

# Radboud Summer School

## Complexity Methods for Behavioural Science

**Dr. Maarten Wijnants**

Guest Lecture: Applications of fractal analysis

Faculty of Social Sciences  
Nijmegen

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# Different faces of variability

## Variability as noise

- $X = T + E$



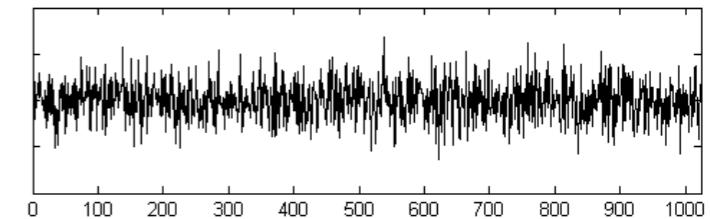
High variability



Low variability

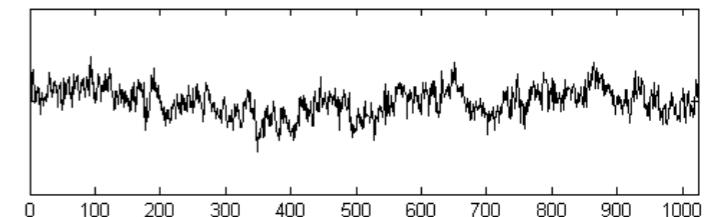
## → Amount of variability

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



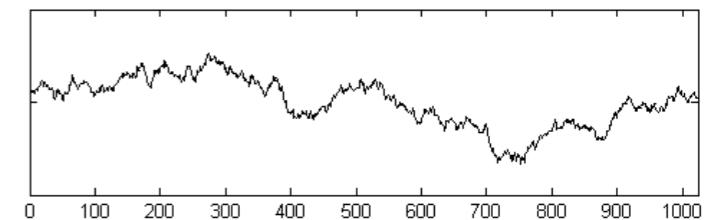
## Variability as structure

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



## → Temporal correlations

- Structured  $\leftrightarrow$  Unstructured
- e.g., ACF, roughness, entropy, DFA, SDA, spectral analysis,...

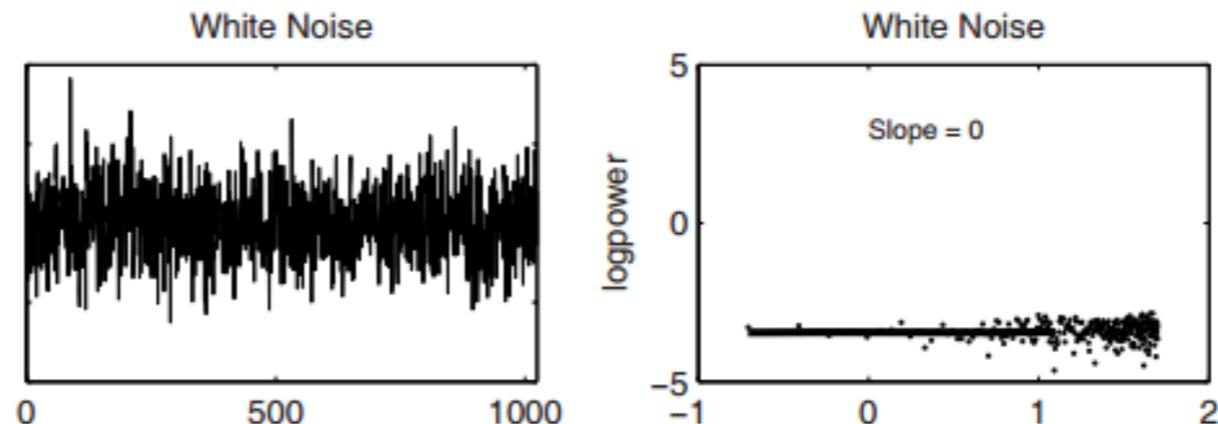


### White noise ( $1/f^0$ )

SDA: FD = 1.5

DFA:  $\alpha = 0.5$

Spectral:  $\beta = 0$

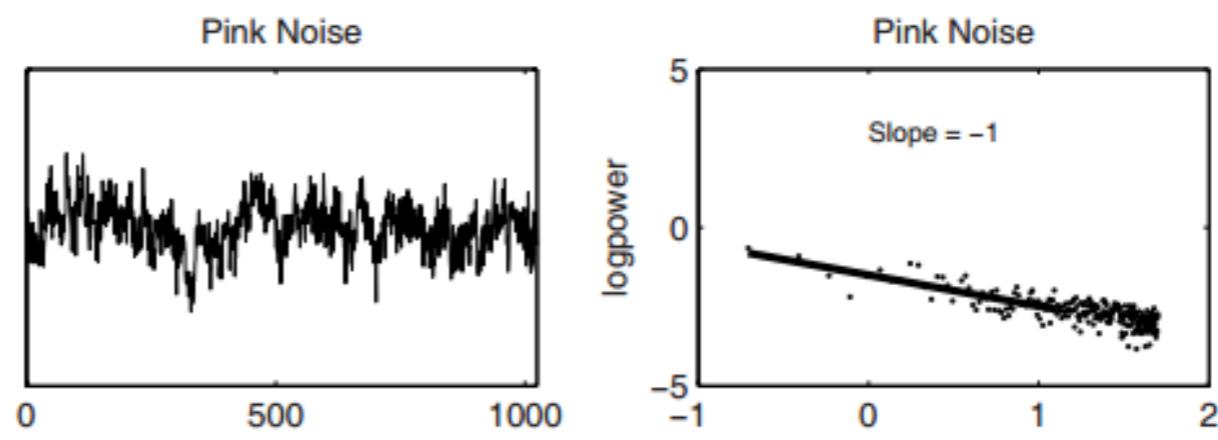


### Pink noise ( $1/f^1$ )

SDA: FD = 1.2

DFA:  $\alpha = 1$

Spectral:  $\beta = -1$



### Brownian noise ( $1/f^2$ )

SDA: FD = 1.1

DFA:  $\alpha = 1.5$

Spectral:  $\beta = -2$

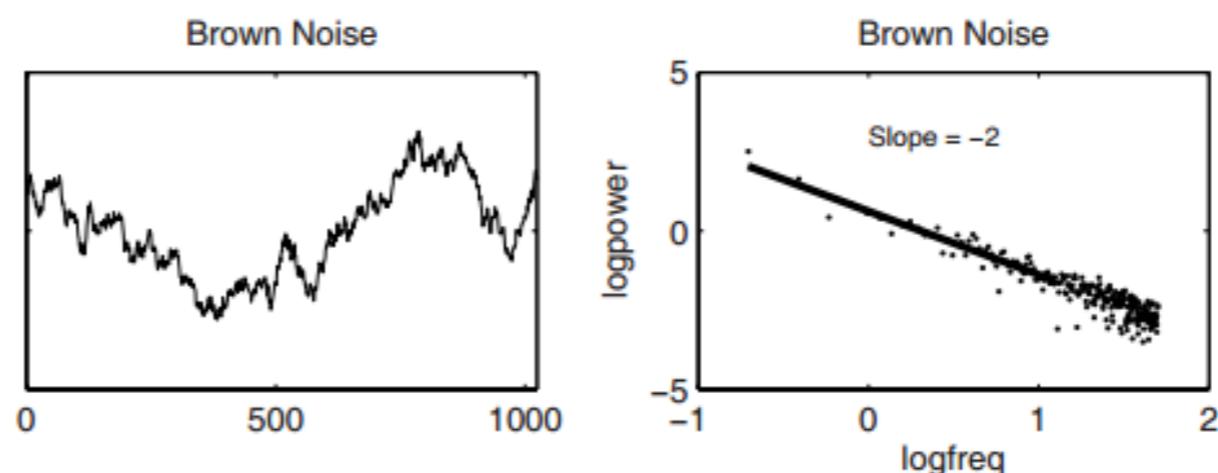
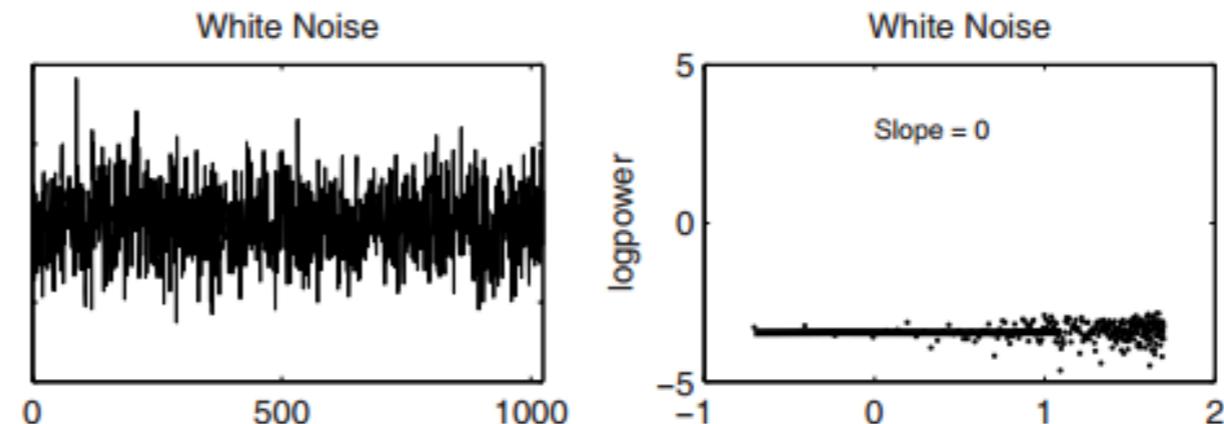


FIGURE 3 | Three different classes of temporal variability, white noise (upper left panel),  $1/f$  scaling (middle left panel), and Brownian noise (lower left panel), and their respective power spectra are shown in the respective panels at the right.

