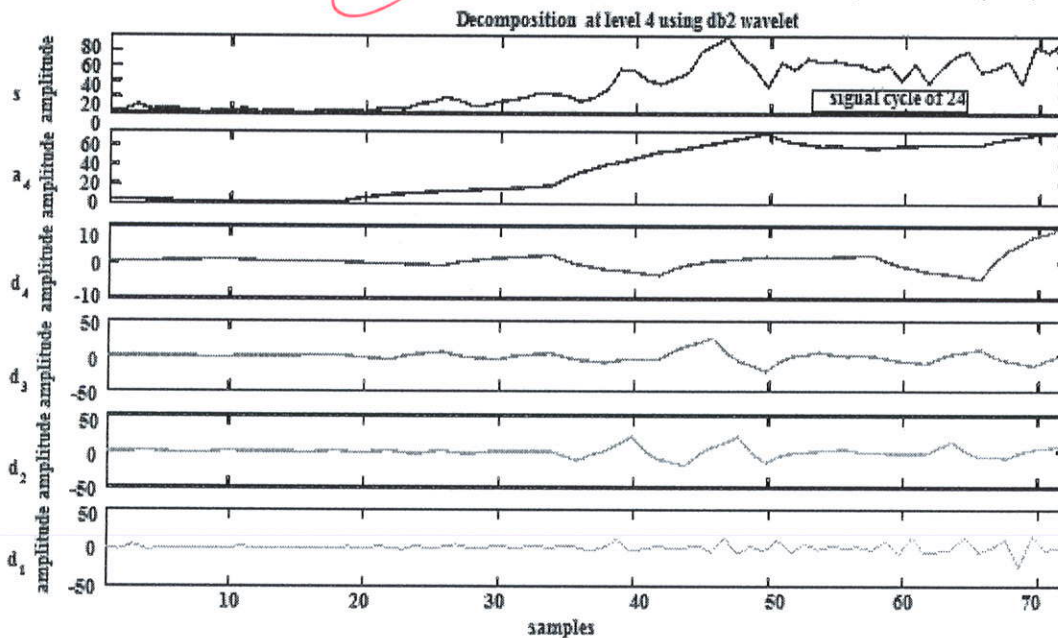


Fig. 3. DWT decompositions of Solar Cycle 24 using Daubechies2 wavelet. Anomalous variation showing from 40 to 50 no. of samples (i.e. at the maximum phase of cycle).

Q5 - Can we find the precise location of the maximum phase?

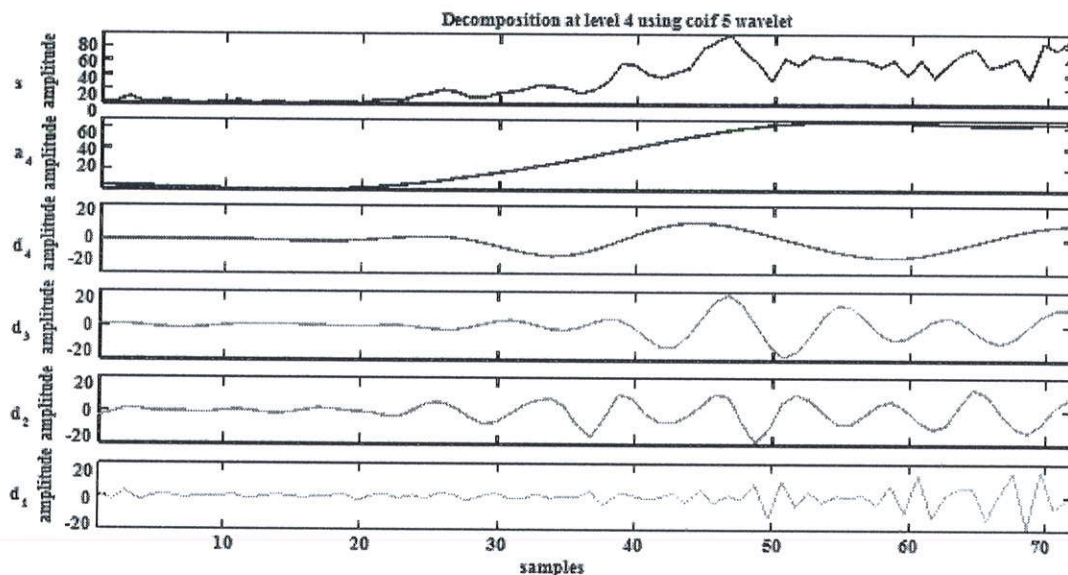


Fig. 4. DWT decompositions of Solar Cycle 24 using Coifman5 wavelet. Anomalous sharp variation should be found between 40 to 50 no. of samples (i.e. at the maximum phase of cycle).

