

# Dynamics of complex systems

## Lecture 4: Long-range correlations and scaling

**Dr. Maarten Wijnants & dr. Fred Hasselman**

Research Master Behavioural Sciences

Faculty of Social Sciences  
Nijmegen

*change perspective*

- Spectral analysis (power spectral density analysis)
- Detrended Fluctuations Analysis (DFA)
- Applications



High variability



Low variability

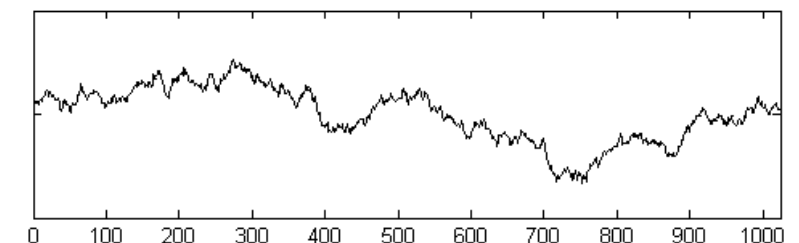
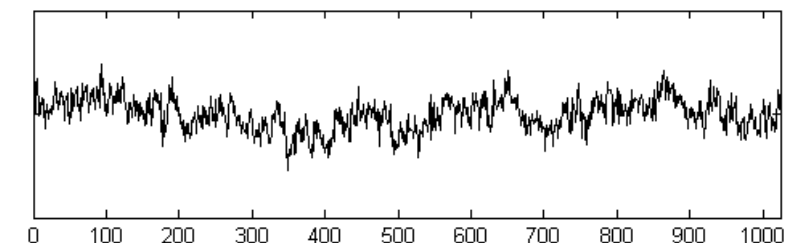
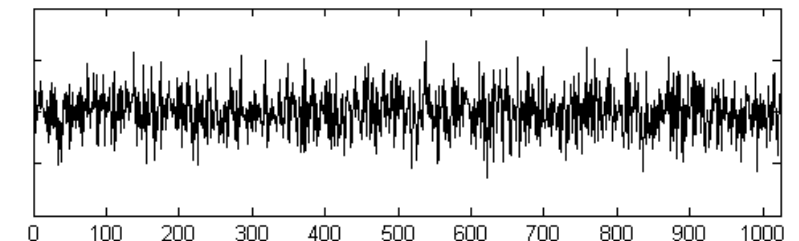
# Different faces of variability

## Variability is noise

- $X = T + E$

### ➔ Amount of variability

- Low  $\leftrightarrow$  High
- e.g., standard deviation



## Variability is structured

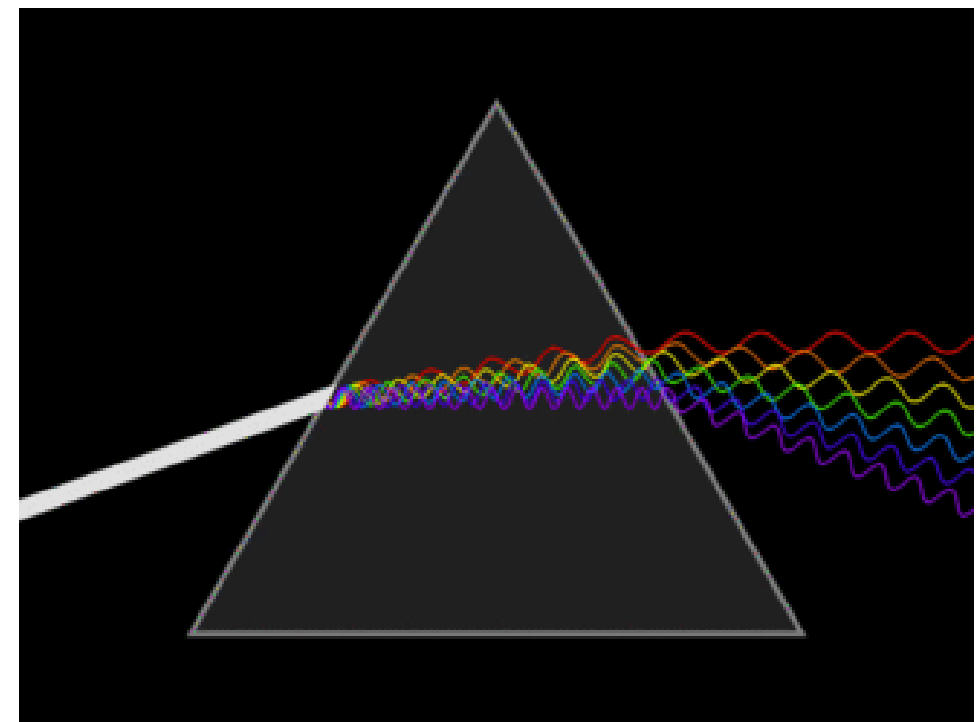
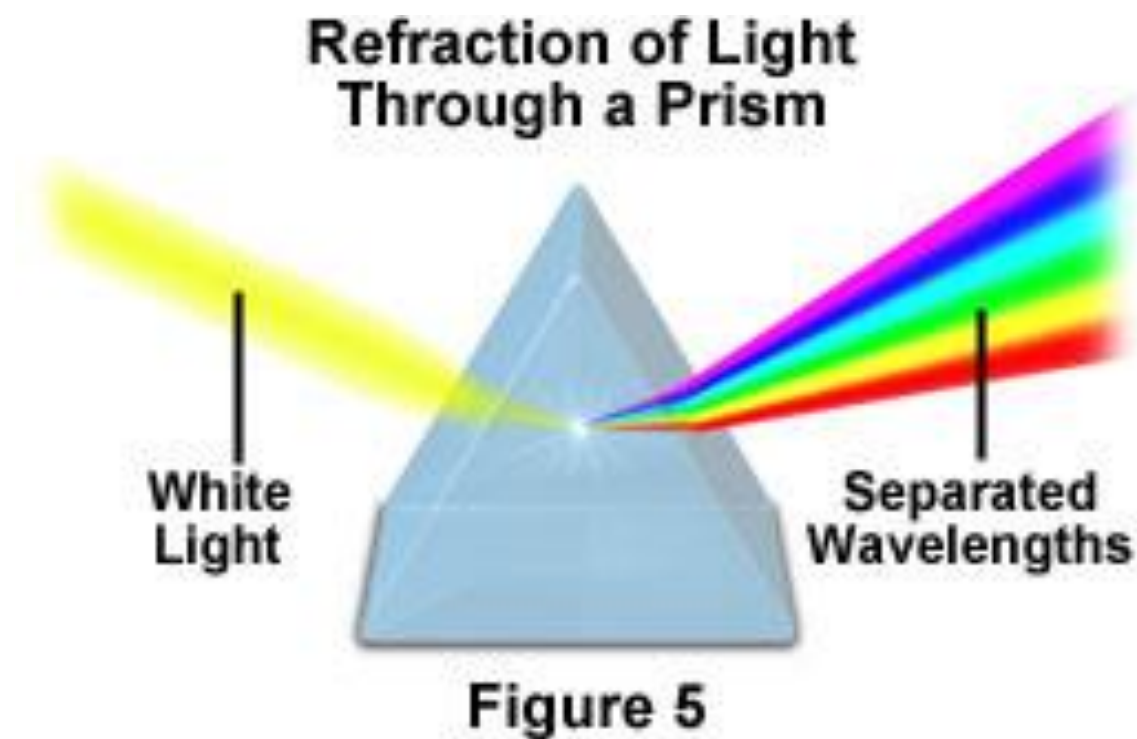
- $f(x) = 1/f^\alpha$
- Correlated vs. uncorrelated temporal structure

### ➔ Temporal correlations

- Structured  $\leftrightarrow$  Unstructured
- e.g., ACF, ...

# Frequency domain analysis

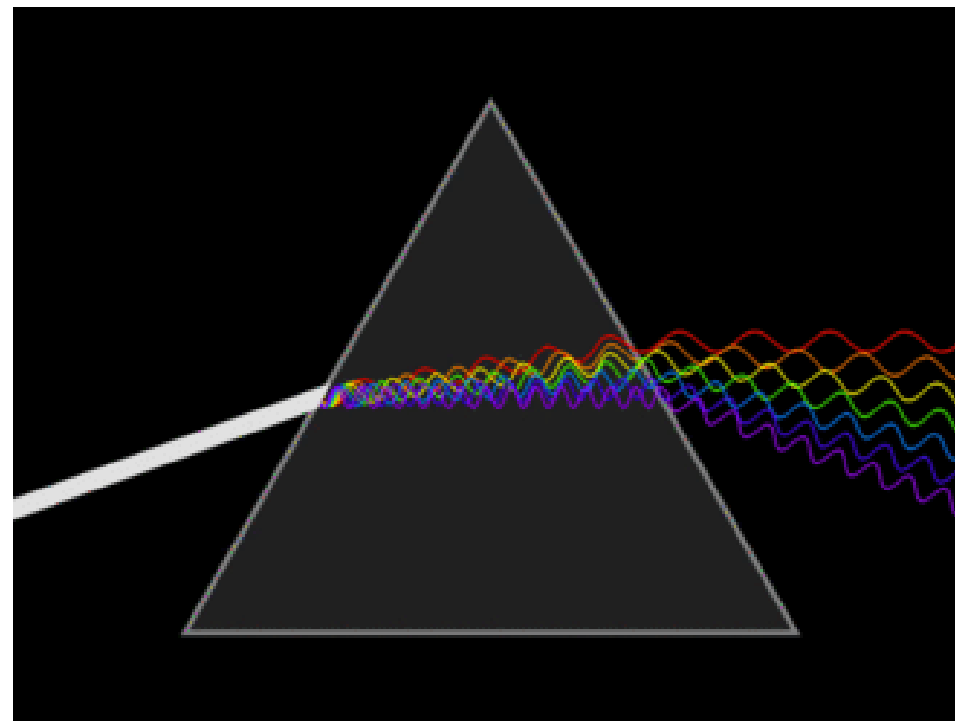
- Changes over time (ms)
- Frequencies of change (Hz)