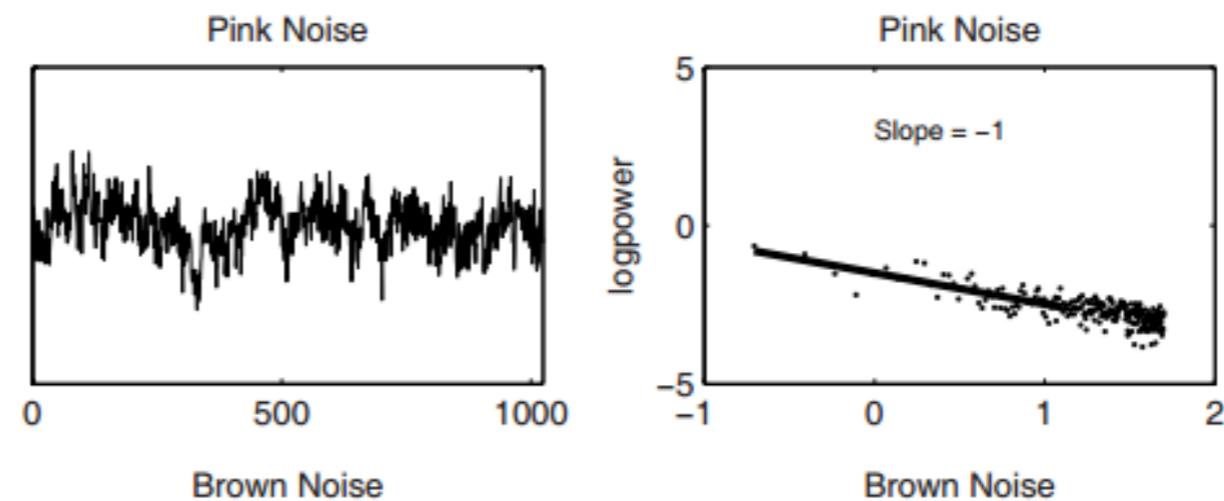
## White noise ( $1/f^0$ )

Unconstrained  
Random  
Independent behavior



## Pink noise ( $1/f^1$ )

Reduced degrees of freedom  
Statistical self-similarity  
Every data point participates  
in a larger, scale-free pattern



## Brownian noise ( $1/f^2$ )

Highly constrained  
Persistent  
Strong serial dependence

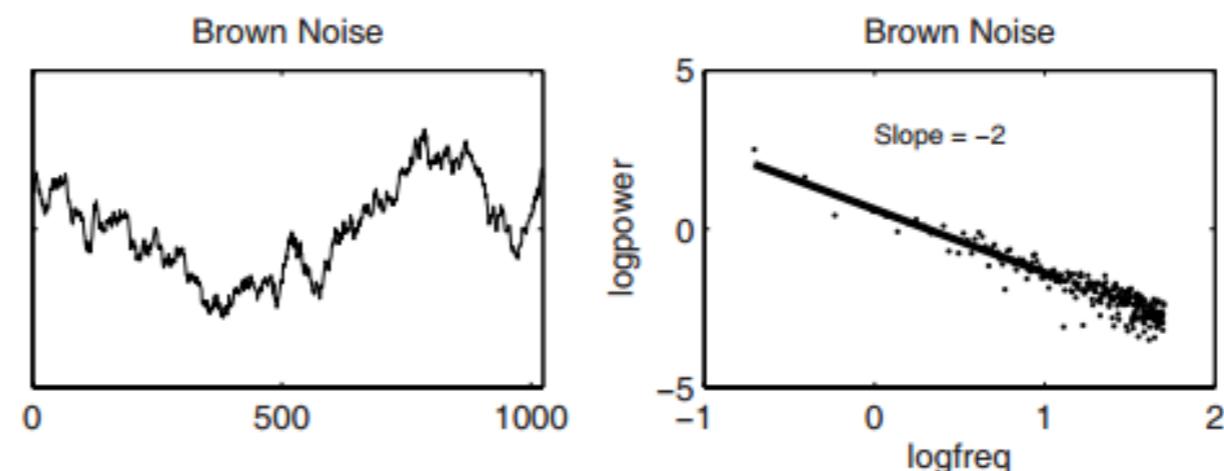
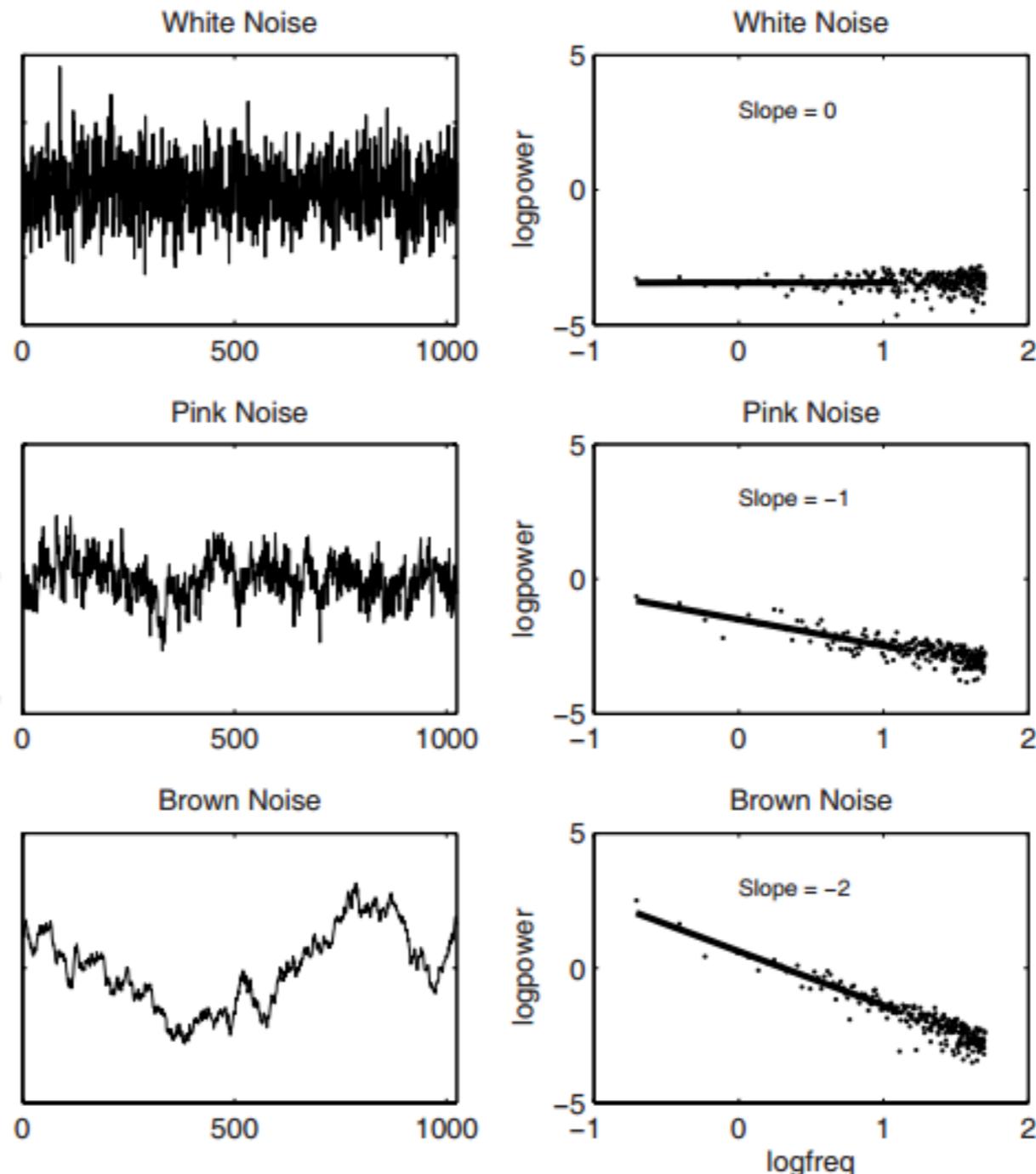
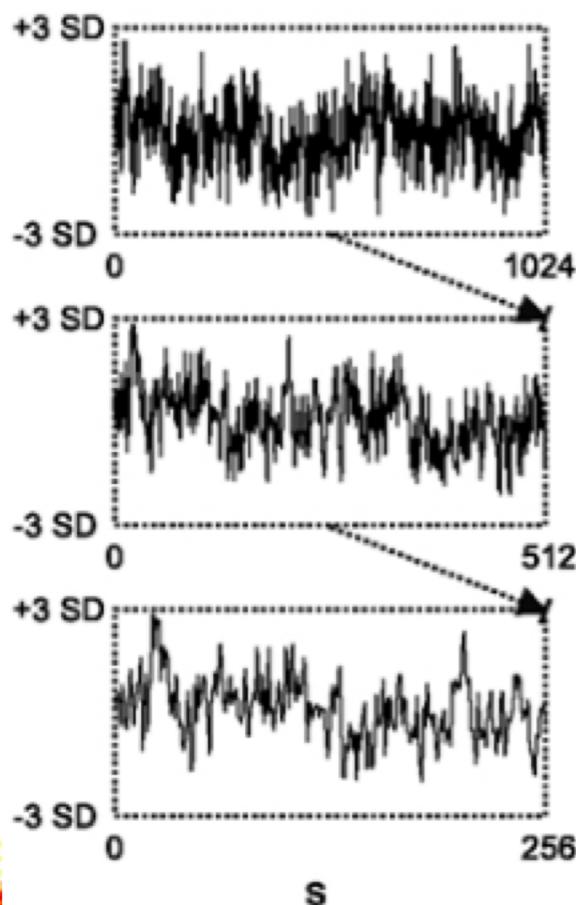
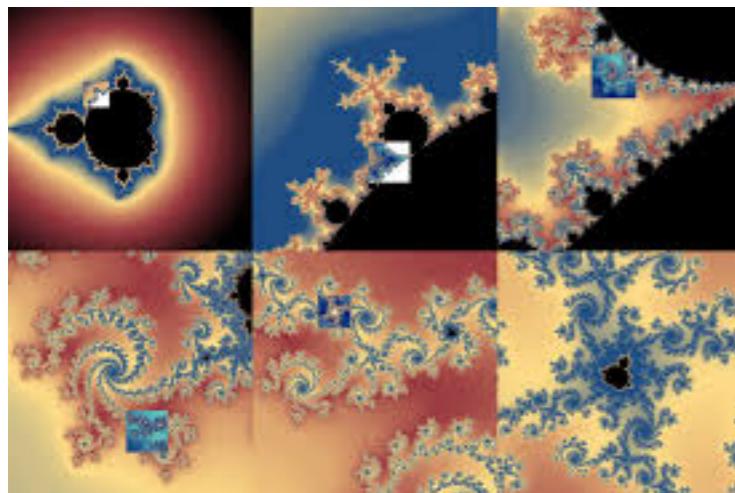


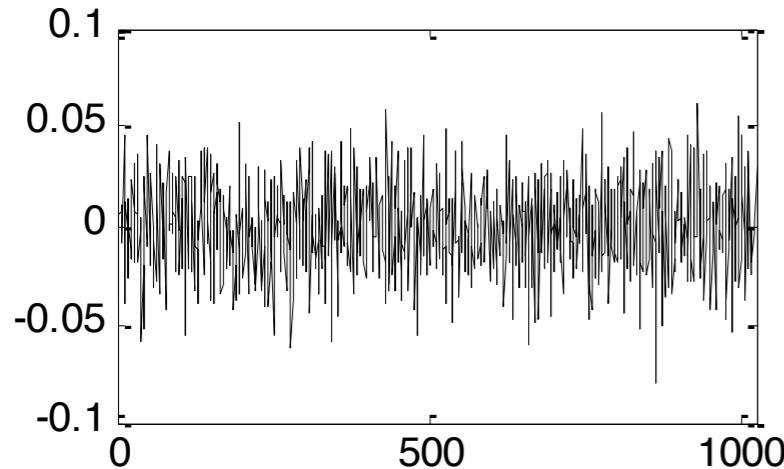
FIGURE 3 | Three different classes of temporal variability, white noise (upper left panel),  $1/f$  scaling (middle left panel), and Brownian noise (lower left panel), and their respective power spectra are shown in the respective panels at the right.