In this section, DWT of the sunspot numbers for Solar Cycle 23 and 24 were carried out in terms of approximations and details coefficients using Daubechies and Coifmann mother wavelets. These wavelets have fractal structure and they include both highly localized wavelets and highly smooth wavelets. The sunspot time series (s) is decomposed into two orthonormal components, frequency (details) and approximations component. The approximations represent the long term of data which is almost identified to original time series and the detail coefficients represent the short period fluctuations in given period range. The result of analysis was shown in Fig. 1 – 4. In figures, the X-axis represents the time period of the sunspot time series. In all figures first panel represents the variation in sunspot number time series. The second panel gives the approximation coefficient " a_4 " of all the figures. It separates the short term anomalous variation from the long term variations.

On the other hand, other four parts d_1 , d_2 , d_3 and d_4 represent detail coefficients of the sunspot time series. The detail coefficients reveal that the field strength changes between positive and negative values. This indicates the existence of a strong and variable magnetic field on the Sun as sunspots. The detail coefficients of sunspot numbers show that high frequency components during the initial and main phase of the solar cycle and their time evolution. Here it is found that very high frequency components present only during the main phase of the respective solar cycles and they are very strong in amplitude and stable for higher level of decomposition. Even though there are some strong fluctuations in the main phase and recovery phase but they are not as persistent as that present during the initial phase and main phase of the cycles.

3.6 MULTIFRACTAL ANALYSIS OF SUNSPOT NUMBERS

In this section, we have done multifractal analysis of sunspot numbers time series as fractal dimensions are best known part of the fractal analysis.

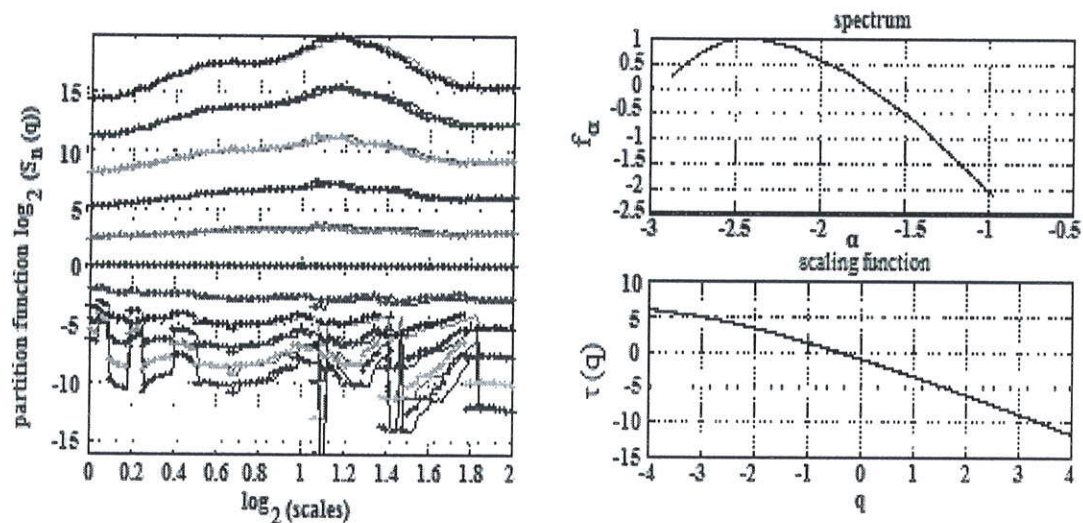


Fig. 5. Multifractal analysis of Solar Cycle 23 using CWT (Morlet wavelet of size 8 and 128 voices, LS regression and local maxima). In figure Left panel show the Legendre spectra and Right (upper) panel gives the singularity spectrum and Right (lower) panel for scaling function of the considered time series.

