

Fractals Everywhere

What is a fractal

- A subset of a Euclidean space for which the Hausdorff dimension (roughness or chaos) strictly exceeds the topological dimension (Mandelbrot & Benoît B. ,1983).
- Fractals exhibit similar patterns at increasingly small scales called self similarity. As finer details are revealed, the form of the details is similar to the whole.

$$L(ar) = k L(r)$$

Fractal physiology

- Physiological signals are commonly irregular. Healthy physiological dynamics exhibit complex temporal fluctuations that are not simply due to uncorrelated random errors. Bassingthwaite et al., 1994
- Physiological systems possess “memory,” such that the change from one output event to the next displays a subtle “hidden” temporal structure.
- Such temporal fluctuations possess interesting properties, one of which is the property of long- range, fractal organization.
- Such dynamics are thought to exhibit the combination of critical properties of a control system: stability, flexibility, and adaptability.
- The presence of fractal properties in physiological and behavioral outputs is one critical marker of physiological complexity in humans

Fractal physiology (Quiet Standing)

- The postural control system does not behave chaotically, but rather it can be modeled as a correlated *random walk*, with almost perfect scaling behavior.
- Balance of a normal human being is the result of a most likely nonlinear superposition of sensory signals originating from the vestibular, visual, and somatosensory systems:
 - Visual system - 0-0.1 Hz
 - Vestibular system - 0.1-0.5 Hz
 - Somatosensory system - 0.5-1 Hz
 - Central nervous system - >1Hz
- In case of damages of one component of the balance system, certain characteristic frequency bands will change their relative weights

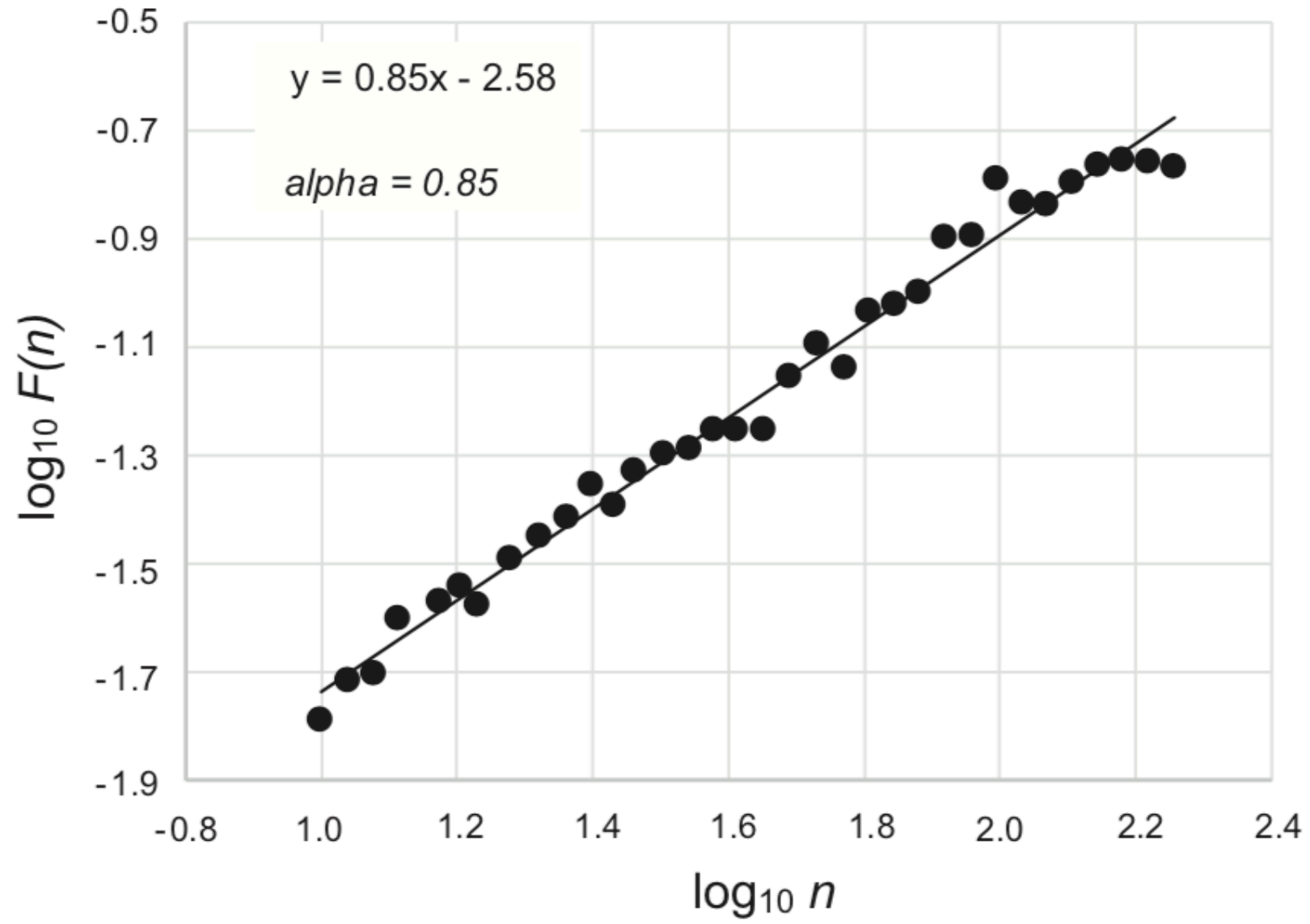
Turner et al., 2000; Blázquez et al, 2008; Sample et al., 2016;

DFA

- **Detrended Fluctuation Analysis:** One of the most popular methods to estimate the self-similarity and scaling property in fractal signals:
 - DFA computes the RMS error of linear fits over progressively larger bins (non- overlapped “boxes” of similar size, the scale) of an integrated time series.
 - Relationship between overall RMS error and the scale is the fluctuation function
 - When the fluctuation function follows a linear trend, the slope of this trend is scaling parameter α , which gives information concerning the correlation properties of a signal

PhysioToolkit (Goldberger et al., 2000)

DFA



DFA

- $\alpha = 0.5$ is characteristic of an uncorrelated random series (or white noise)
- $\alpha < 0.5$ the signal presents negative correlations
- $\alpha > 0.5$ positive correlations
- DFA correlation means that the displacement of CoM/CoP depends not only on the value of the most recent values, but also on the values of those at relatively remote times, thus indicating a memory effect in the behavior of the control system. ---> Very complex dynamics, whose optimal functioning seems to be chaotic.
- DFA allows us to discover the presence of long-range correlations, even in seemingly non-stationary time series
- DFA is insensitive to the length of the series

Warning

- **Choice of measurement and DFA parameters (box size, number of scales, etc) has a distinct influence on the computed metric**
- **Signal-to-noise ratios always influence outcomes, and time signals always contain noise (SNR has to be specified for assessing results)**

Müller et al., 2016

References

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