# Every data point participates in a larger, scale-free pattern

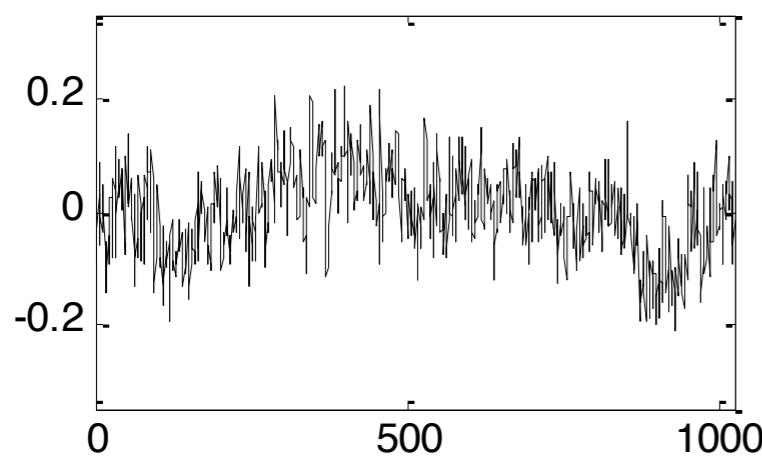


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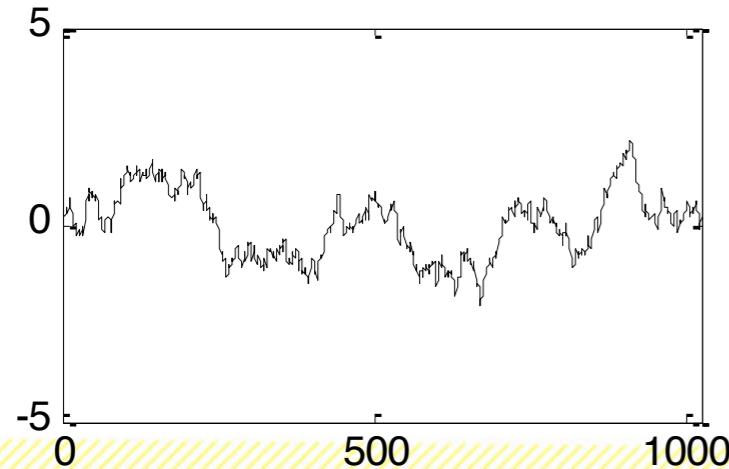
White Noise



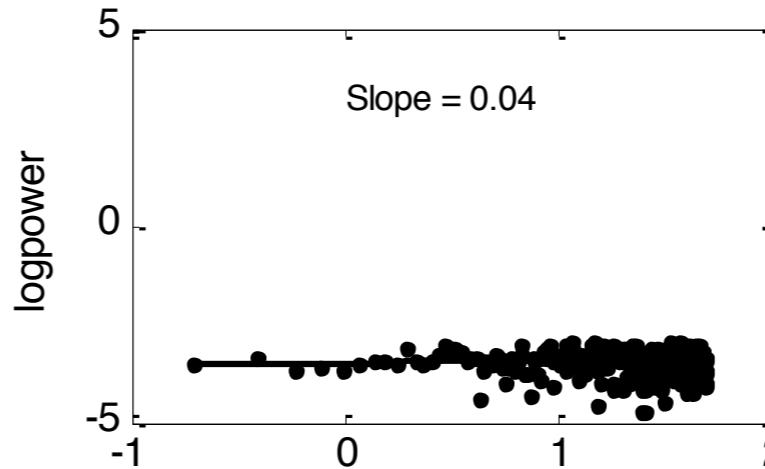
Pink Noise



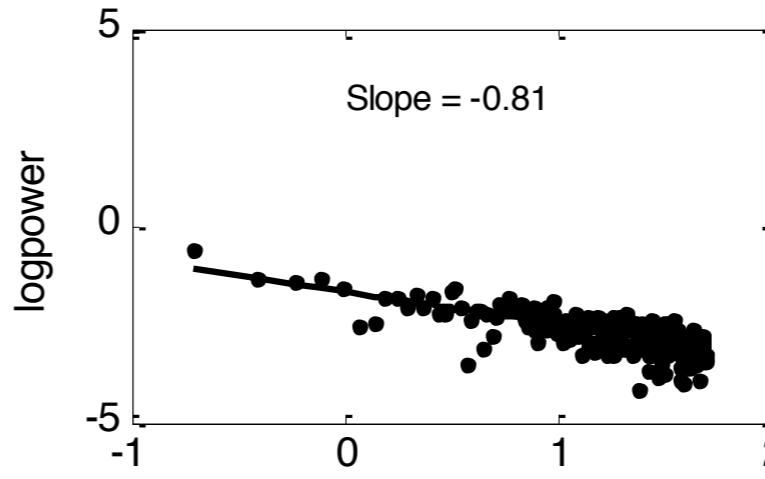
Brown Noise



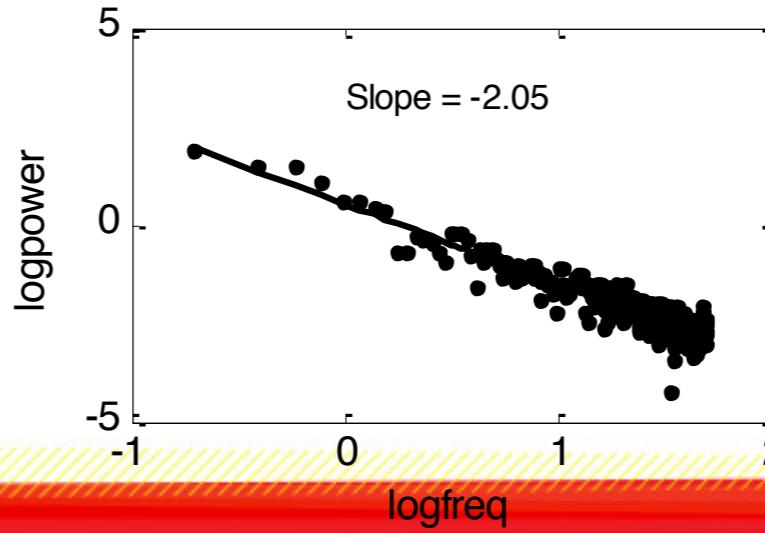
White Noise



Pink Noise



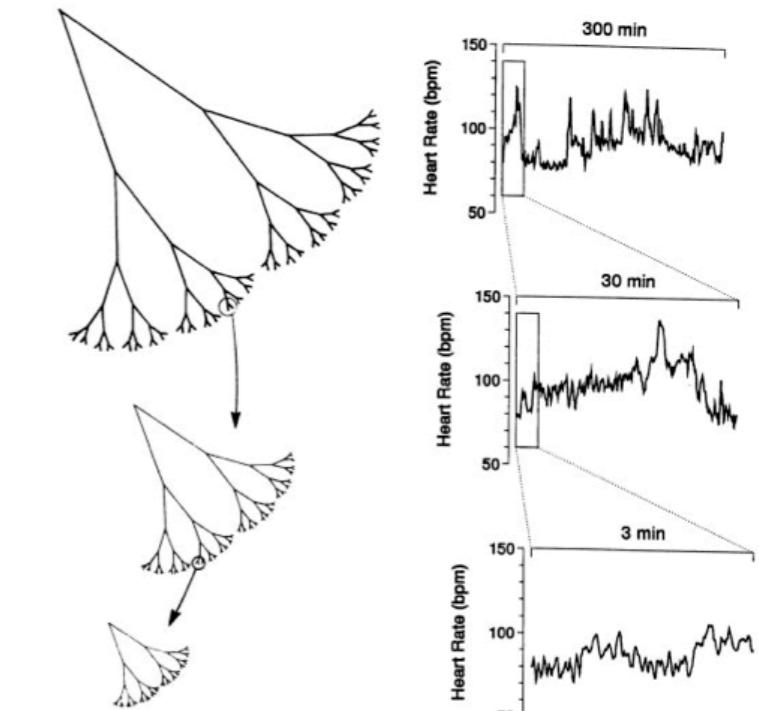
Brown Noise



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

Disorganized

Spatial Self-Similarity      Temporal Self-Similarity



Rigid & persistent

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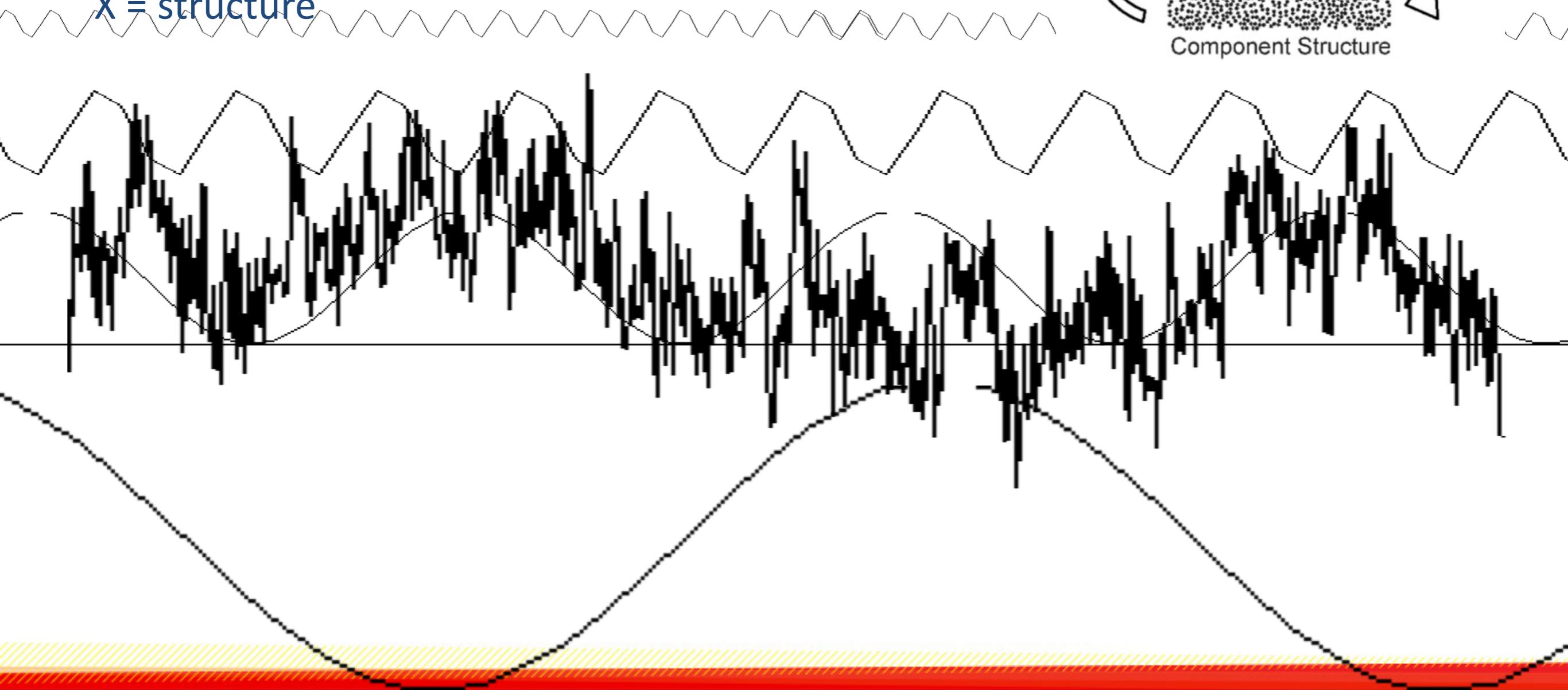
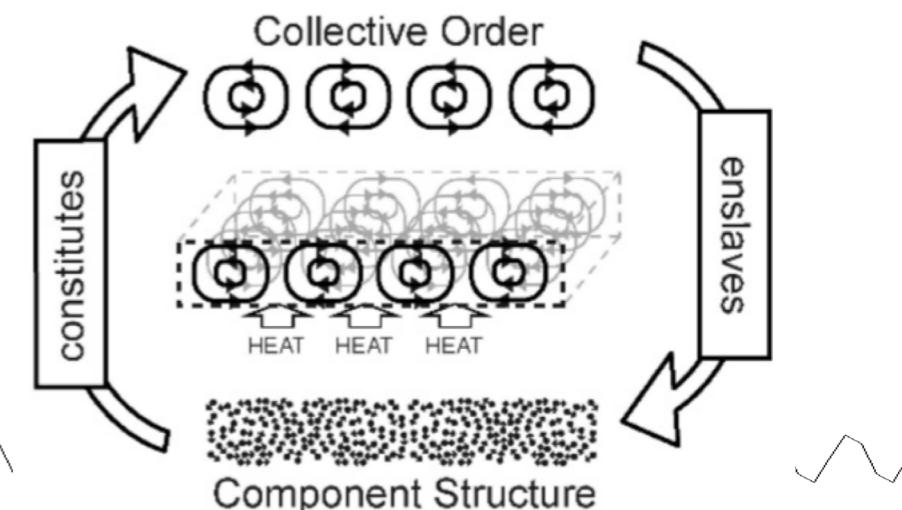


## Interaction-dominant dynamics

$1/f$  noise = fingerprint of self-organized systems

Part-whole relation

X = structure



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## Empirical timeseries (Applications)

Do empirical timeseries actually reveal 1/f noise?  
Does the value of the scaling exponent matter?