# Frequency domain analysis

Changes over time →



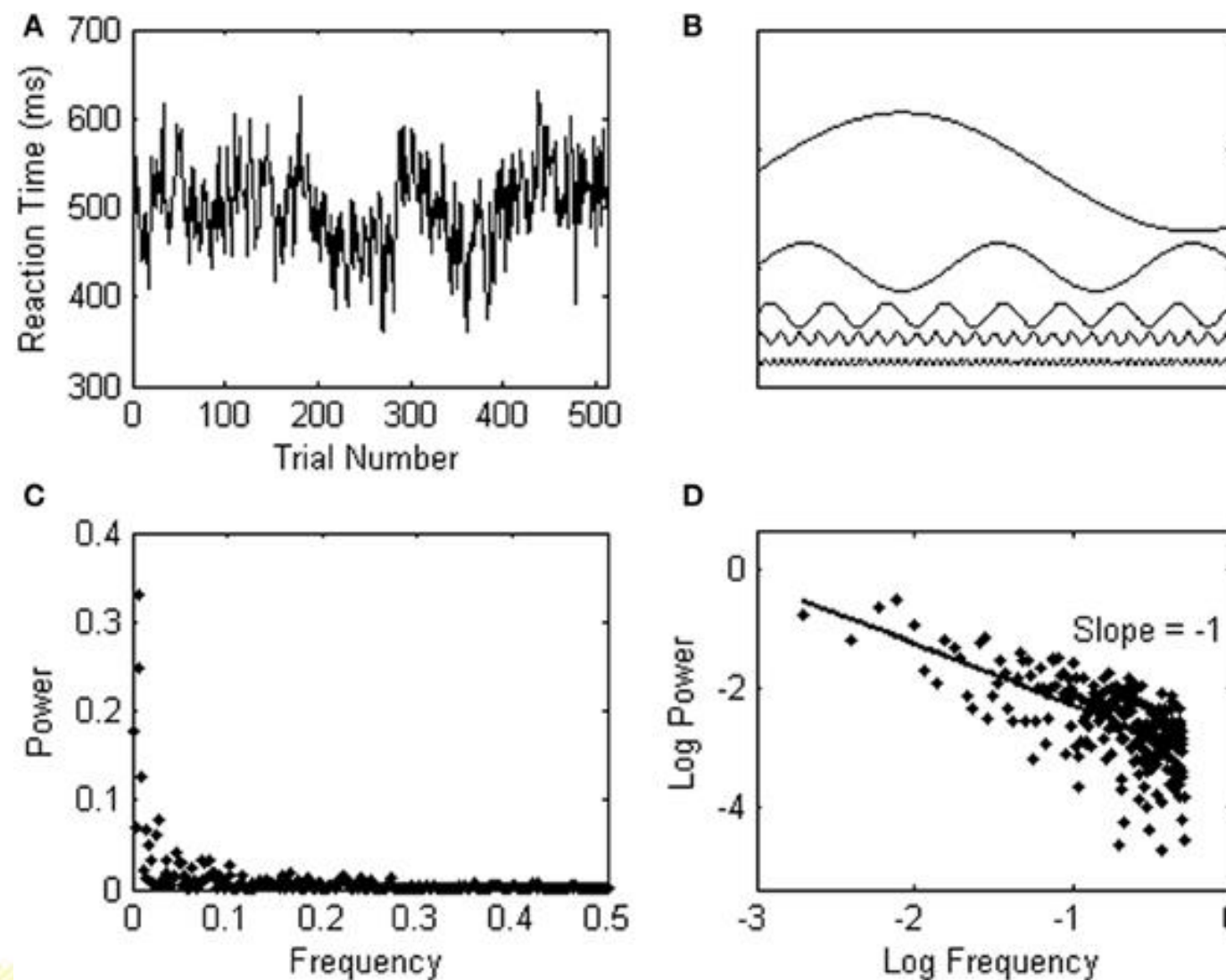
→ Constituent frequencies  
of a given amplitude

## Fourier Transform

# Spectral analysis: from time to frequency domain

Fourier transformation:

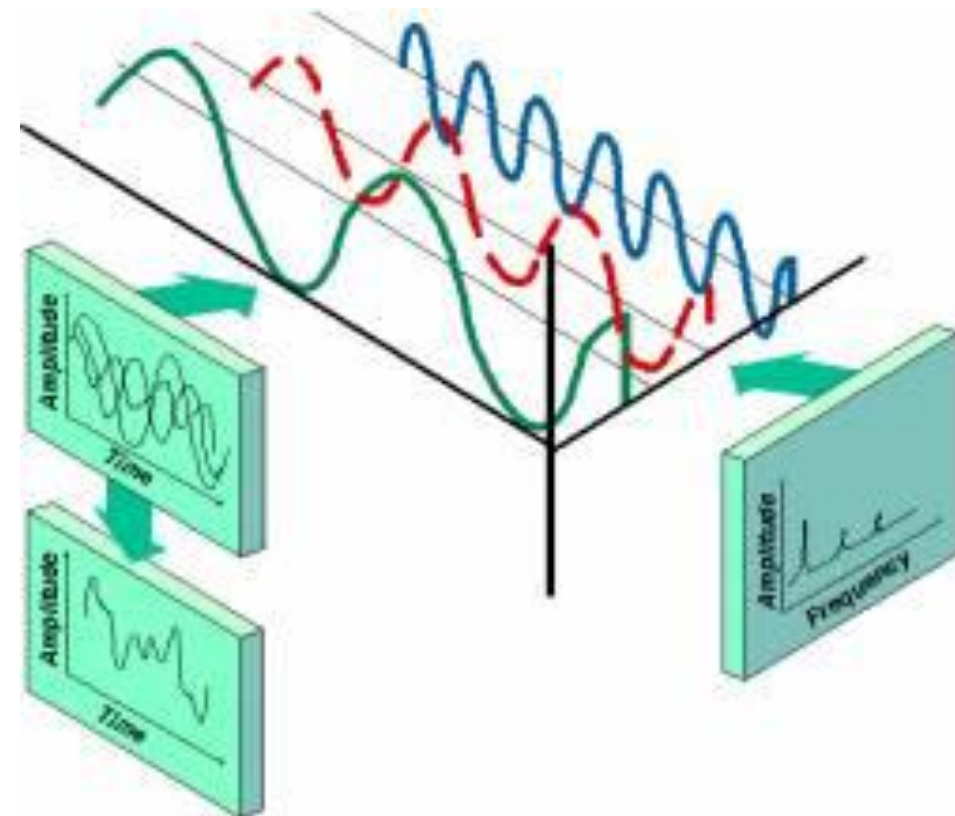
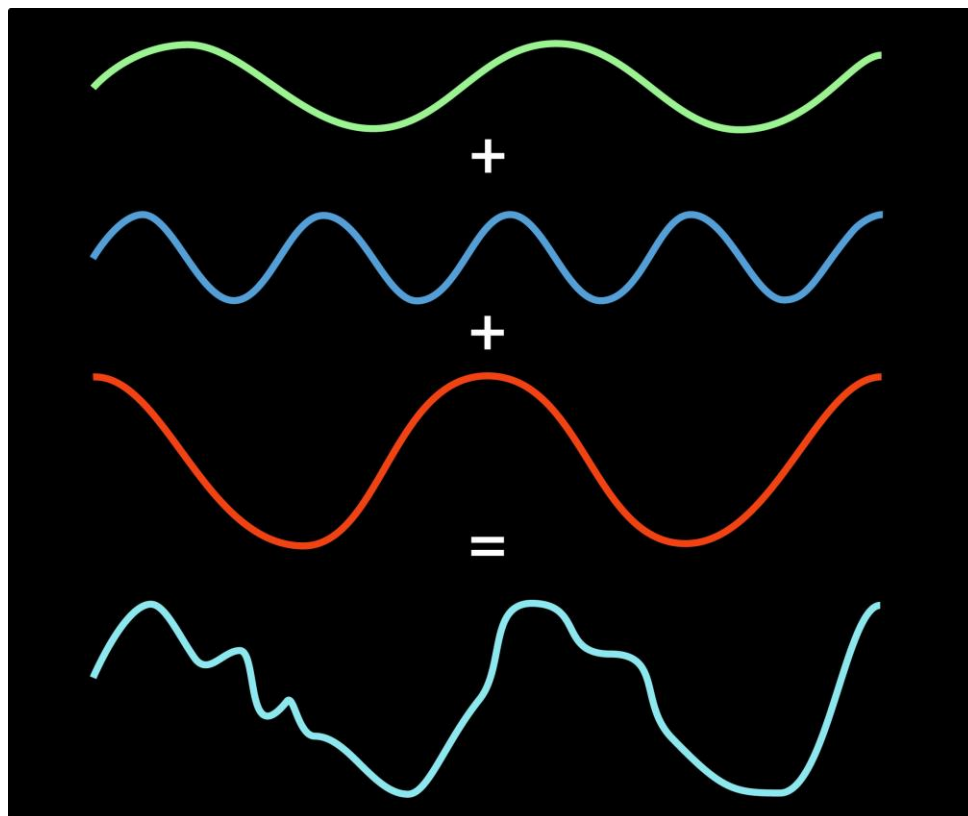
Any waveform can be duplicated by the superposition of sine and cosine waves



# Spectral analysis: from time to frequency domain

Fourier transformation:

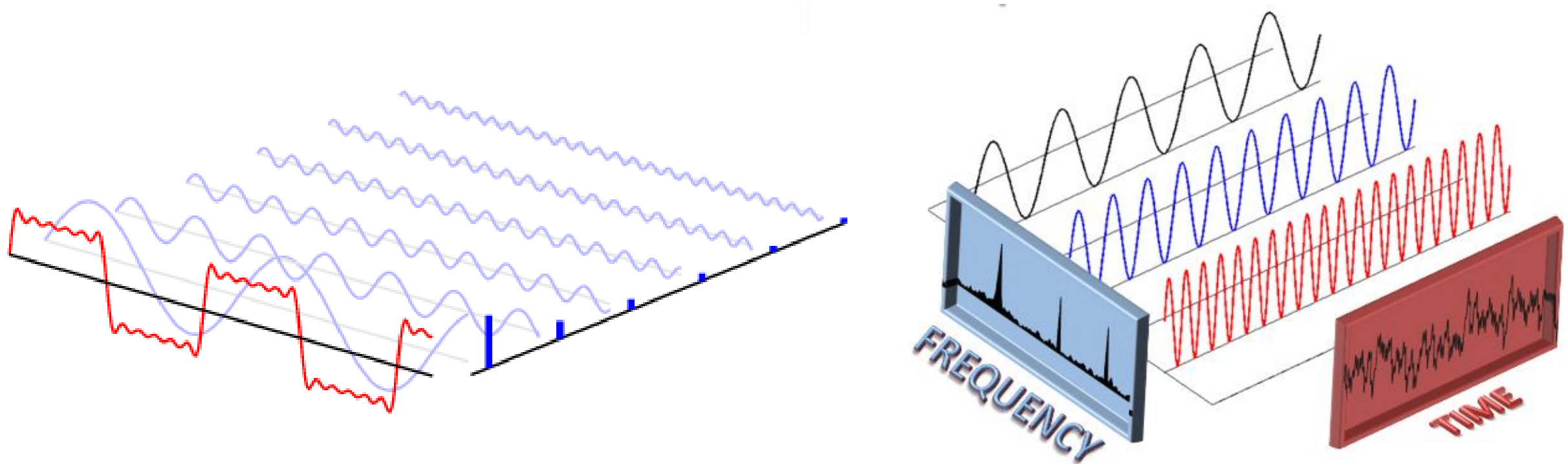
Any waveform can be duplicated by the superposition of sine and cosine waves



# Spectral analysis: from time to frequency domain

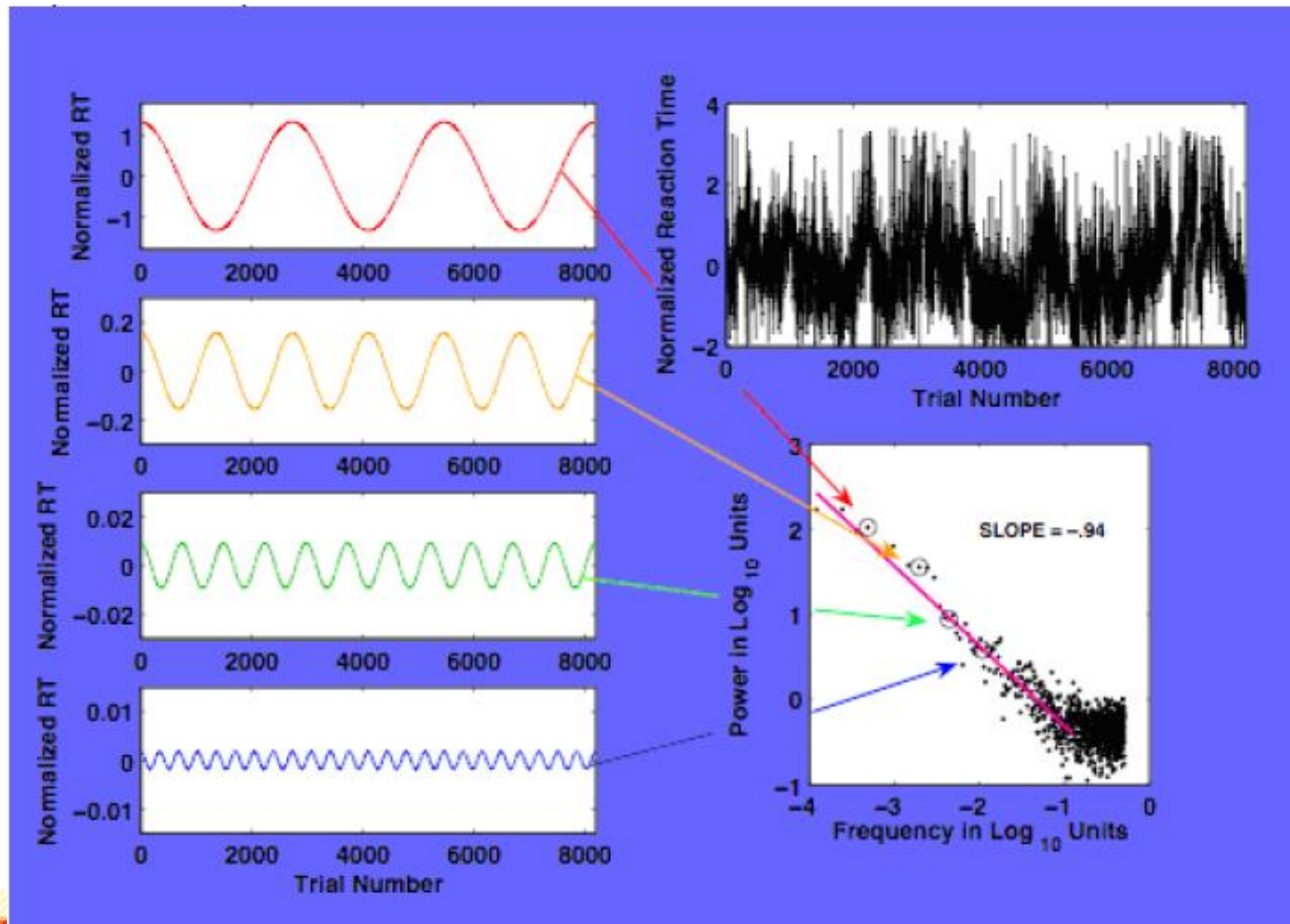
Fourier transformation:

Any waveform can be duplicated by the superposition of sine and cosine waves

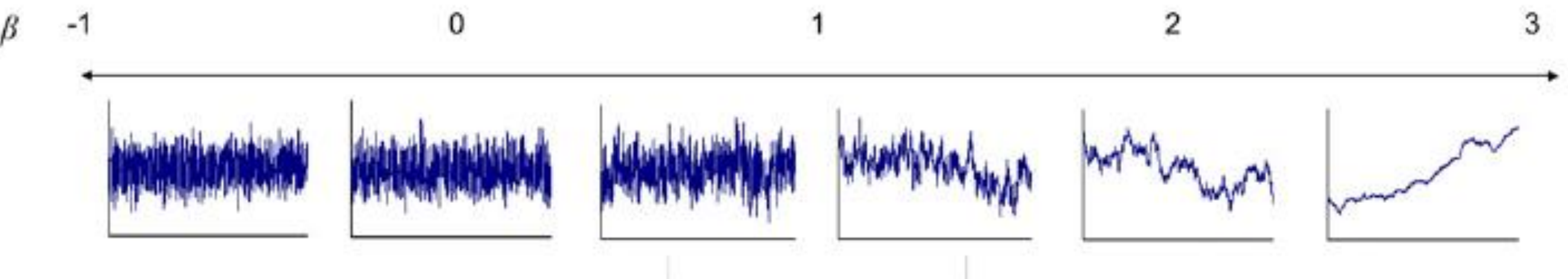




## Spectral analysis: from time to frequency domain



## Spectral analysis: from time to frequency domain

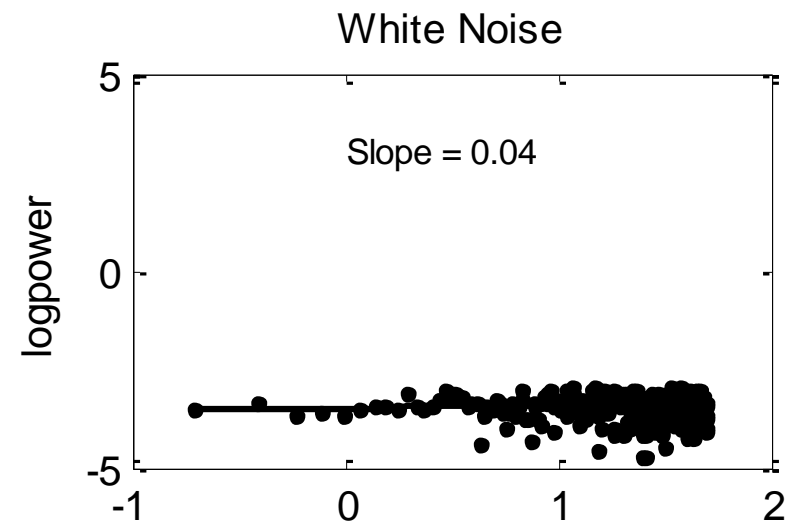
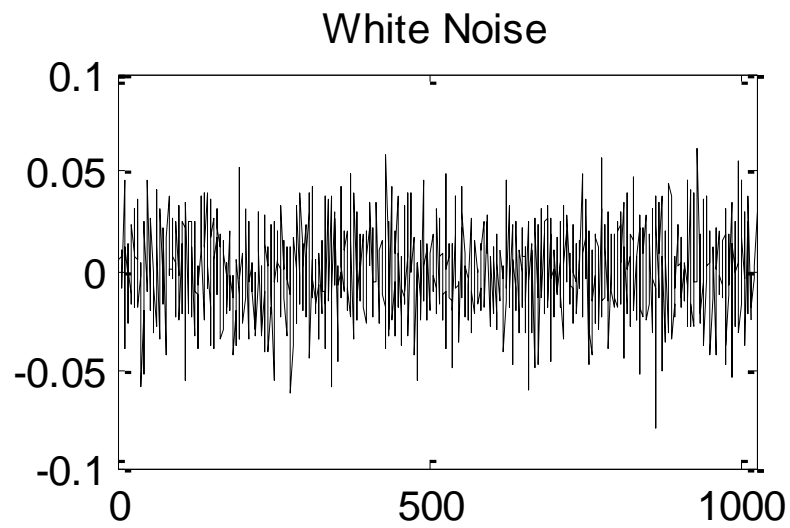


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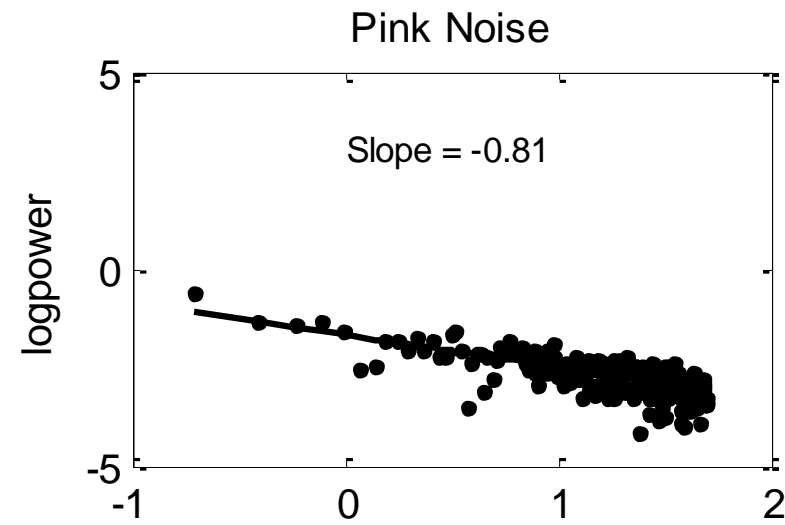
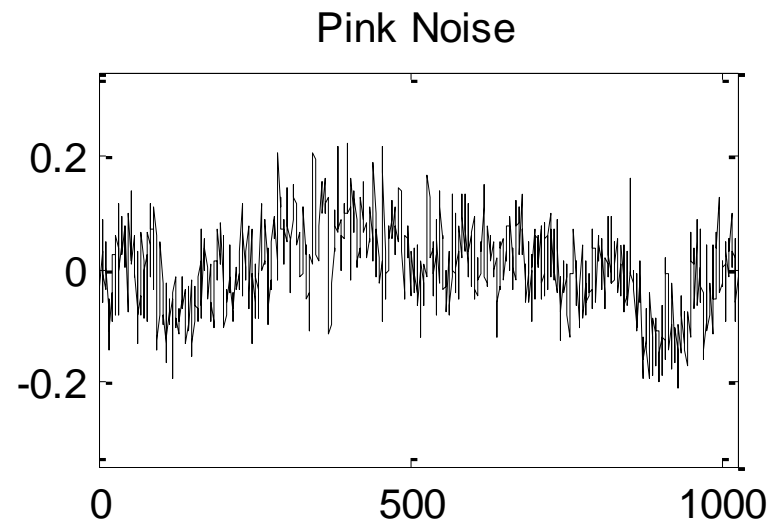
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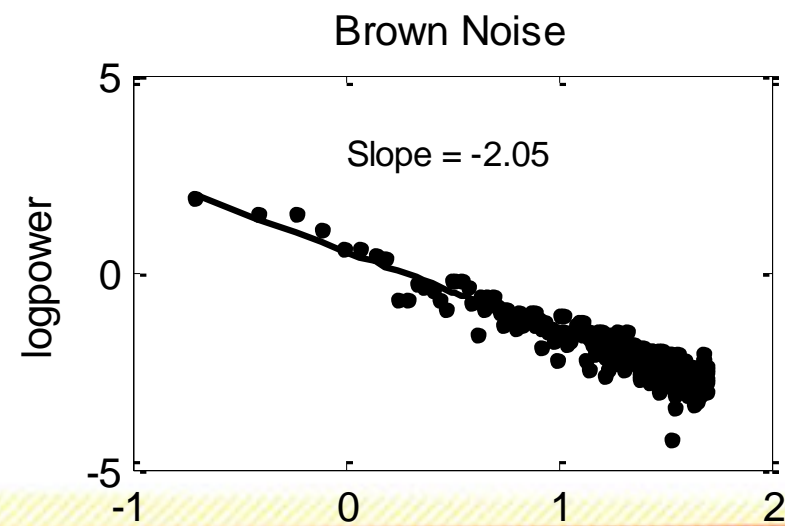
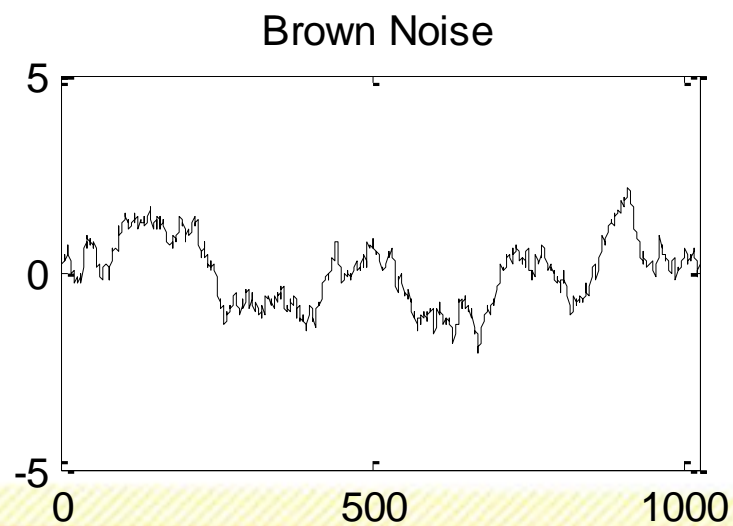
$$f(x) = 1/f^\alpha$$



Disorganized



Self-affinity



Rigid & persistent



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## Spectral analysis

- Only stationary signals
- Time series length
  - Power of 2
    - Truncation or zero-padding
  - Sufficiently long
- Linear trends may disrupt the outcome
  - May be seen as a low-frequency fluctuation
    - Linear (and quadratic) detrending prior to analysis