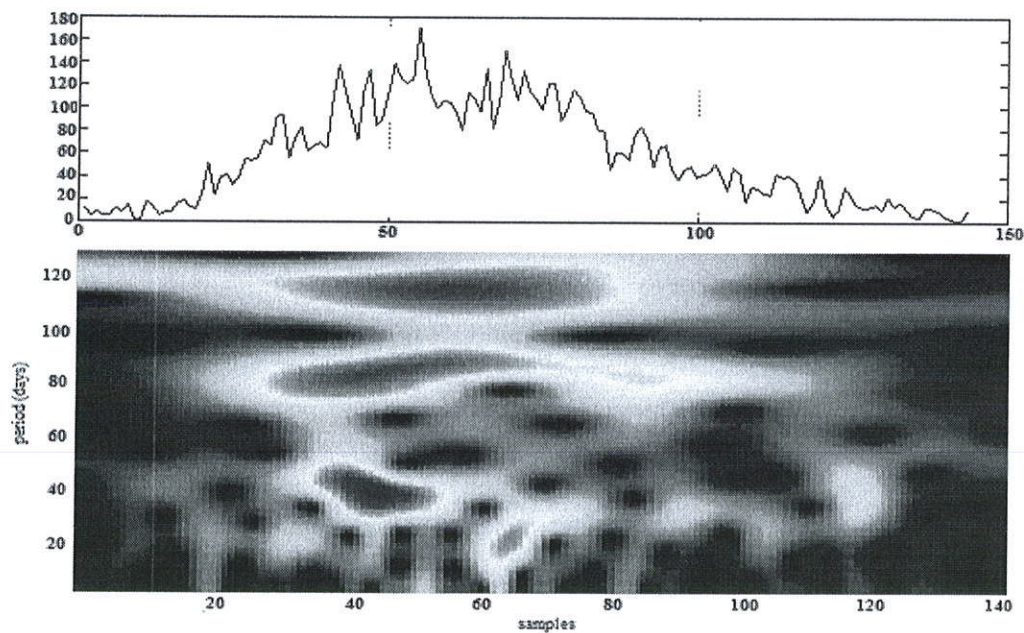


Fig. 6. CWT of Solar Cycle 23 using a Morlet Wavelet of size 8 and 128 voices. In figure X-axis represents the number of samples and Y-axis represents scale. The upper panel shows the raw data of the time series and lower panel its continuous wavelet transform (CWT).

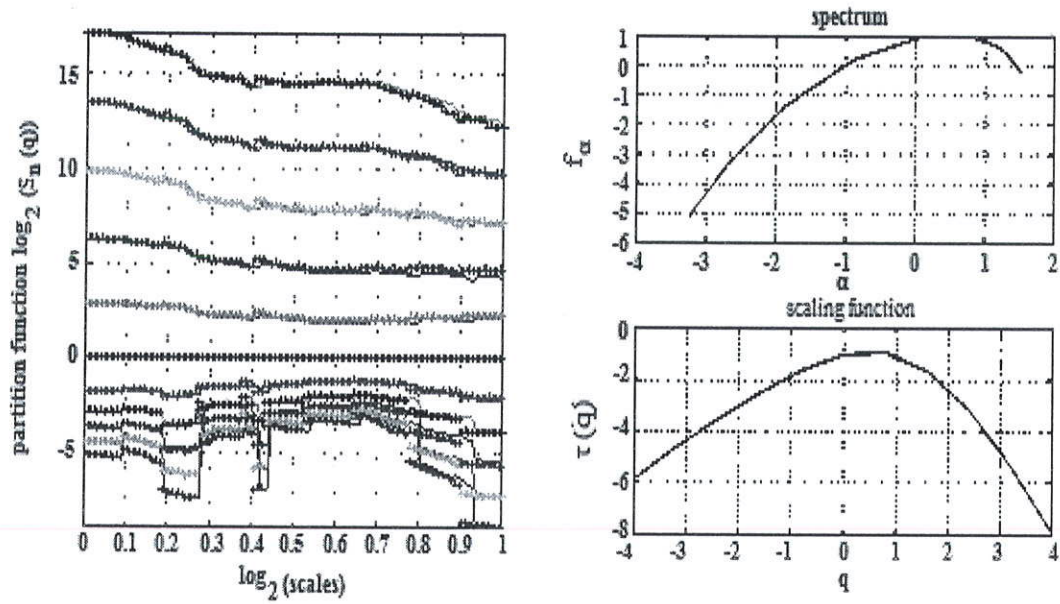


Fig. 7. Multifractal analysis of Solar Cycle 24 using CWT (Morlet wavelet of size 8 and 128 voices, LS regression and local maxima). In figure Left panel show the Legendre spectra and Right (upper) panel gives the singularity spectrum and Right (lower) panel for scaling function of the considered time series.