- Medicine & Physiology
- Neuroscience
- Cognition

# Fractal physiology

A healthy heart fluctuates as  $1/f$  noise

Loss of function = loss of complexity

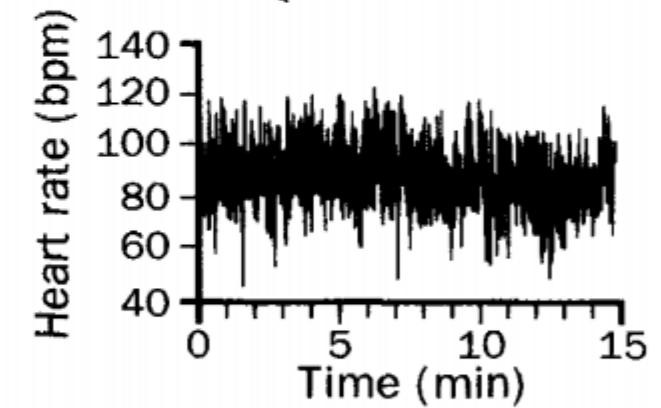
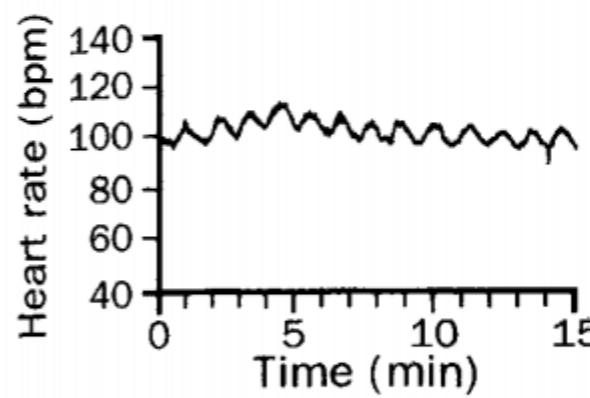
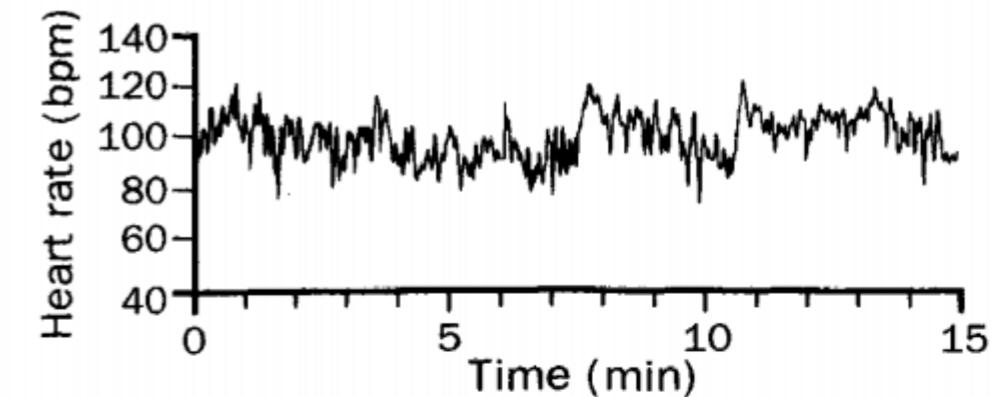


Figure 2: **Healthy dynamics (top), showing multiscale, long-range order; pathological breakdown of fractal dynamics, leading to single-scale (bottom left) or uncorrelated randomness (bottom right)**

Top heart-rate time-series is from a healthy individual; bottom left is from patient with heart failure, and bottom right from patient with atrial fibrillation.

# Fractal physiology

A healthy heart fluctuates as  $1/f$  noise

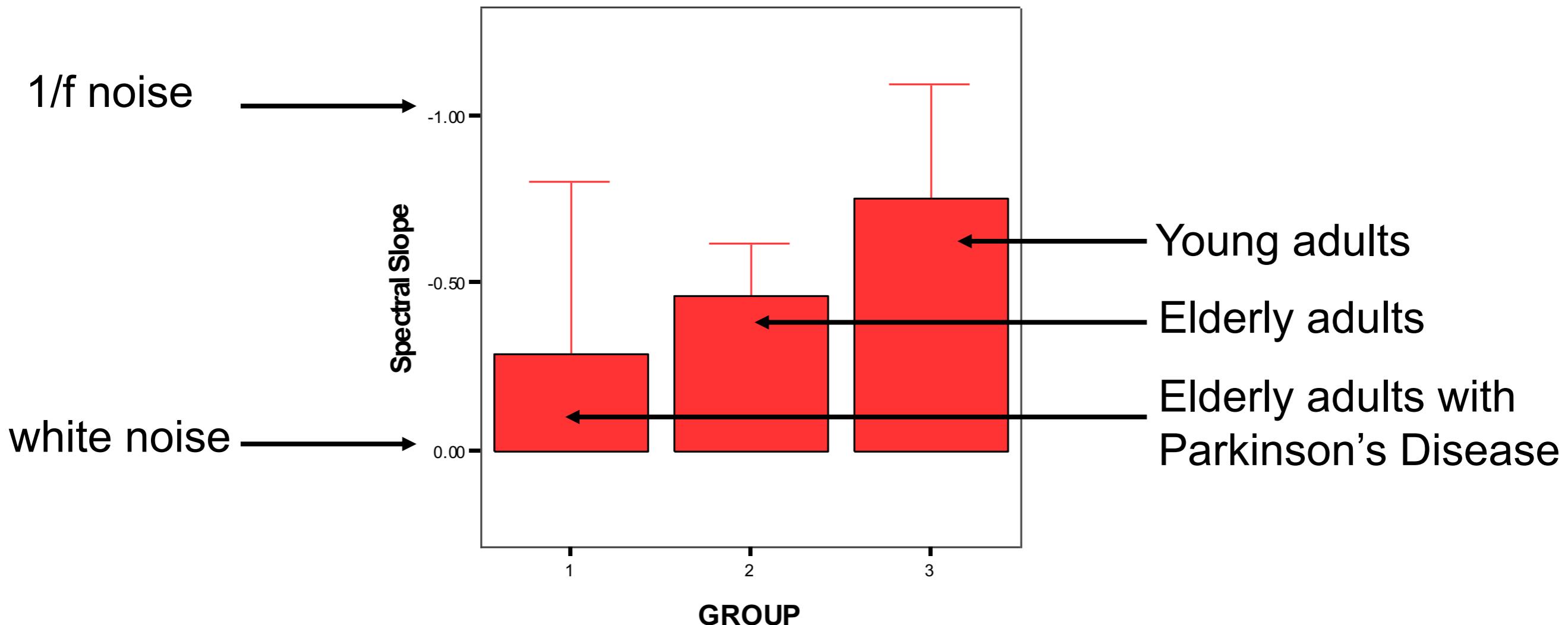
Deviations from  $1/f$  noise correlate with mortality risk (Mäkikallio et al., 2001)

- Congestive heart failure
- Ventricular arrhythmia (Goldberger, 1997; Peng et al., 1995)

More subtle deviations from  $1/f$  noise

- Aging (Goldberger, 2002)
- Obese children (Vanderlei, Pastre, Júnior, & de Godoy, 2010)
- Adults with down syndrome (Mendonca, Pereira, & Fernhall, 2011)

## Gait intervals



(Hausdorff, 2007)

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# Fractal physiology

A healthy respiratory system emits  $1/f$  noise

Breathing rhythm

- $1/f$  noise → white noise
  - Aging (Peng et al., 2002; West, 2006)
- White noise →  $1/f$  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

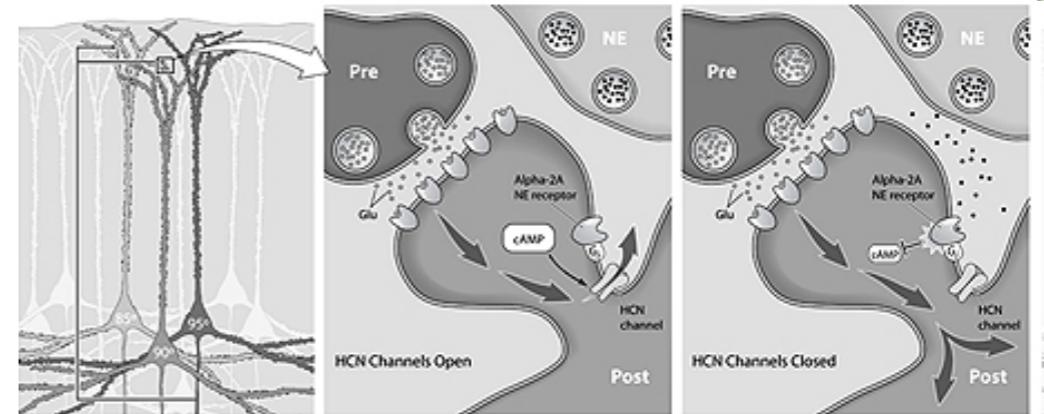
## Colon pressure

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

# Fractal neuroscience

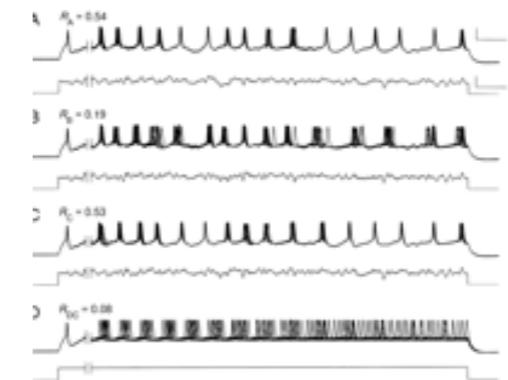
## Ion Channels Opening and Closing Times