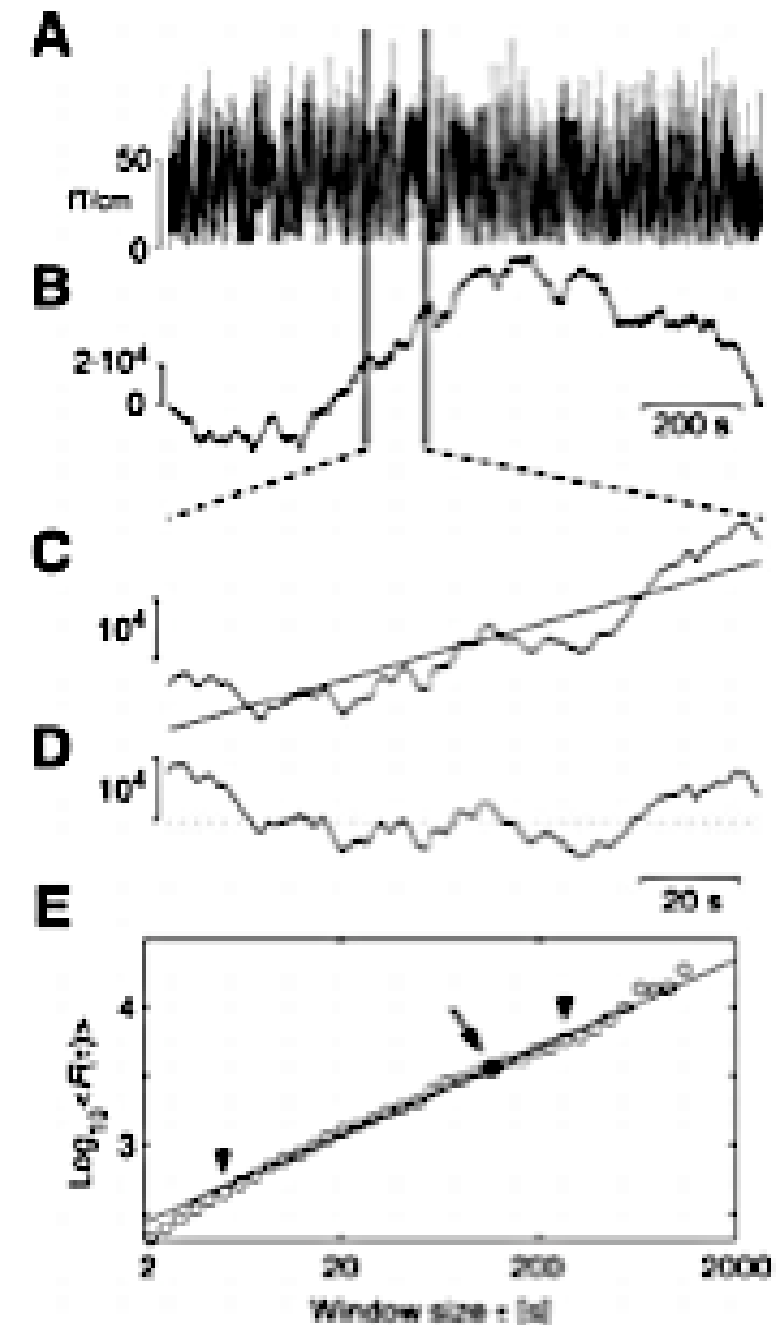


## Detrended Fluctuation Analysis

Same logic as SDA except:

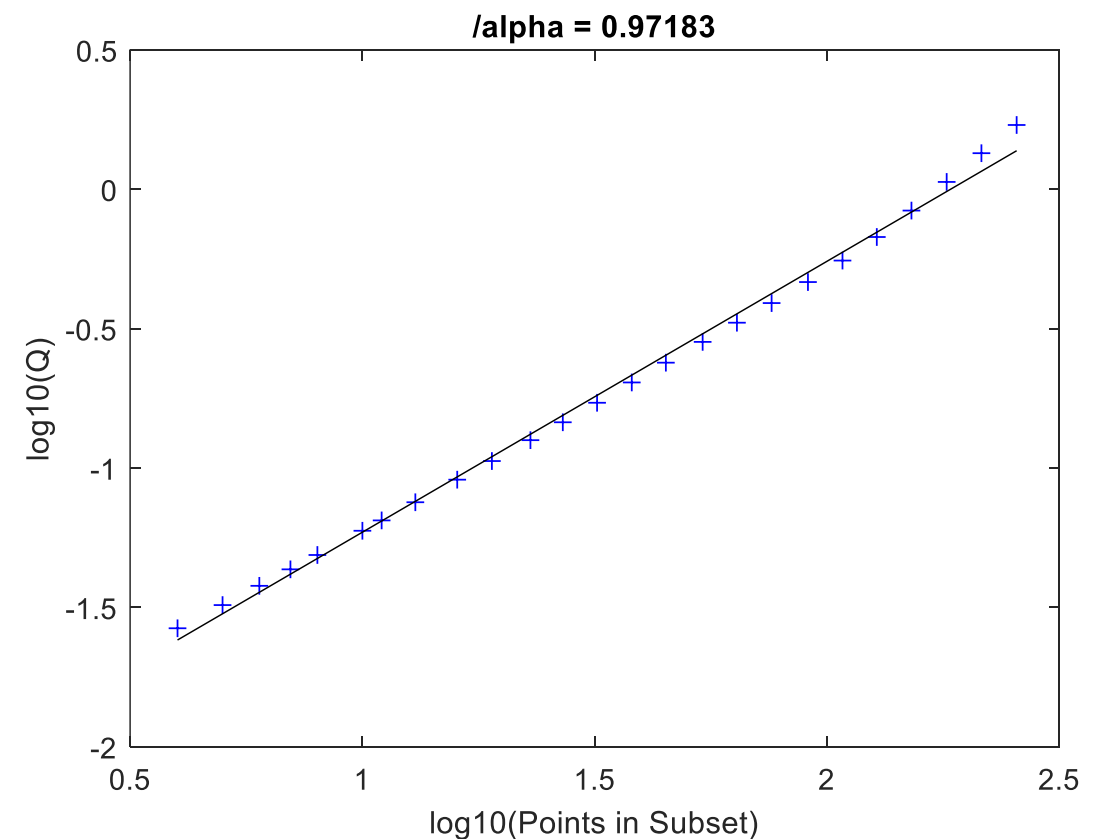
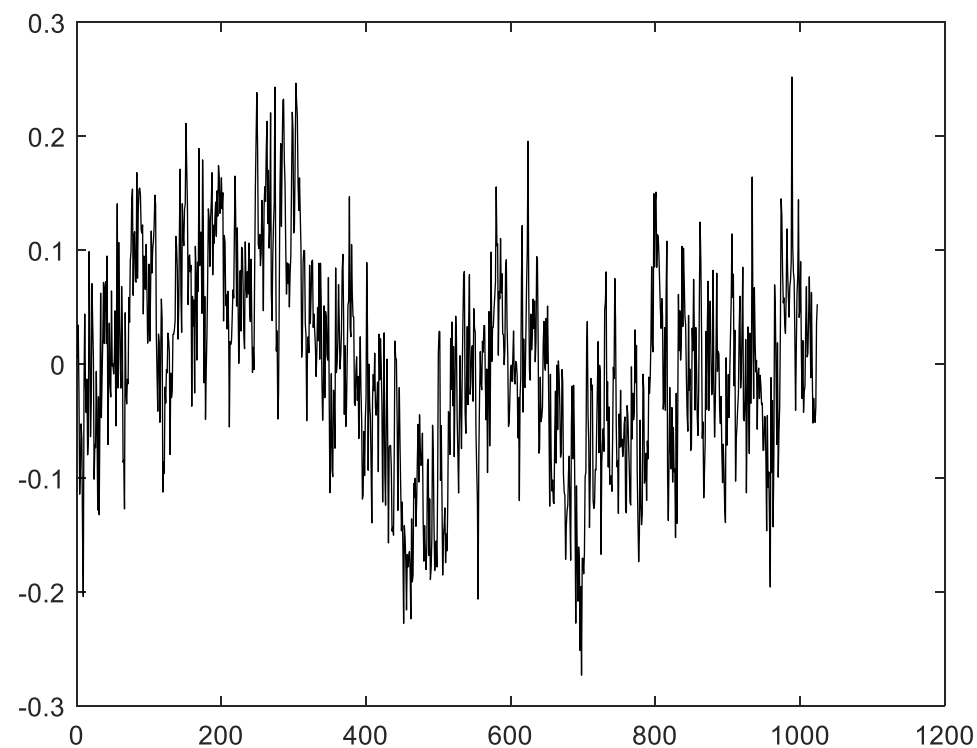
- **A.** Signal is integrated
- **B.** Divided into bins
- **C.** Detrended (linear)
- **D.** Remaining SD is the dispersion measure
- **E.** Plot on  $\text{Log}_{10}$  scale and calculate slope (alpha)

Or **C** & **D** in one step: fit a line in the bin and take SD of residuals... same result.



## DFA: Pink noise

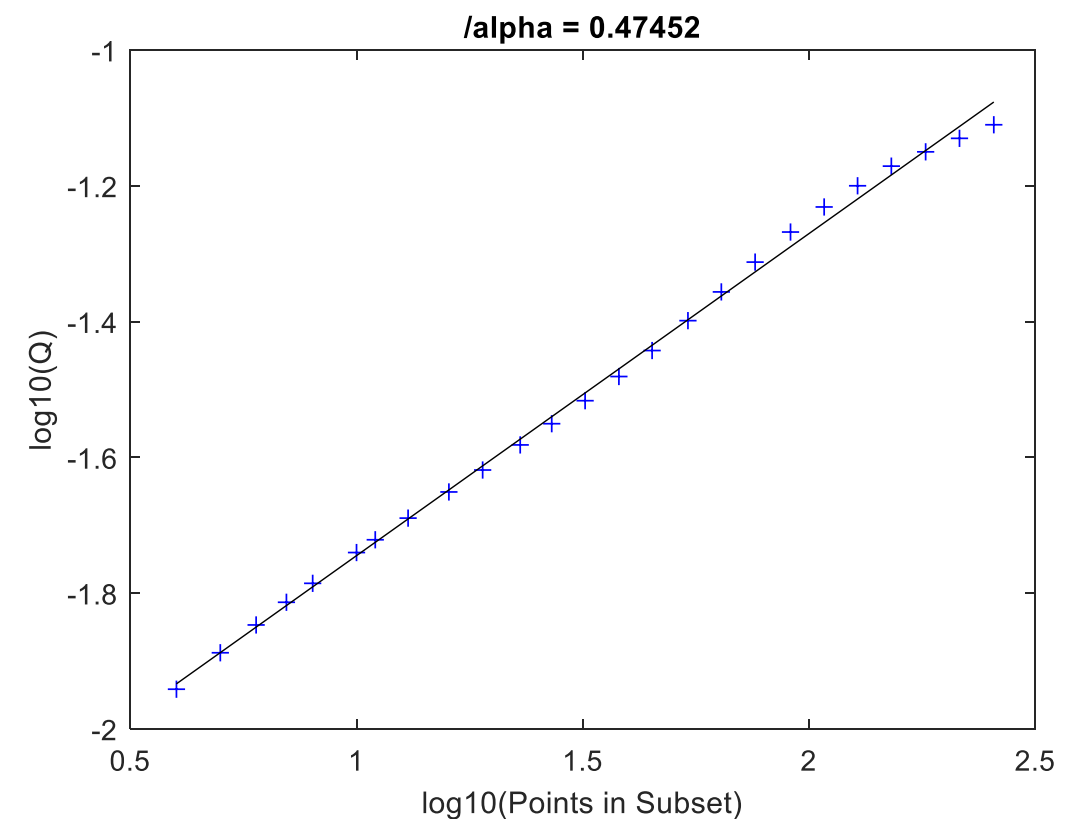
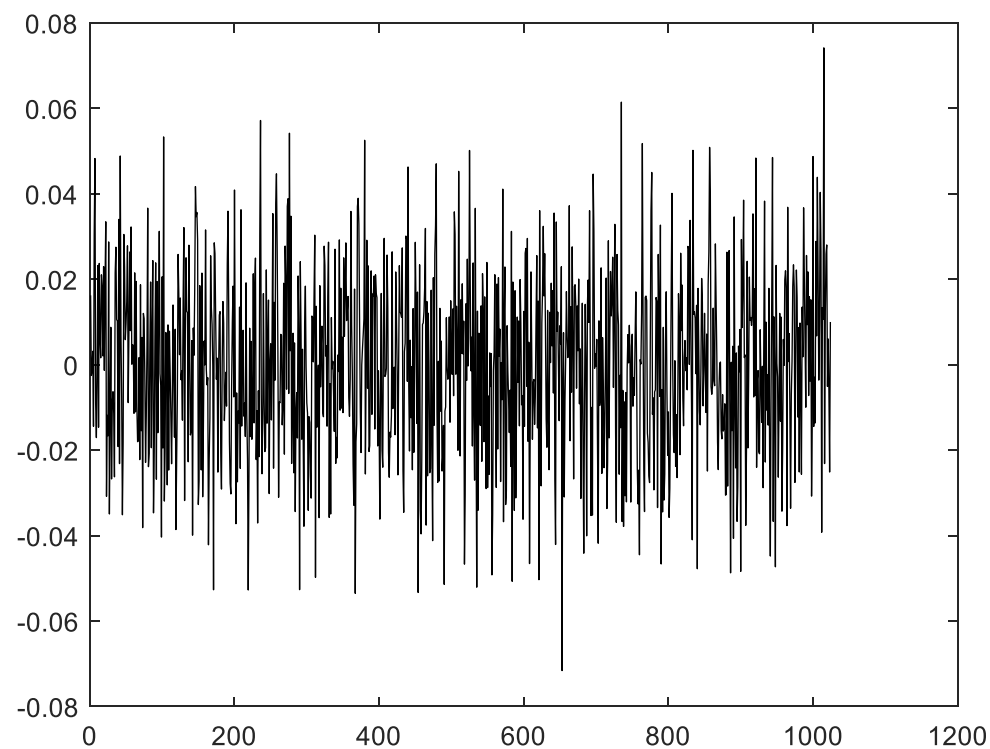
- Fit the slope:
  - $\alpha = .5 \rightarrow$  white noise
  - $\alpha = 1 \rightarrow$  pink noise
  - $\alpha = 1.5 \rightarrow$  Brownian noise