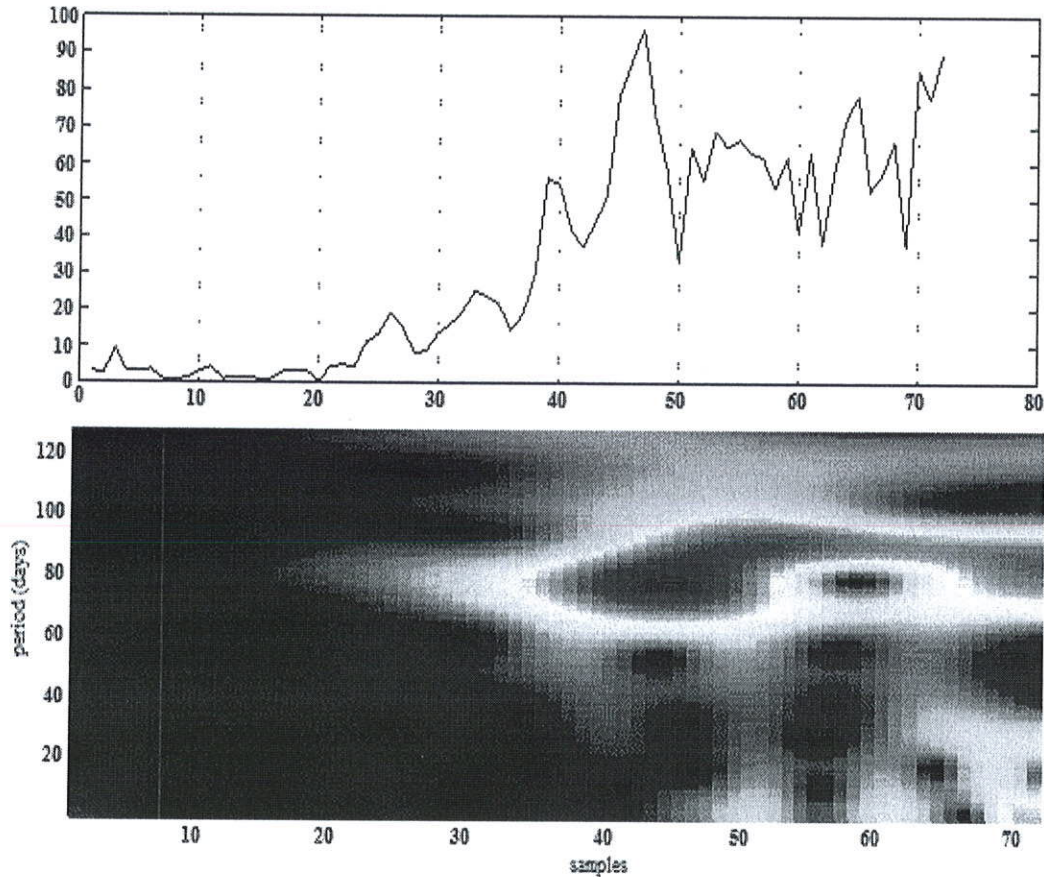


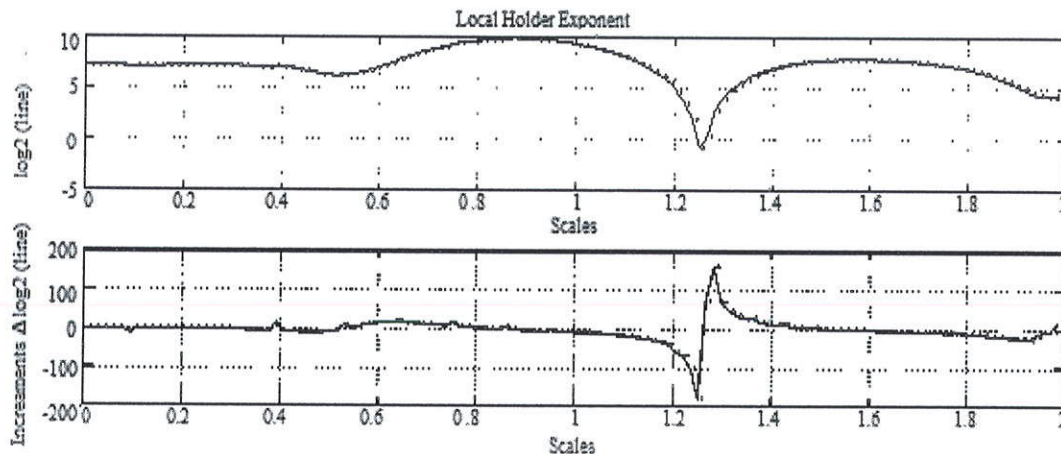
Fig. 8. CWT of Solar Cycle 24 using a Morlet Wavelet of size 8 and 128 voices. In figure X axis represents the number of samples and Y axis represents scale. The upper panel shows the raw data of the time series and lower panel its continuous wavelet transform (CWT).

We performed the multifractal analysis, calculating the Legendre spectra by means of the software FRACLAB, developed at INRIA and available online at <http://www-rocq.inria.fr>. Since the sunspot data have some missing values we take the longest segments of sunspot data without any missing values. The order of the magnitude of the length of each segment is about 120, thus permitting to obtain reliable estimates of the singularity spectrum and multifractal parameters. Fig. 5 and 7 (Right panel) shows the Legendre spectra for the selected segments for both solar cycle. All the spectra present the typical single-humped shape that characterizes multifractal nature of sunspot number. The spectra of the segments for each sunspot time series, calculated for different time intervals, are not identical, although they are very similar to each other. Nonlinear fluctuations are possibly due to the presence of multifractal processes. The