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



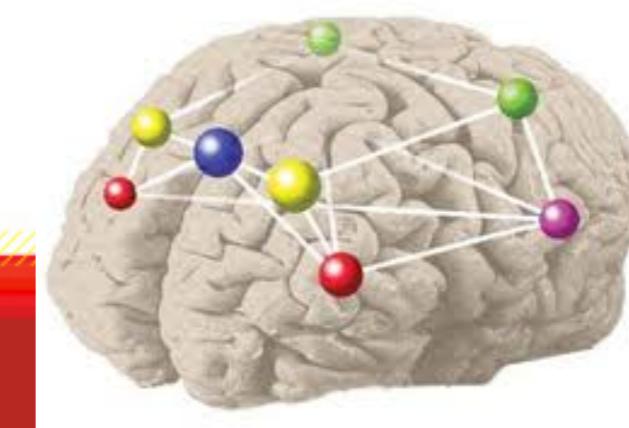
## Neural Spike Intervals

- (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)



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# Fractal neuroscience

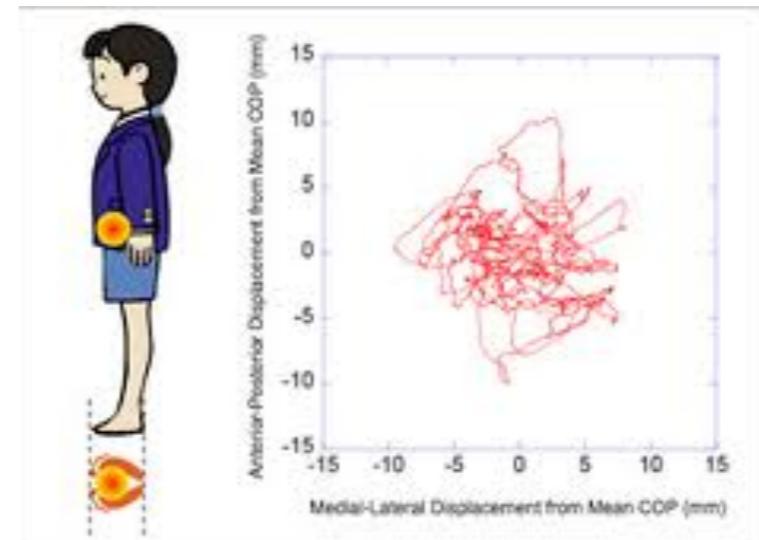
## Deviations from $1/f$ noise in EEG

- 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)
- ...

# Motor Control

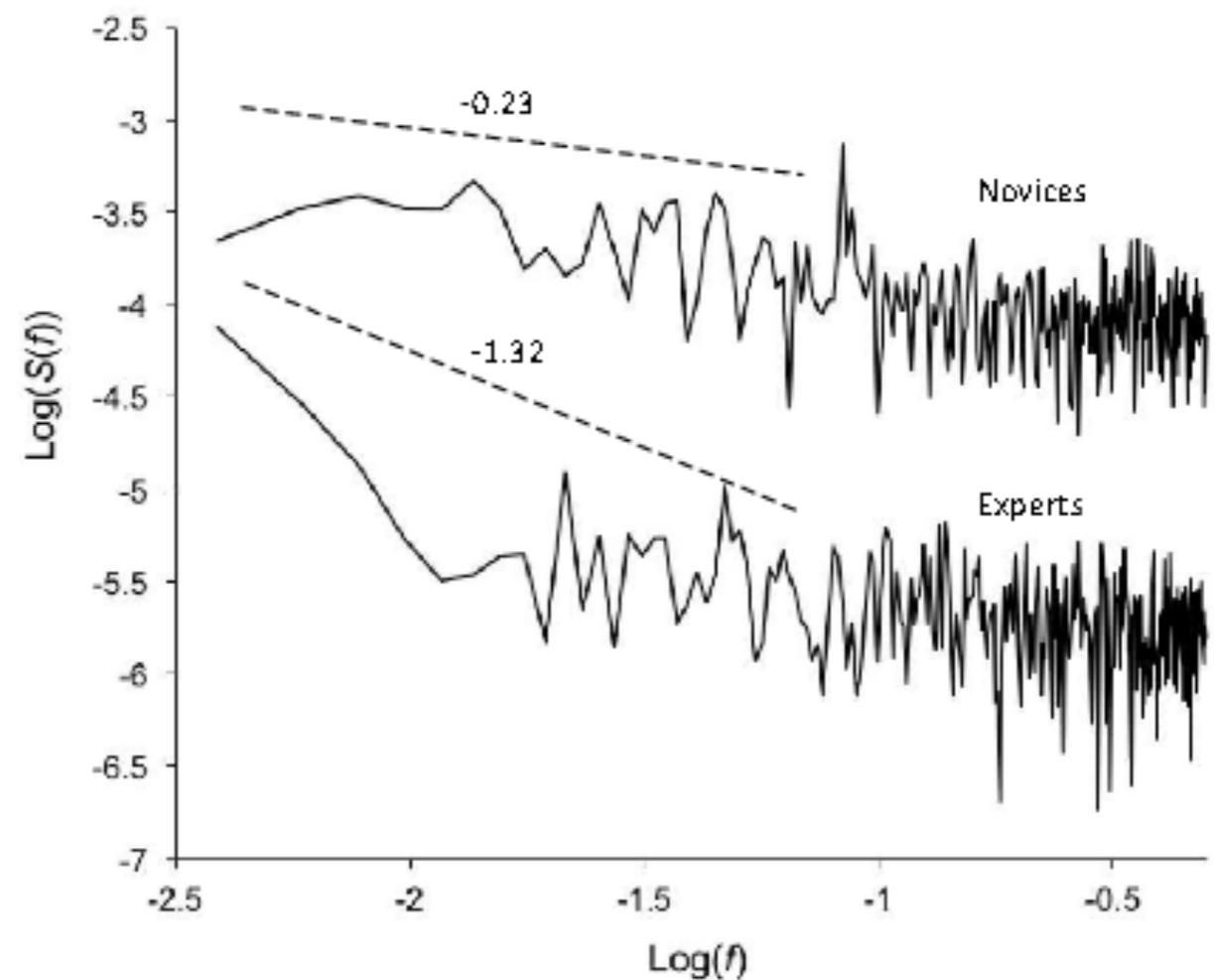
## Posture

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



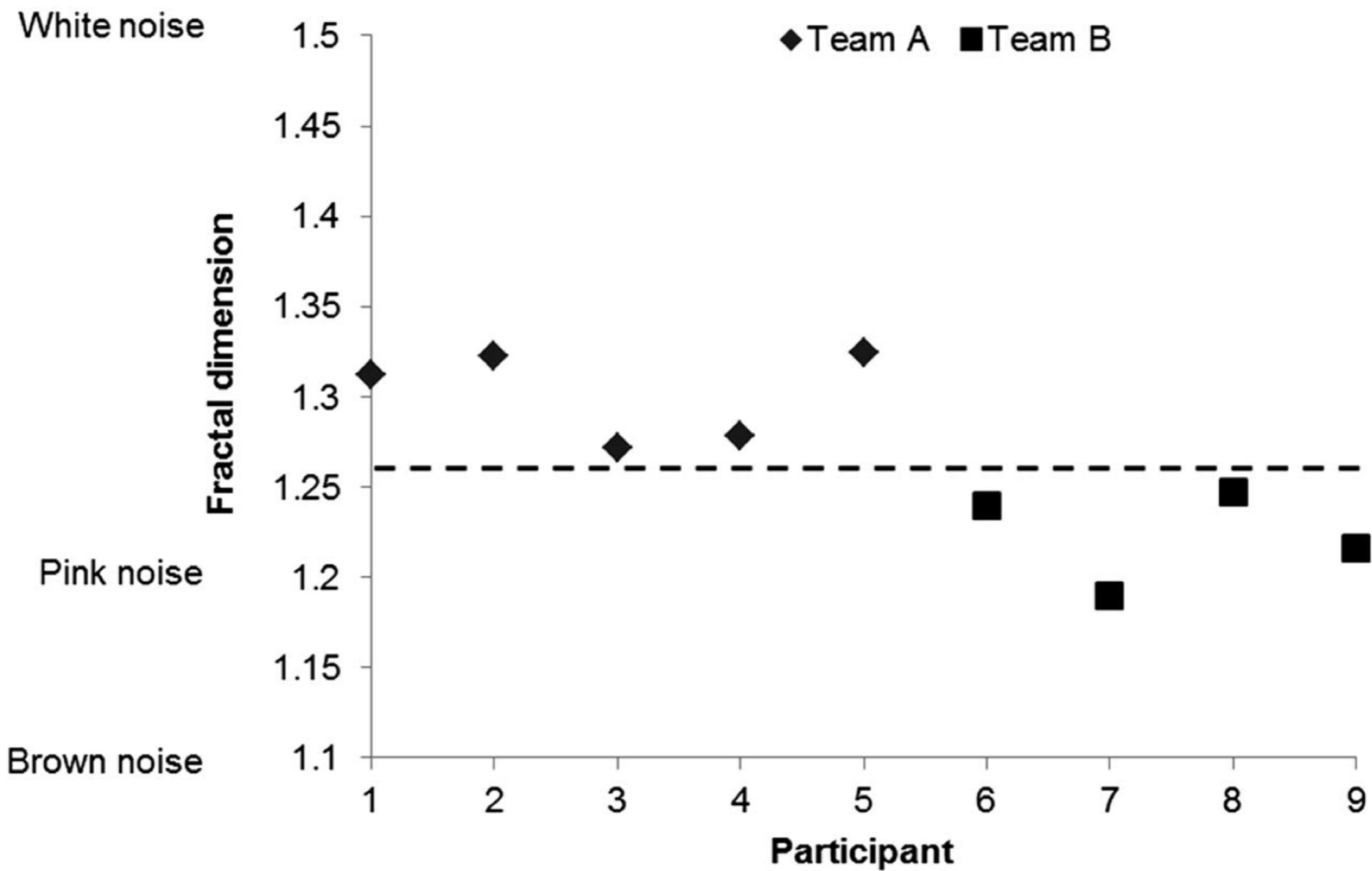
Correlation fractal dimension and falling risk in elderly  
( $r = .78$ ,  $p < .0005$ ; Hausdorff, 2007)