$$0.97183x + -2.20174 = y, \quad r^2 = 1.00, \quad H = 0.972, \quad D = 1.028$$

## DFA: White noise

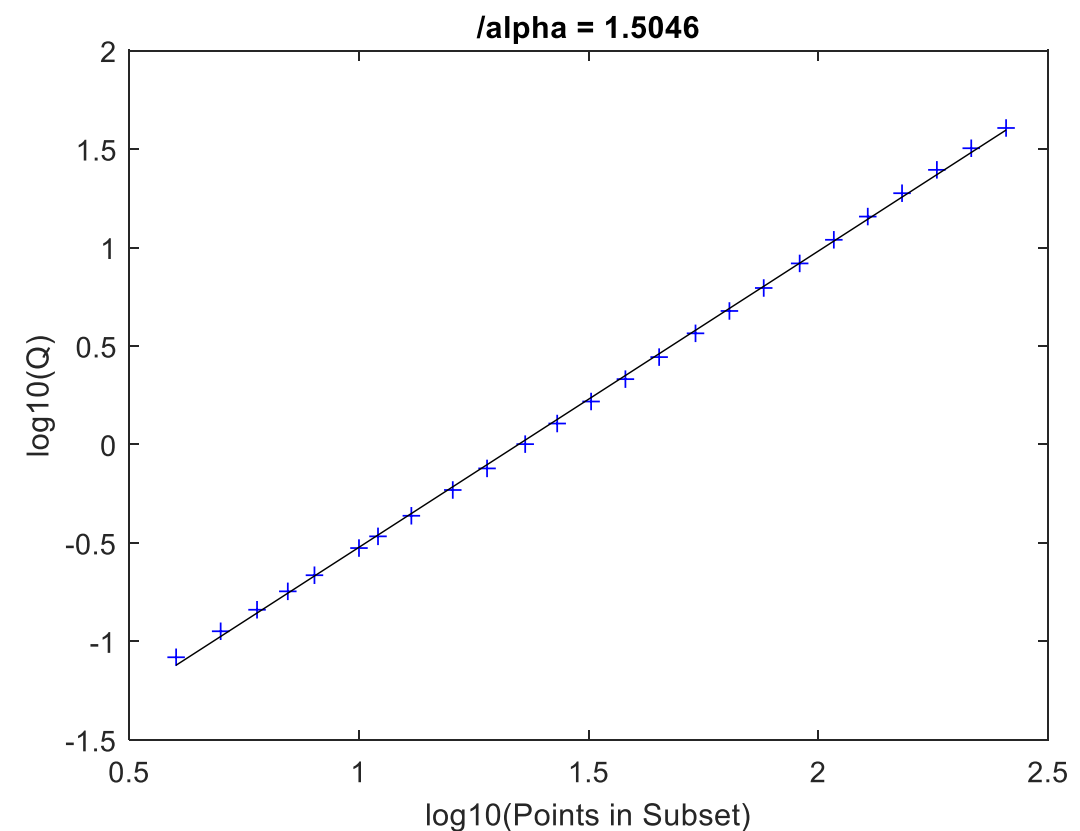
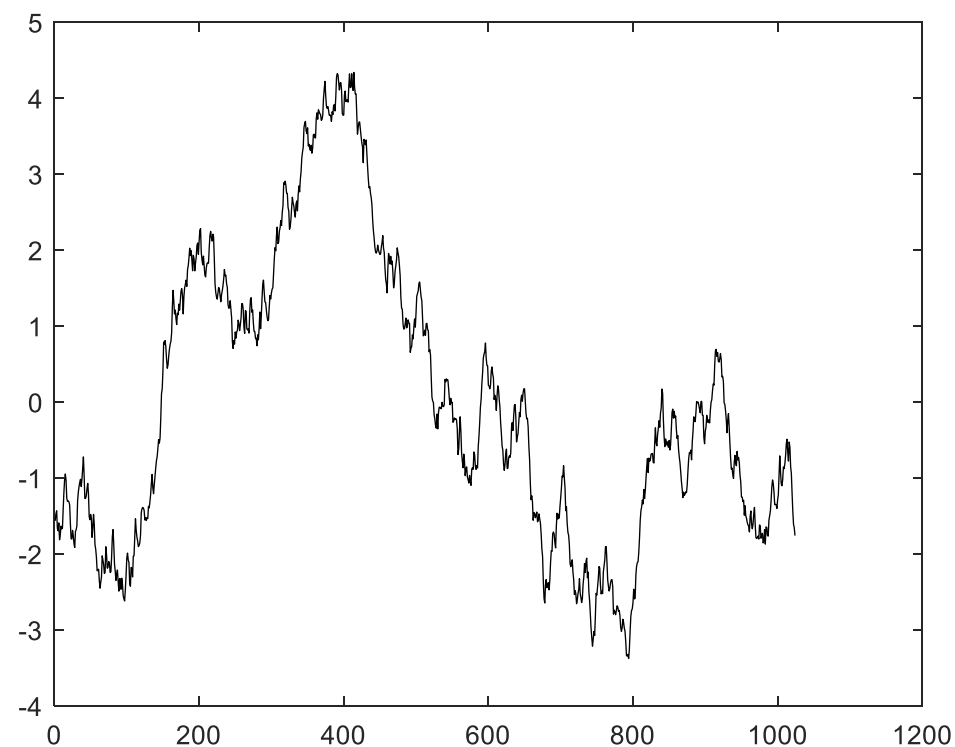
- Fit the slope:
  - $\alpha = .5 \rightarrow$  white noise
  - $\alpha = 1 \rightarrow$  pink noise
  - $\alpha = 1.5 \rightarrow$  Brownian noise



$$0.47452x + -2.21923 = y, \quad r^2 = 1.00, \quad H = 0.475, \quad D = 1.525$$

## DFA: Brownian noise

- Fit the slope:
  - $\alpha = .5 \rightarrow$  white noise
  - $\alpha = 1 \rightarrow$  pink noise
  - $\alpha = 1.5 \rightarrow$  Brownian noise



$$1.50456x + -2.02672 = y, \quad r^2 = 1.00, \quad H \approx 1.505 \quad D = 0.495$$



## DFA

- Works for non-stationary signals
- Time series length
  - Power of 2
  - Sufficiently long
- Linear trends may do not disrupt the outcome
  - Detrending is an inherent part of the calculation

# Applications

# Empirical timeseries

- Do empirical timeseries actually reveal  $1/f$  noise?
- Does the value of your scaling exponent matter?
  - Medicine & Physiology
  - Neuroscience
  - Cognition

## Fractal physiology

- A healthy heart fluctuates a  $1/f$  noise
- Deviations from  $1/f$  noise correlate with mortality risk (Mäkikallio et al., 2001)
  - Congestive heart failure
  - Ventricular arrhythmia (Goldberger, 1997; Peng et al., 1995)
- Smaller deviations from  $1/f$  noise
  - Aging (Goldberger, 2002)
  - Obese children (Vanderlei, Pastre, Júnior, & de Godoy, 2010)
  - Adults with down syndrome (Mendonca, Pereira, & Fernhall, 2011)



# Fractal physiology

- A healthy respiratory system emits  $1/f$  noise
- Breathing rhythm
  - $1/f$  noise  $\rightarrow$  white noise
    - Aging (Peng et al., 2002; West, 2006)
  - White noise  $\rightarrow 1/f$  noise
    - With gestational age in fetal development (Govindan, Wilson, Murphy, Russel, & Lowery, 2007)
- Also:
  - Asthma patients with more pronounced  $1/f$  signatures in breathing rhythm show better recovery after treatment (Frey et al., 2005).

# Fractal physiology

- Blood pressure fluctuates as  $1/f$  noise (Mutch et al., 2000, Brogan et al., 2007)
  - Diabetic patients show reduced  $1/f$  noise in glucose fluctuations in the blood flow compared with healthy controls (Ogata et al., 2007; Yamamoto et al., 2010).

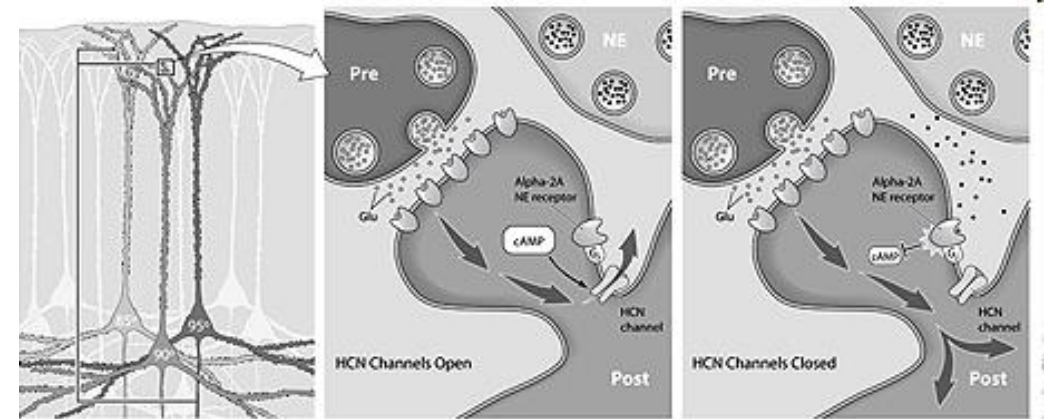
# Fractal physiology

- Colon pressure
  - Patients hospitalized for slow transit constipation showed colon pressure fluctuations deviating from  $1/f$  noise towards Brown noise (Yan, Yan, Zhang, & Wang, 2008).