smaller values of α correspond to the burst of events, while higher values of α correspond to events occurring sparsely by Voros [55]. The spectrum gives geometrical information pertaining to the dimension of sets of points in a signal having a given Holder exponent. This is the most precise spectrum from a mathematical point of view, but is also difficult one to estimate. Large deviation spectrum yields statistical information, related to the probability of finding a point with a given Holder exponent in the signal. More precisely, it measures how this probability behaves with the change in resolution.

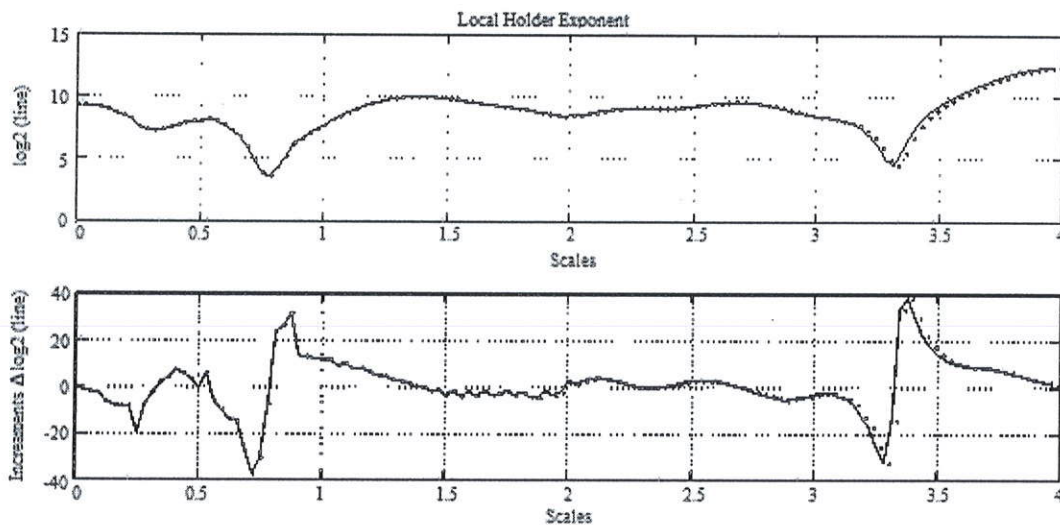
Legendre spectrum is a concave approximation to the large deviation spectrum. Its main purpose is to yield much more robust estimates, though at the expense of a loss of information. It could base on box method or CWT techniques. In the sequel we show some sample results for the spectra computed with the Legendre technique. Figures 6 and 8 show the results of the CWT (Morlet wavelet) based estimation of the Legendre spectrum which represents an approximation of the spectrum for two different Solar Cycles. Each of the Figures 6 and 8 consist of two parts. The first part on the top of each of these figures

— value of Holder Exponent?