# Skiing



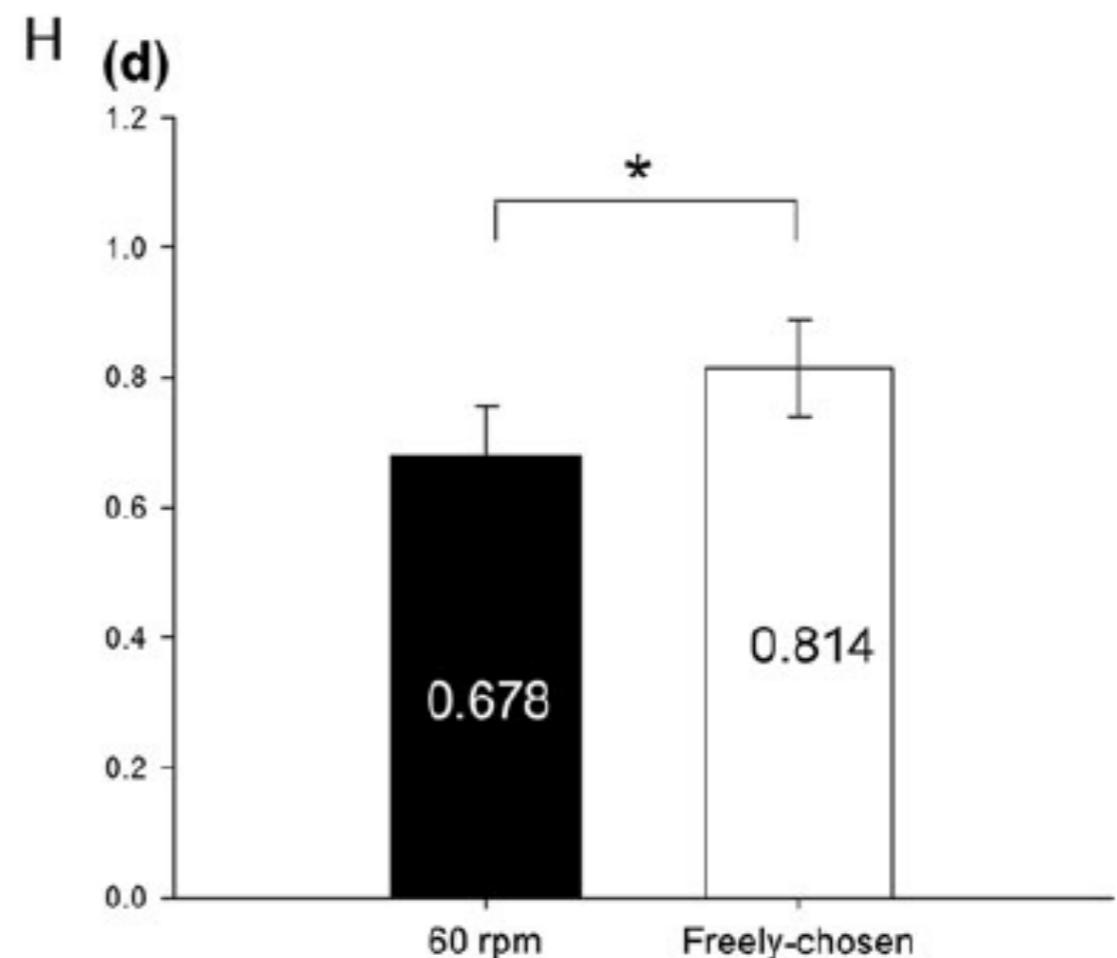
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# Rowing



**Figure 4** — Fractal dimensions of lower-skilled participants (Team A) and high-skilled participants (Team B). The dashed line separates the rowers of the two teams.

# Cycling



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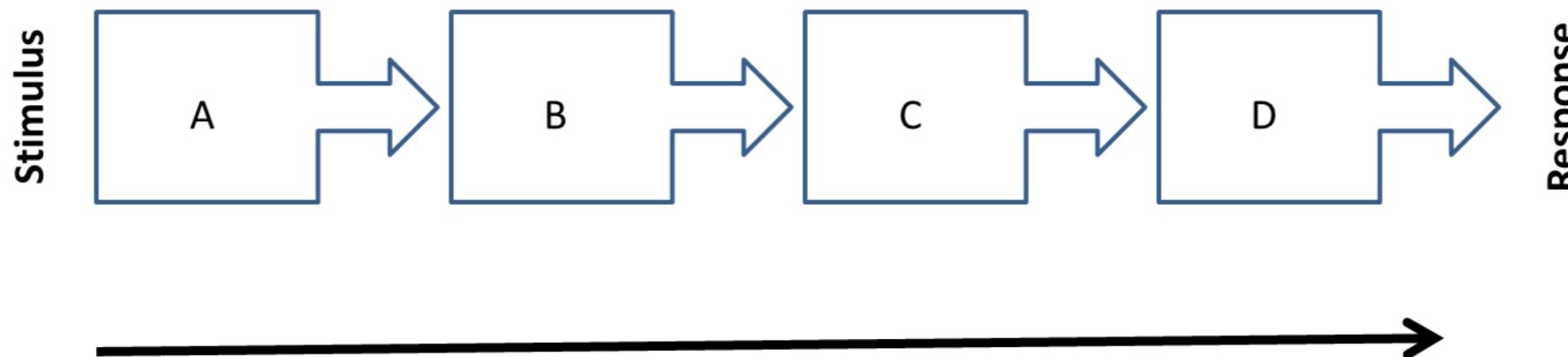
# **What can we learn from 1/f noise in cognitive performances?**

Maarten Wijnants

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$$A + B + C + D = RT$$

E.g., 84ms + 46ms + 128ms + 304ms = 562ms

## Component-dominant dynamics ???

### 1/f Noise in Human Cognition

D. L. Gilden,\* T. Thornton, M. W. Mallon

When a person attempts to produce from memory a given spatial or temporal interval, there is inevitably some error associated with the estimate. The time course of this error was measured in a series of experiments where subjects repeatedly attempted to replicate given target intervals. Sequences of the errors in both spatial and temporal replications were found to fluctuate as 1/f noises. 1/f noise is encountered in a wide variety of physical systems and is theorized to be a characteristic signature of complexity.

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

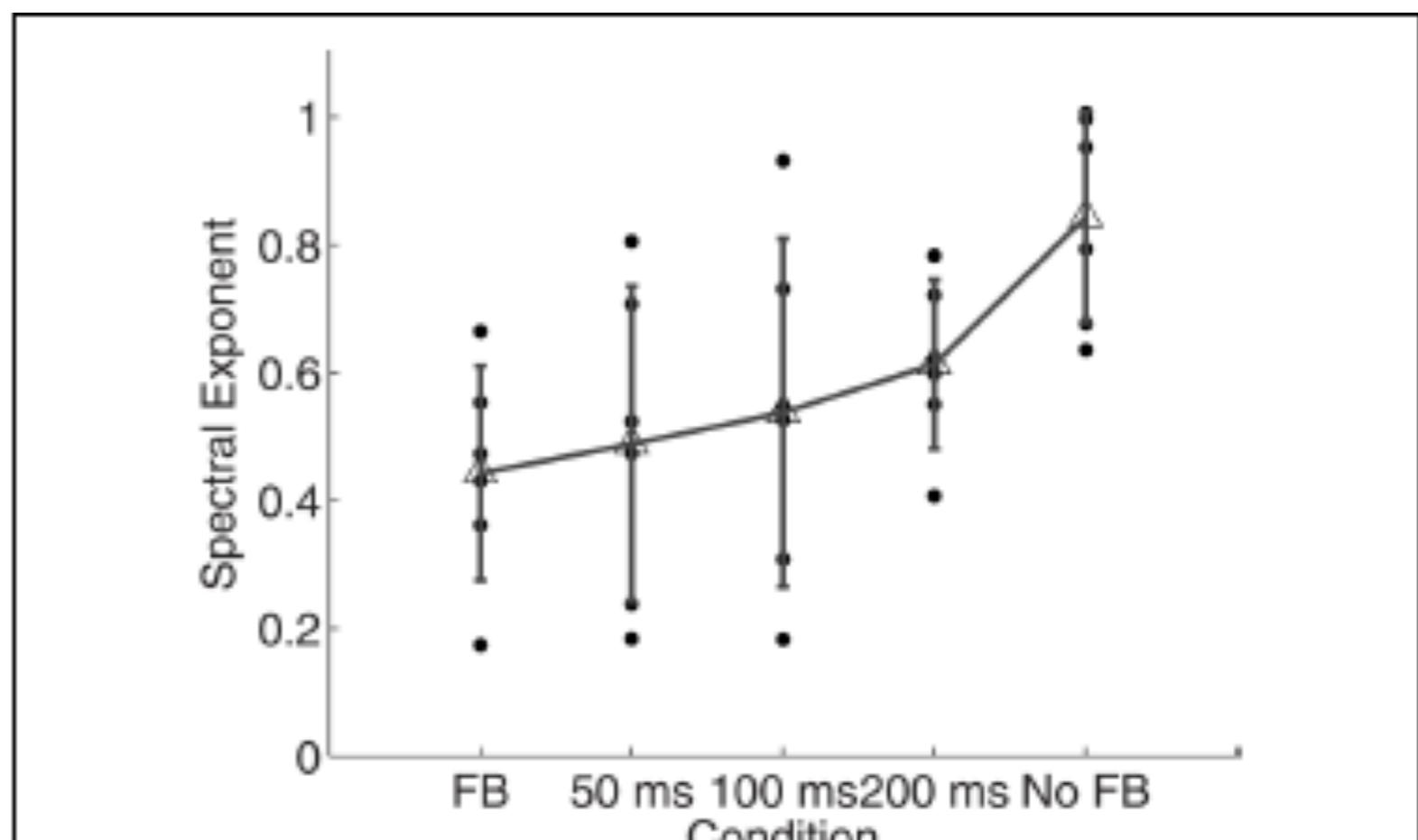
## Proof of the pudding

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

## Task constraints

### Temporal estimation

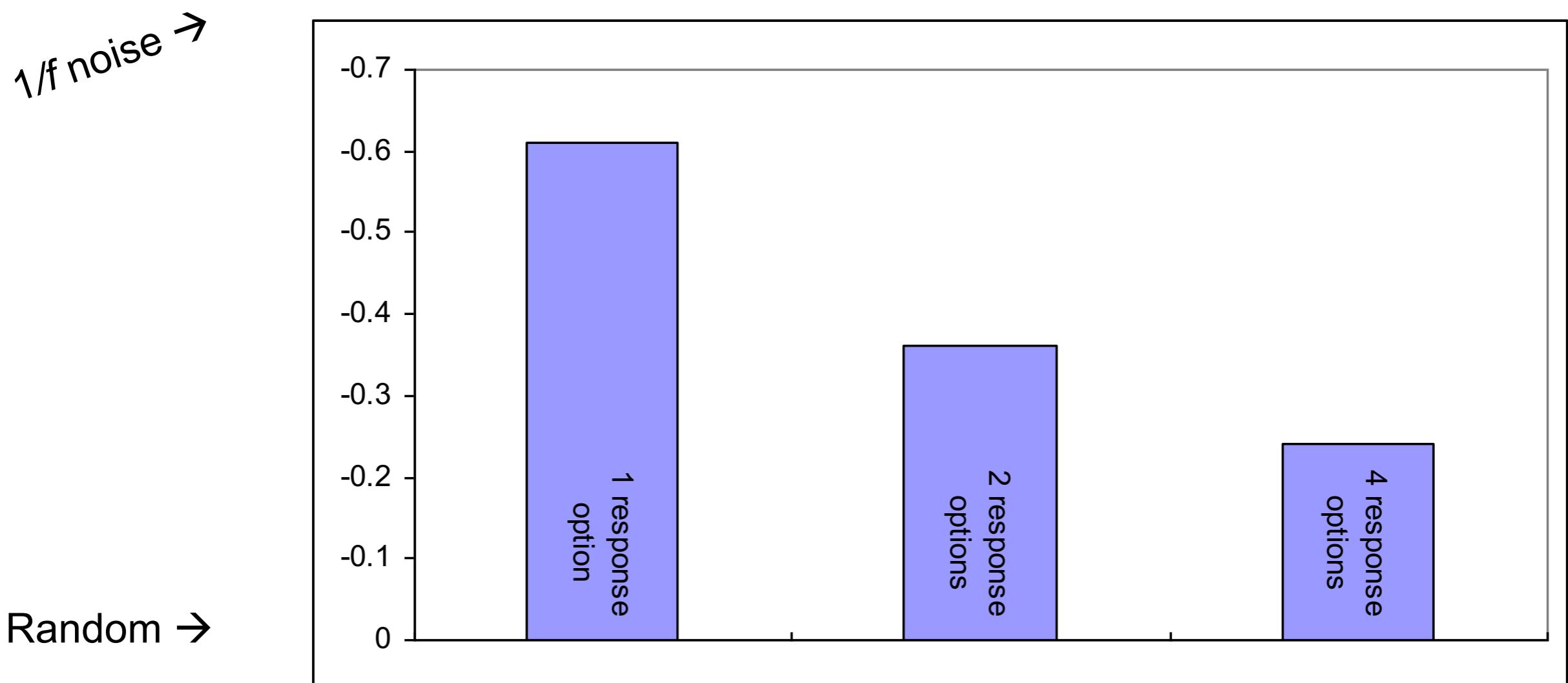
External perturbations add extraneous random variation to the measured performances: Accuracy feedback



**FIGURE 3 | Spectral exponents of the time estimates.** Spectral exponents  $\alpha$  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.