# Fractal neuroscience

- Ion Channels Opening and Closing Times

- (Liebovitch & Krekora, 2002; Liebovitch & Shehadeh, 2005; Lowen, Cash, Poo, & Teich, 1997; Takeda, Sakata, & Matsuoka, 1999, Varanda, Liebovitch, Figueiroa, & Nogueira, 2000)

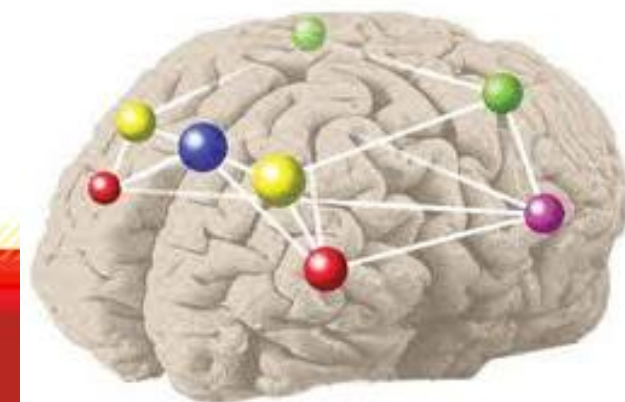
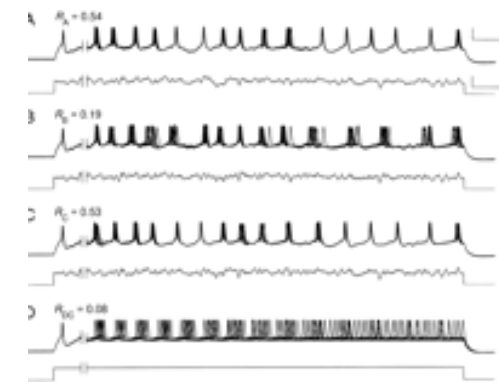


- Neural Spike Intervals

- (Bhattacharya, Edwards, Mamelak, & Schuman, 2005; Giugliano, Darbon, Arsiero, Luescher, & Streit, 2004; Grüneis et al., 1993, West & Deering, 1994)

- Larger Scale Neural Assemblies

- (Buzsàki, 2006; Bressler & Kelso, 2001; Freeman, Holmes, Burke, & Vanhatalo, 2003; Spasic, Kesic, Kalauzi, & Saponjic, 2010; Tognoli & Kelso, 2009; Varela, Lachaux, Rodriguez, & Martinerie, 2001; Werner, 2007)





# Fractal neuroscience

- Deviations from  $1/f$  noise in EEG
  - Major-Depressive Disorder (Linkenkaer-Hansen et al., 2005)
  - Mania (Bahrami, Seyedsadjadi, Babadi, & Noroozian, 2005)
  - Autism (Lai et al., 2010)
  - Epilepsy (Ramon, Holmes, Freeman, McElroy, & Rezyanian, 2008)
  - Alzheimer's Disease (Abásolo, Hornero, Gómez, García, & López, 2008)
  - ...

## Some comprehensive review and discussion papers

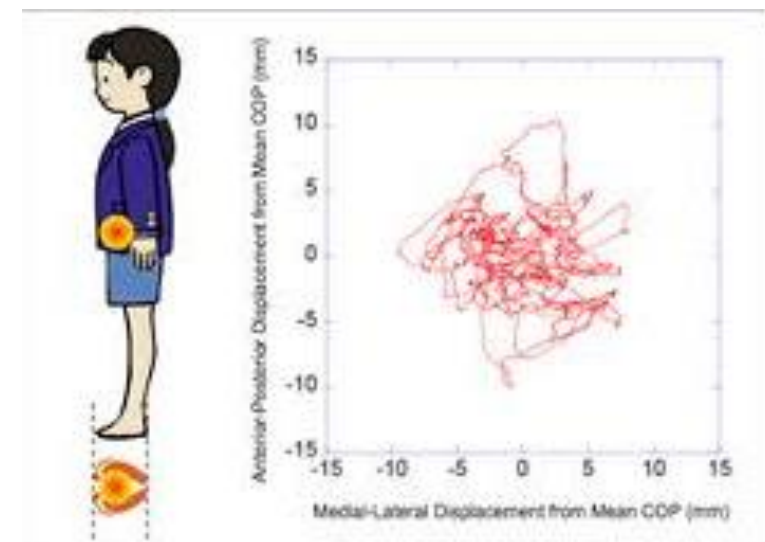
Temporal complexity is commonplace in different disciplines:

- Fractals in medicine
- Fractals in dentistry
- Fractals in pathology
- Fractals and cancer
- Fractals in finance
- Fractals in cell biology
- Fractals in cosmology
- Fractals in the neurosciences
- A link to the Frontiers in Fractal Physiology journal

# Motor Control

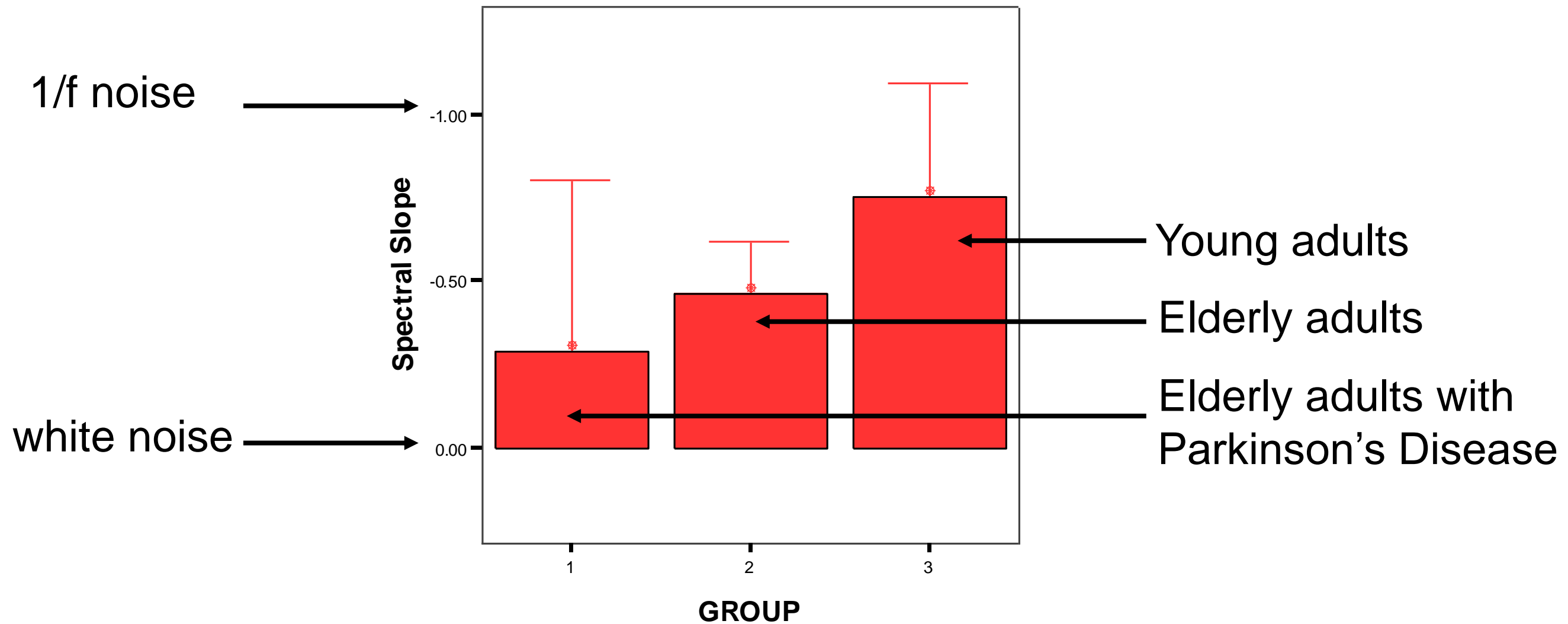
## Posture

- A decrease in postural stability is accompanied by deviations from 1/f noise (Hong et al., 2006)



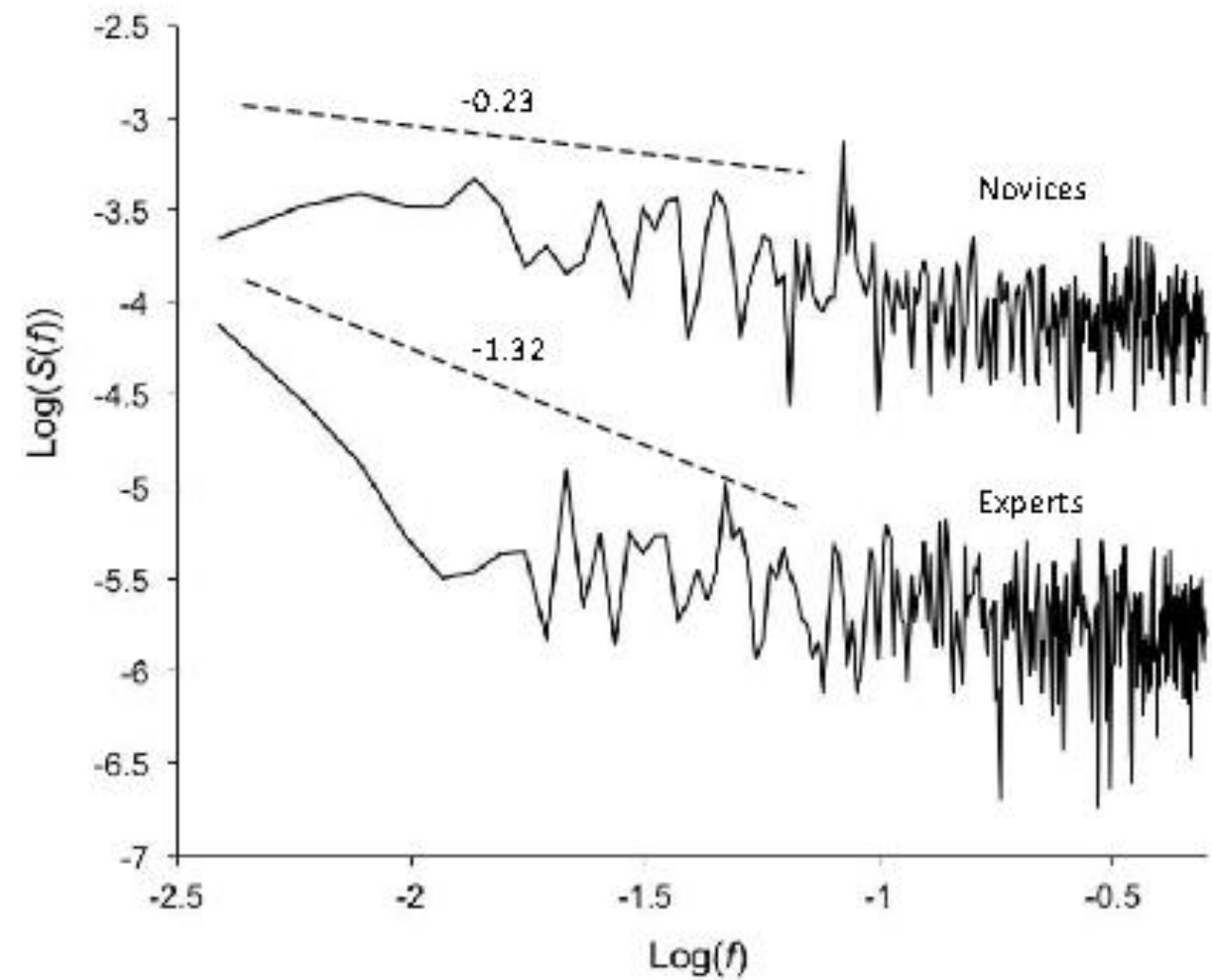
Correlation fractal dimension and falling risk in elderly ( $r = .78$ ,  $p < .0005$ ; Hausdorff, 2007)

# Gait intervals



(Hausdorff, 2007)