269 represents the signal or raw data. The second part of these figures shows the analyzed pattern with the
 270 application of Morlet wavelet of size 8 and 128 voices of different Solar Cycles. From the figures 9 and 10
 271 it is cleared the exponents that arise sharply in the multifractal analysis of dataset to assess the regularity
 272 of a sunspot time series. The Holder exponent is used for this study that characterizes the regularity of
 273 the measure (function) of the magnetic field strength of sunspot under consideration at either pointwise
 274 regularity or local regularity. The non-parametric point wise Holder regularity approach (from Fig. 9 and
 275 10) used with CWT with Morlet wavelet for the Solar Cycle 23 and ascending phase of Cycle 24.



276
 277 Fig. 9. Non-parametric point wise Holder regularity estimation using CWT with Morlet wavelet for
 278 the Solar Cycle 23.



279
 280 Fig. 10. Non-parametric point wise Holder regularity estimation using CWT with Morlet wavelet for
 281 the Solar Cycle 24.

282 4. CONCLUSION

283
 284 Wavelet based multifractal analysis is a useful way to characterize the multifractality in time series. The
 285 result of our analysis shows that sunspot time series reveal multifractal character during Solar Cycle 23
 286 and ascending phase of Solar Cycle 24. The DWT (Fig. 1 - 4) of Sunspot time series shows the
 287
 288

(C9) Can summarize results in Table form by showing the values of Holder, etc.

phenomenological difference between Solar Cycle 23 and Cycle 24. It present Solar Cycle 23 has large range fluctuation due to the contribution of highly magnitude solar activities whereas during the Cycle 24 nit is not noticed hence it is a quite Cycle except during the maximum phase reveals the large fluctuations. The wavelet spectrum depicts in Fig. 6 and 8 reveals the time at which large variations in time series occur, which is more important for the investigation of multifractality. Various phenomenons related to Sun such as solar flares, sunspot, coronal mass ejection (CME) etc. Three main multifractal spectra viz. the Housedroff, large deviation and Legendre spectra (see Fig. 5 and 7) provides information as to which singularities occur in sunspot time series and which are dominant. Fig. (9 and 10) display the one dimensional curve where abscissa represents the Holder exponents actually present in sunspot data and ordinates are related to the amount of points where we encounter a given singularity. Multifractal analysis provided both a local and a global description of the singularities of a signal. The wavelet spectrum displayed in Fig. 6 and 8 reveals the time at which large variations and hence important in the investigation of multifractality. Various phenomena related to sun and space weather conditions show fractal and multifractal behavior when plotted against time. Multifractals have infinite number of dimensions associated with them.

(fact or authors opinion?)

The results of this work provide a general view of the multifractal pattern of sunspot number time series of between Solar Cycles 23 and ascending phase of Solar Cycle 24. This study reveals the multifractality in nature of sunspots and highly fractal towards the maximum phase of cycles. However, recent studies have shown that non-gaussian fluctuation is responsible for the presence of extreme events in space plasmas. Recently, using a non-extensive approach by Balasis et al. [56] found out the emergence of two distinct phases: (i) the phase where the intense magnetic storms cause a higher degree of magnetic field organization, and (ii) the phase which characterizes the normal periods with lower magnetic field coherence. The phase (i) may be associated with the presence of different kinds of large scale coherent structures, also pointed out by Chang et al. [57]. However, from the analysis it is found that Solar Cycle 23 commonly shows the multifractality nature around the initial, maximum and declining phase of cycle rather than Solar Cycle 24 it is not effective during the initial phase of cycle. Solar Cycle 24 has more fractal nature at the maximum phase and we wait for more result till complete the ongoing cycle.

Explanation

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