## Choice reaction task

External perturbations add extraneous random variation to the measured performances: Number of response options

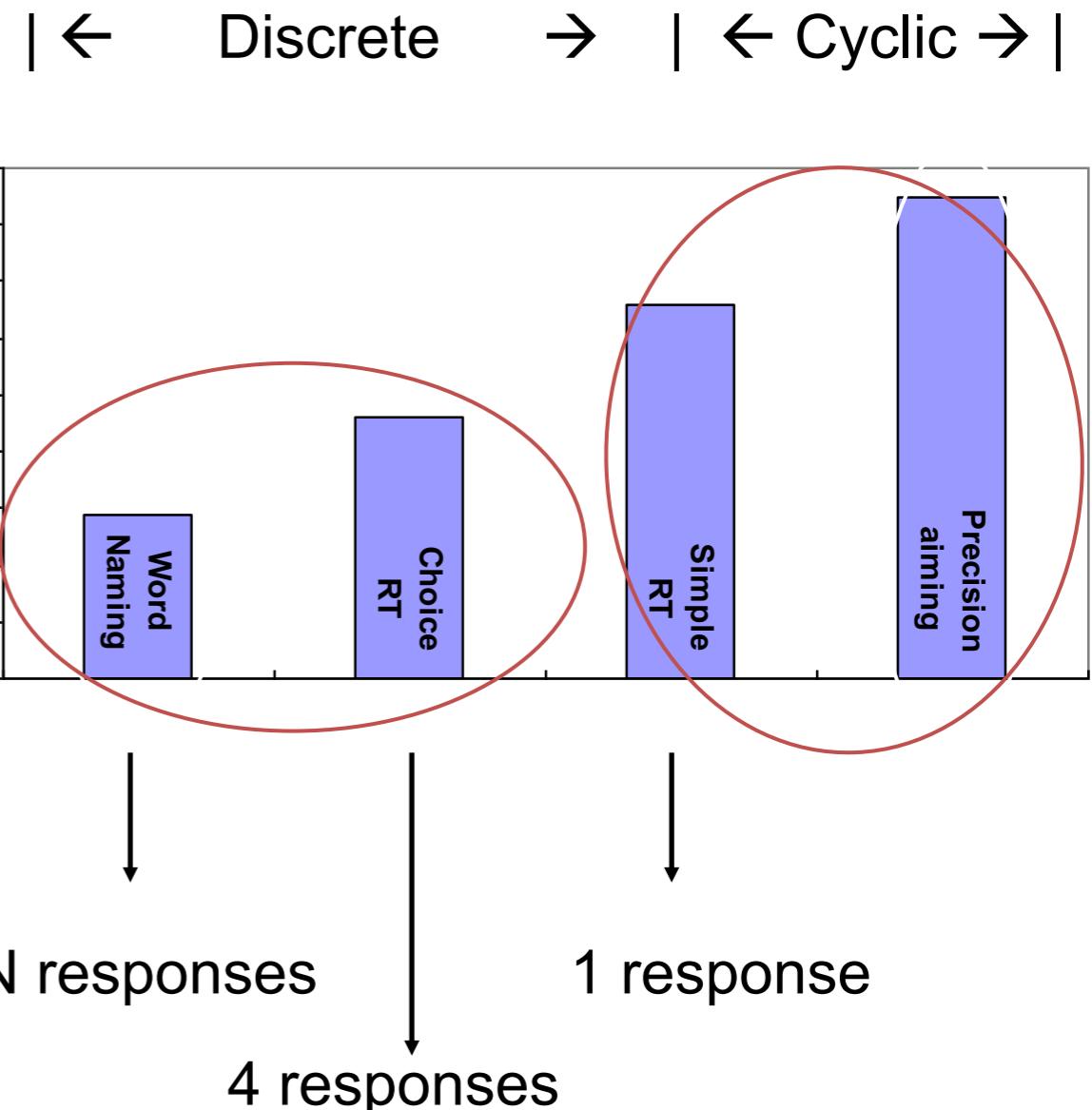


Choice Reaction Task

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# Task constraints: across tasks

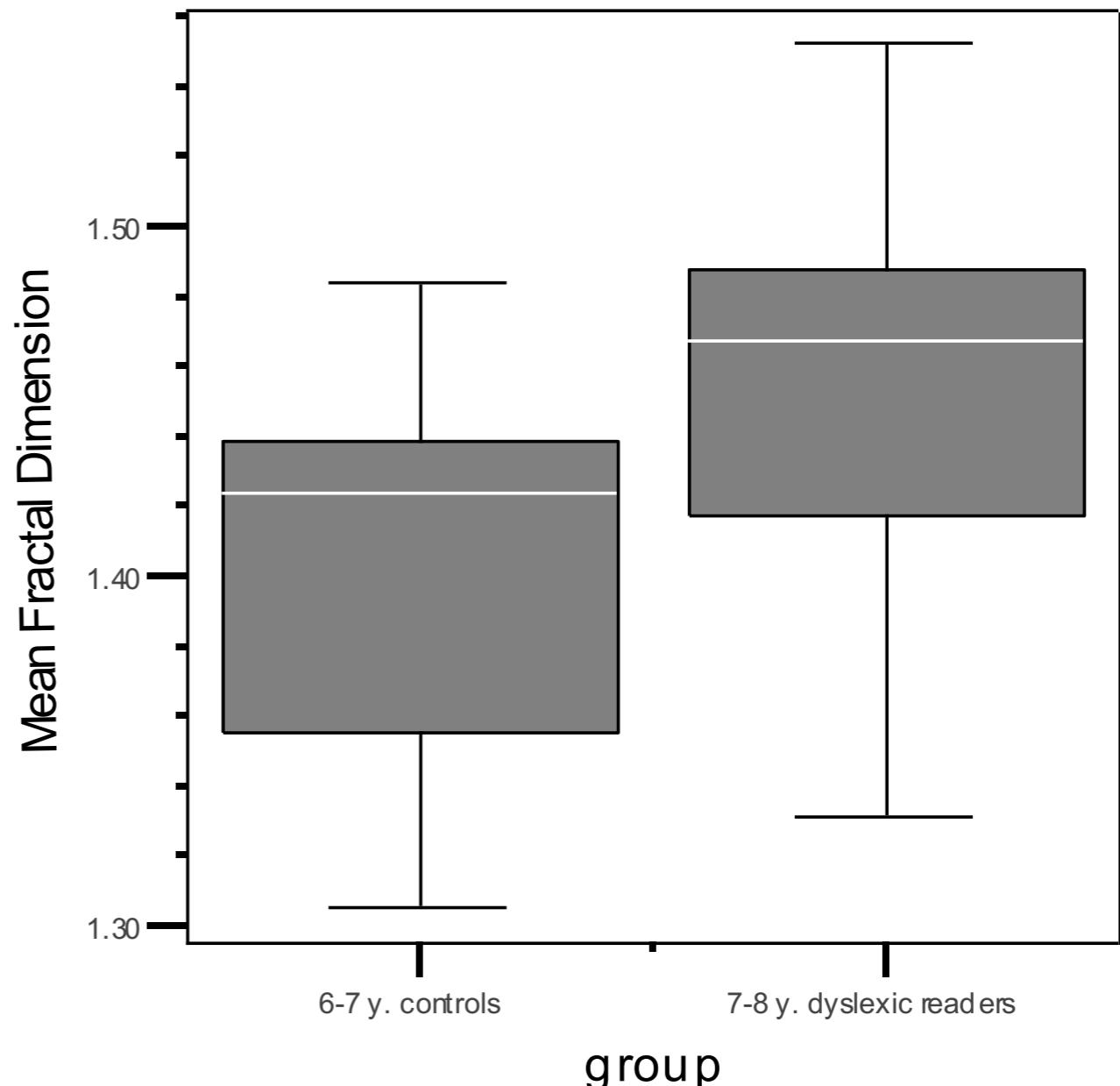
- **Simple RT, Precision aiming:**
  - Each trial is identical: same SIGNAL to respond and same RESPONSE
  - EXTERNAL sources of variation in Response Time are minimized
- → Variation must largely reflect INTERNAL sources
- **Choice RT, Word-naming**
  - Experimental trials differ: A different SIGNAL to respond and a different RESPONSE
  - EXTERNAL sources of variation in Response Time are introduced to the measured values
- → Variation must reflect INTERNAL sources to a lesser extent



Data from: Van Orden, Holden, & Turvey, 2003;  
Kello, Beltz, Van Orden, & Turvey, 2007; Wijnants et al., 2009

## Word-naming

- 7-8 year old Dyslexic Readers
- 6-7 year old Controls
- 1 Block of 560 Word Stimuli
- → Dyslexic Readers Show Reduced 1/f Noise in RTs



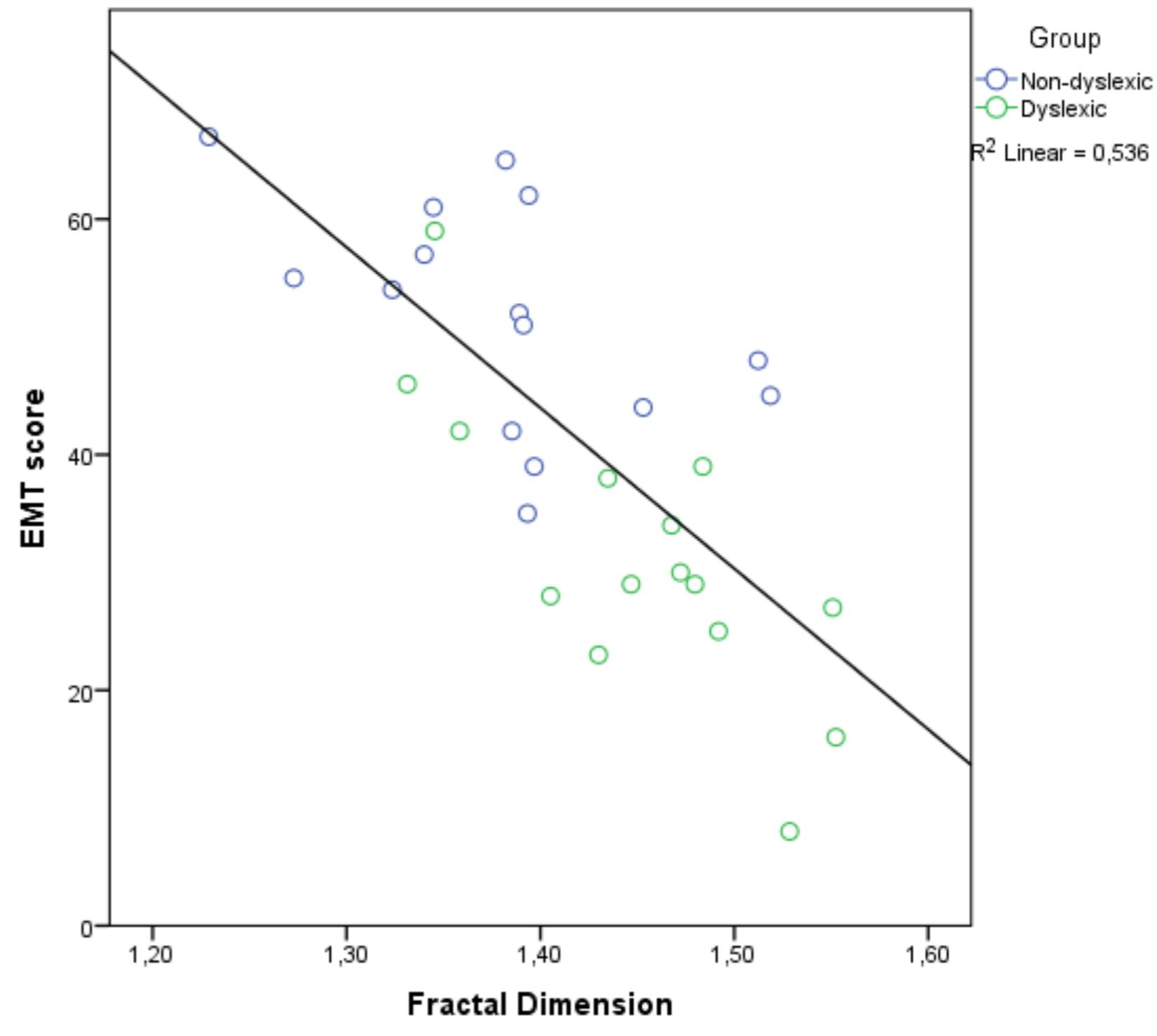
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## Word-naming