# Skiing



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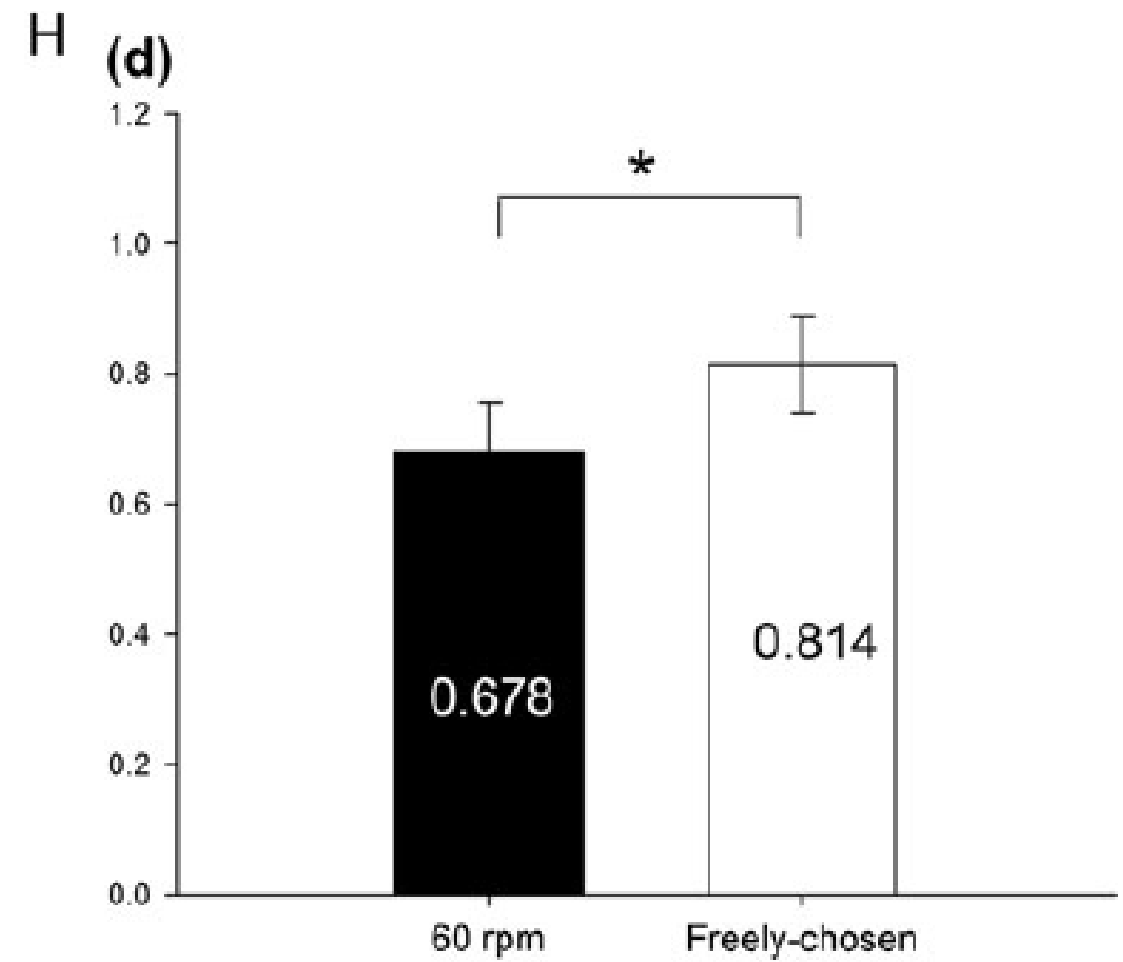
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Nourrit-Lucas et al., 2014





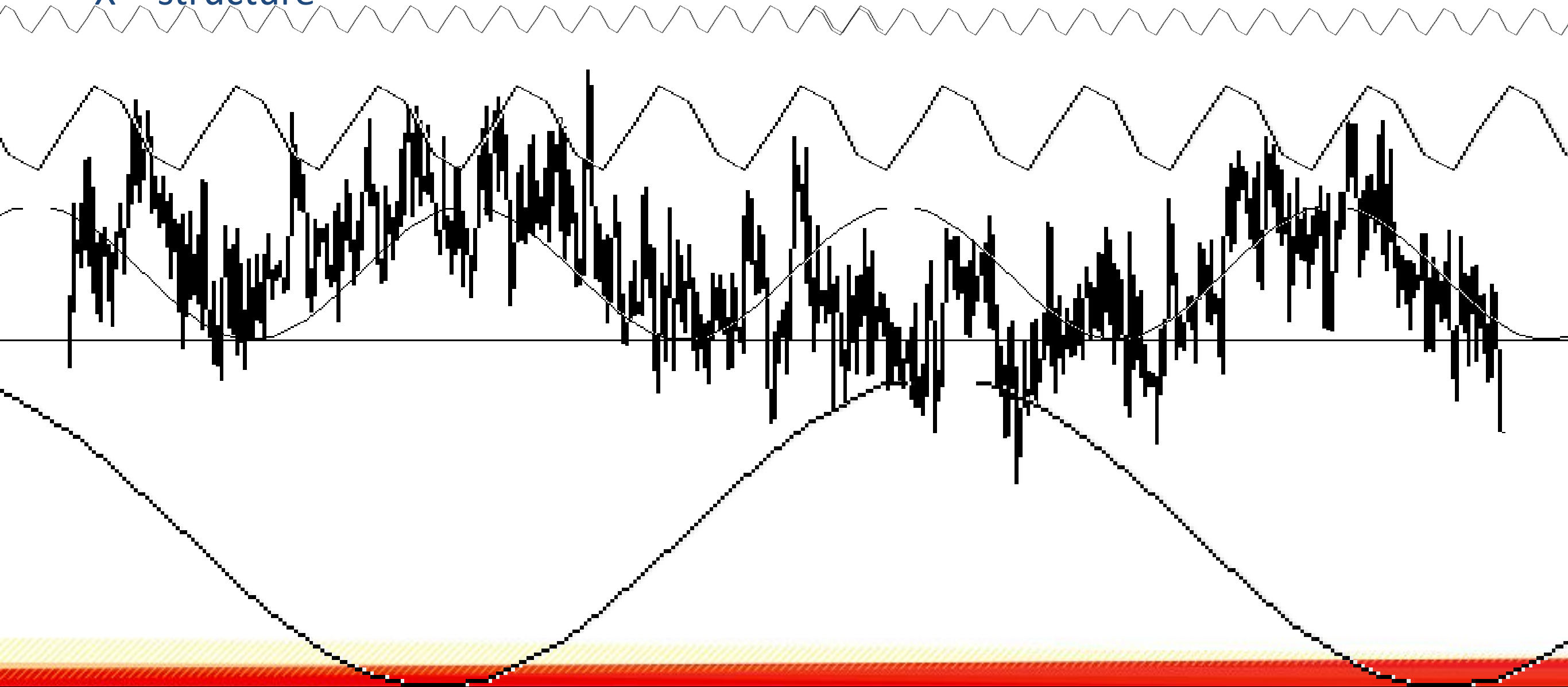
# Cycling



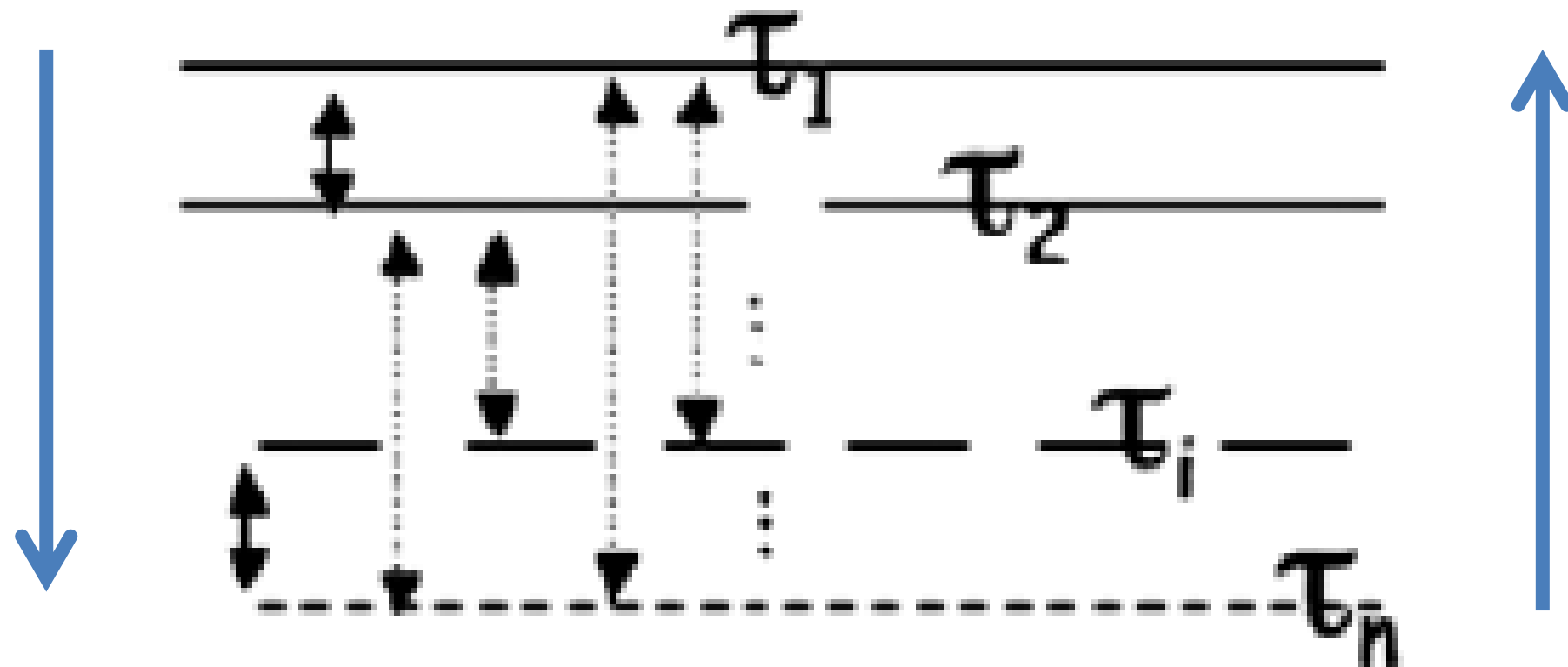
# What can we learn from $1/f$ noise in cognitive performances?

Maarten Wijnants

$X = \text{structure}$



*change perspective*



Changes on multiple time scales are coupled to changes on other timescales

- 1/f noise turns up in a lot of time series in psychological experiments

## Elementary “production” tasks:

Repeated production of a Spatial or a Temporal interval

## Elementary “Motor” tasks:

Postural sway

Tapping and walking

Swinging pendula

Juggling

## Patterns of mood change:

Repeated judgments of self-esteem (2 per day, over 512 days!)

## Classic “Cognitive” tasks:

Simple reaction time

Perceptual learning

Visual search

Classification

Word naming

Lexical decision

Mental rotation



# 1/f scaling and cognition

- Individual response times provide an incomplete description of actual cognitive performances
  - 'basic features of a performance cannot simply be averaged out.'
- Interaction-dominant dynamics
  - 1/f emerges through coordinated interactions between components
  - Components at different scales change each others dynamics
  - No statistically independent components:
    - A single process extends across all time scales of variation

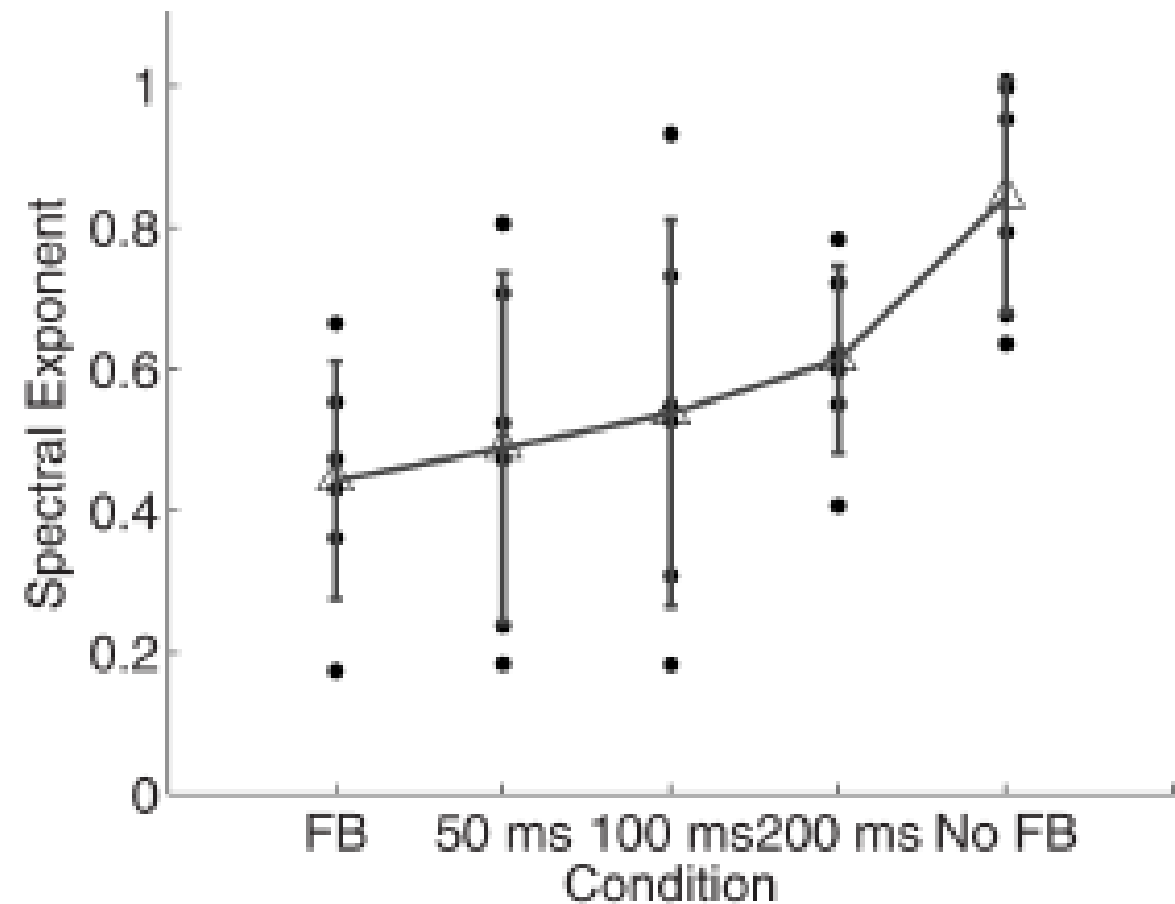
e.g. Holden, Van Orden & Turvey, 2008

# Coordination

- Proof of the pudding
  - Can we systematically manipulate the strength of long-range correlations?
    - E.g. skilled performances
  - Can we separate internal fluctuations (system dynamics) from external fluctuations (perturbations)?

## Task constraints

External perturbations add extraneous random variation to the measured performances: Accuracy feedback



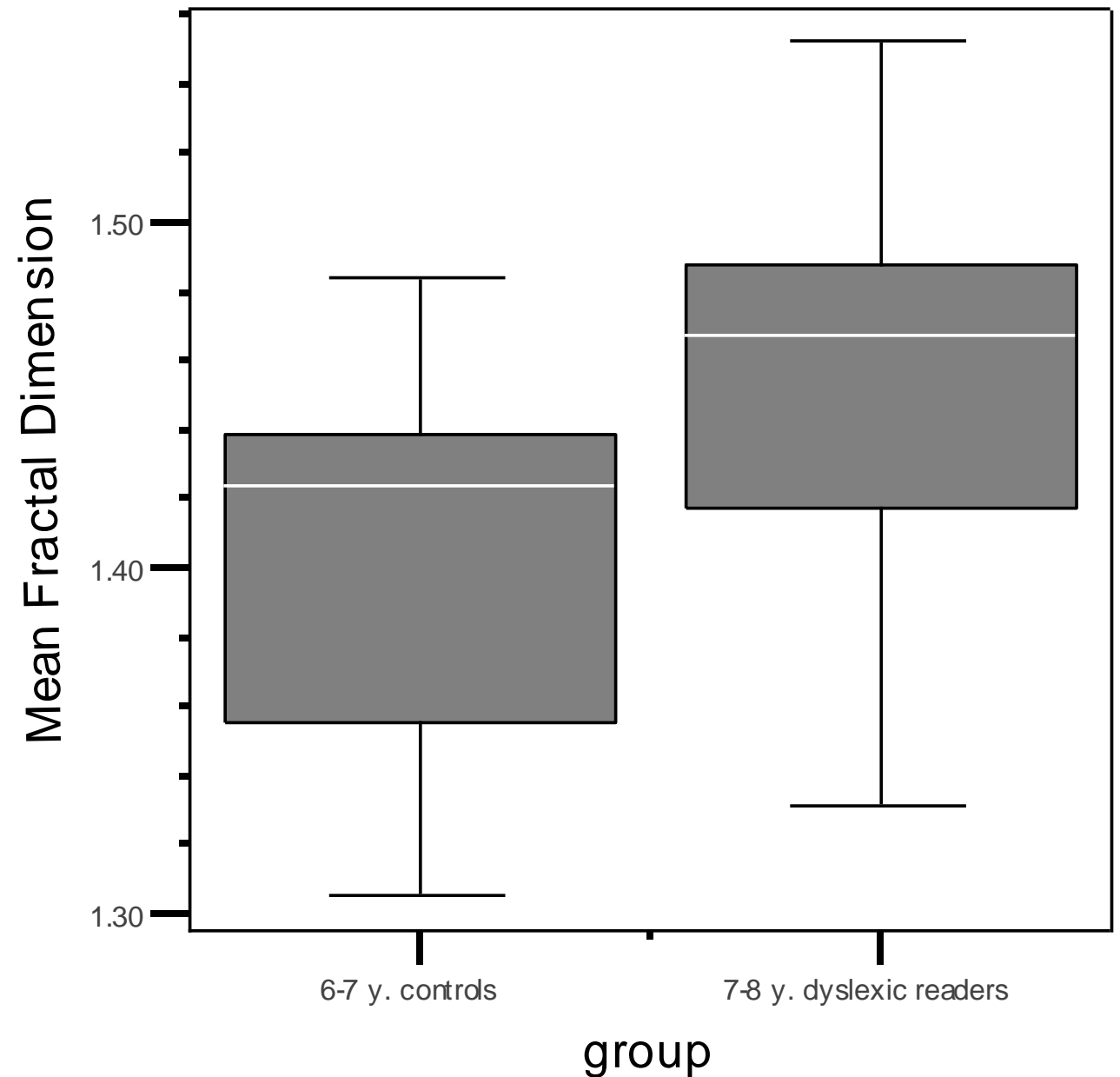
**FIGURE 3 | Spectral exponents of the time estimates.** Spectral exponents  $\propto$  closer to 0 imply presence of white noise whereas values closer to 1 suggest pink noise. Individual points represent observations from individual participants. Error bars plot within-condition SD.

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Kuznetsov & Wallot (2011)

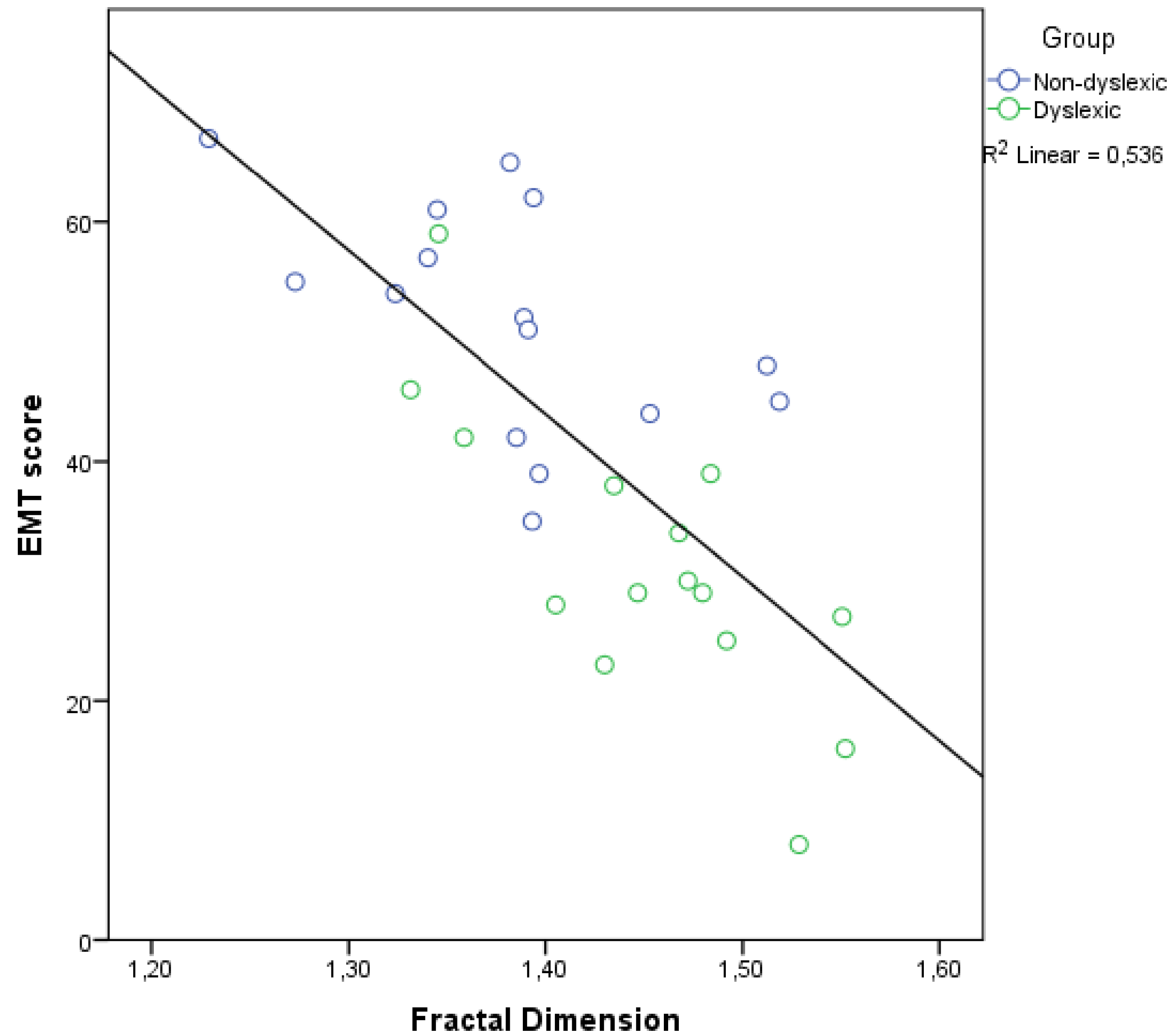
## Word-naming

- 7-8 year old Dyslexic Readers
- 6-7 year old Controls
- 1 Block of 550 Word Stimuli
- ➔ Dyslexic Readers Show Reduced 1/f Noise



## Word-naming

- Oral reading fluency is regarded as the sole best indicator of reading problems (Fuchs, Fuchs, Hosp, & Jenkins, 2001)





## Within-Group Correlations: dyslexics vs. non-dyslexics

		Fractal Dimension	Recurrence Rate	Determinism	Entropy	Meanline
Dyslexics (N = 15)	Mean RT	.56*	-.70**	-.88**	-.81**	-.77**
	St. Dev. RT	.68**	-.71**	-.84**	-.76**	-.74**
	EMT	-.77**	.84**	.75**	.79**	.83**
Non- Dyslexics (N = 15)	Mean RT	.24	-.10	-.12	-.25	-.24
	St. Dev. RT	.49	.39	.33	.23	.24
	EMT	-.28	.37	.53*	.43	.41

\*\*  $p < 0.01$ , \*  $p < 0.05$  (two-tailed).

➔ 1/f noise and RQA outcomes are strongly correlated with the severity of the reading impairment