- Oral reading fluency is regarded as the sole best indicator of reading problems (Fuchs, Fuchs, Hosp, & Jenkins, 2001)
- $1/f$  noise correlates strongly with reading fluency



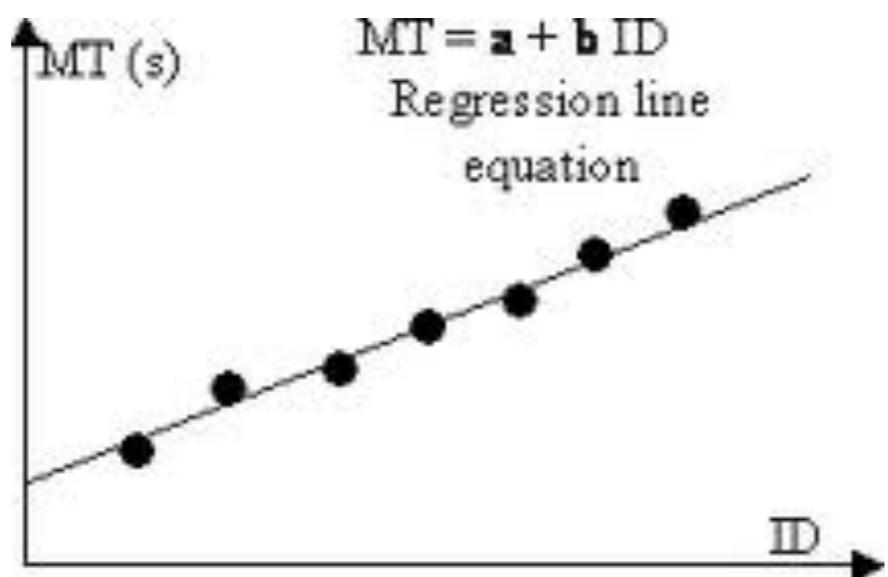
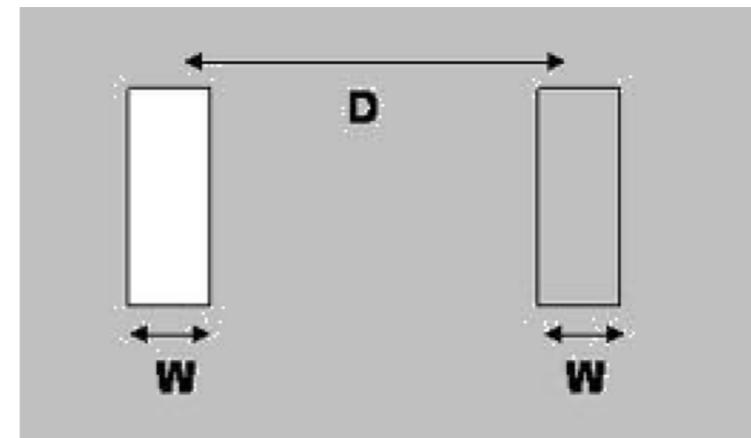
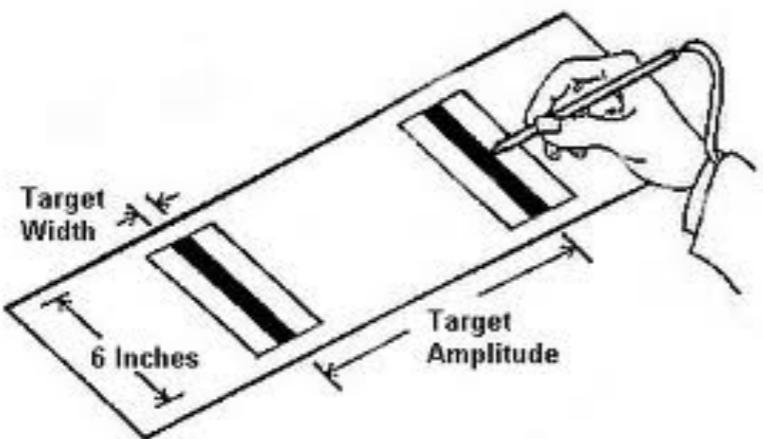
## Within-Group Correlations: dyslexics vs. non-dyslexics

		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

# Fitts task: Motor learning



A diagram illustrating the Fitts' Law equation. It shows 'Time' (T) on the left and 'Distance' (D) on the right. Between them is the equation  $T = a + b \log_2( 2 \frac{D}{W} )$ . Arrows point from 'Coefficients' to 'a' and 'b', and from 'Width' to 'W'.

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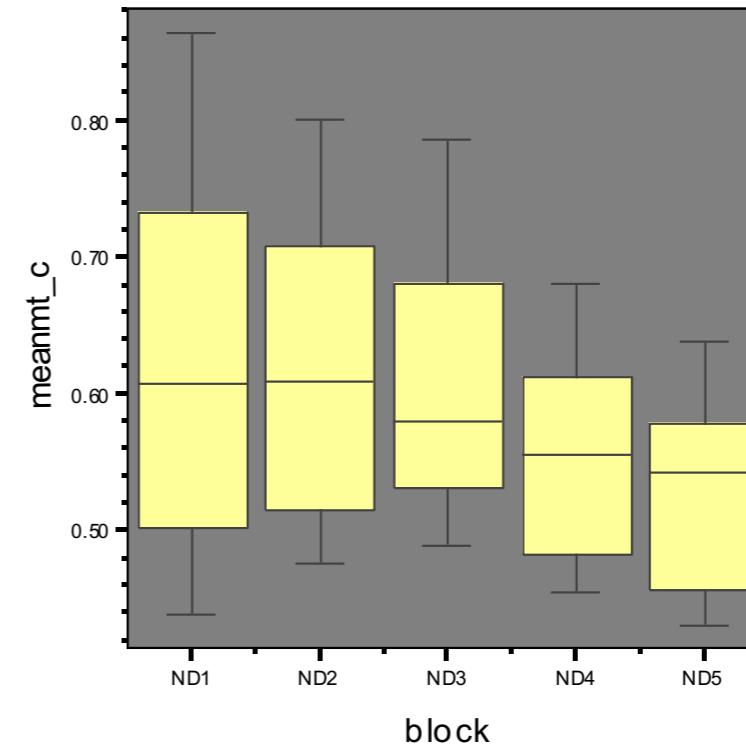
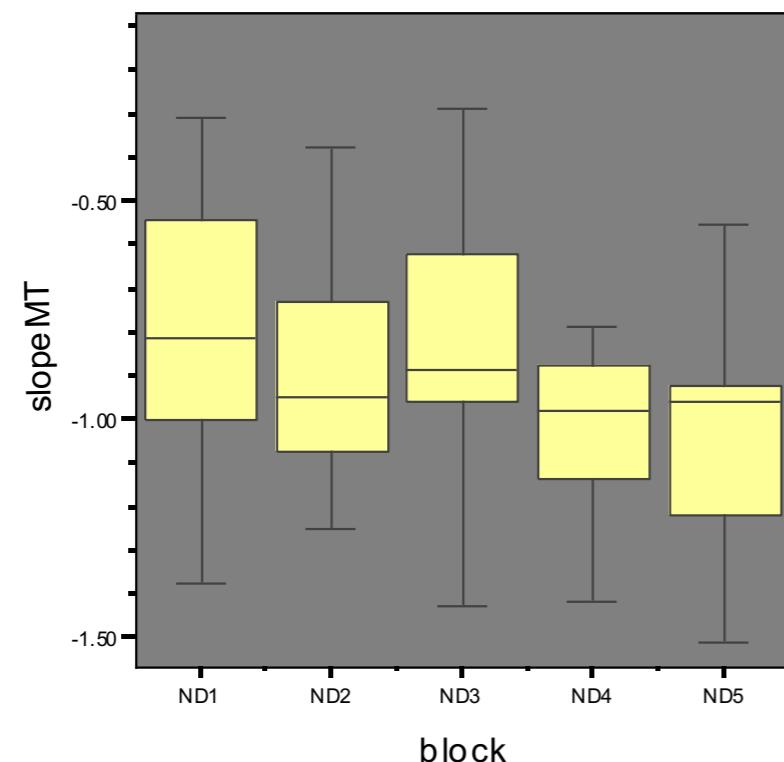


# Motor Learning

$W = 0.8 \text{ cm}$



- 5 blocks x 1100 trials
- Non-dominant hand



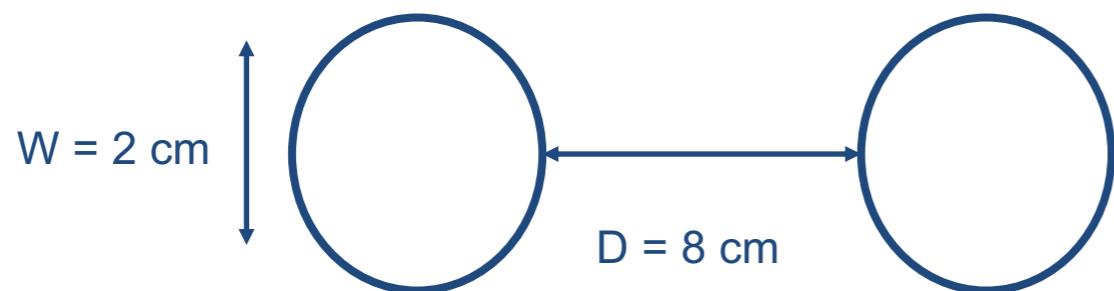
(Wijnants, Bosman, Hasselman, Cox, & Van Orden, 2009)

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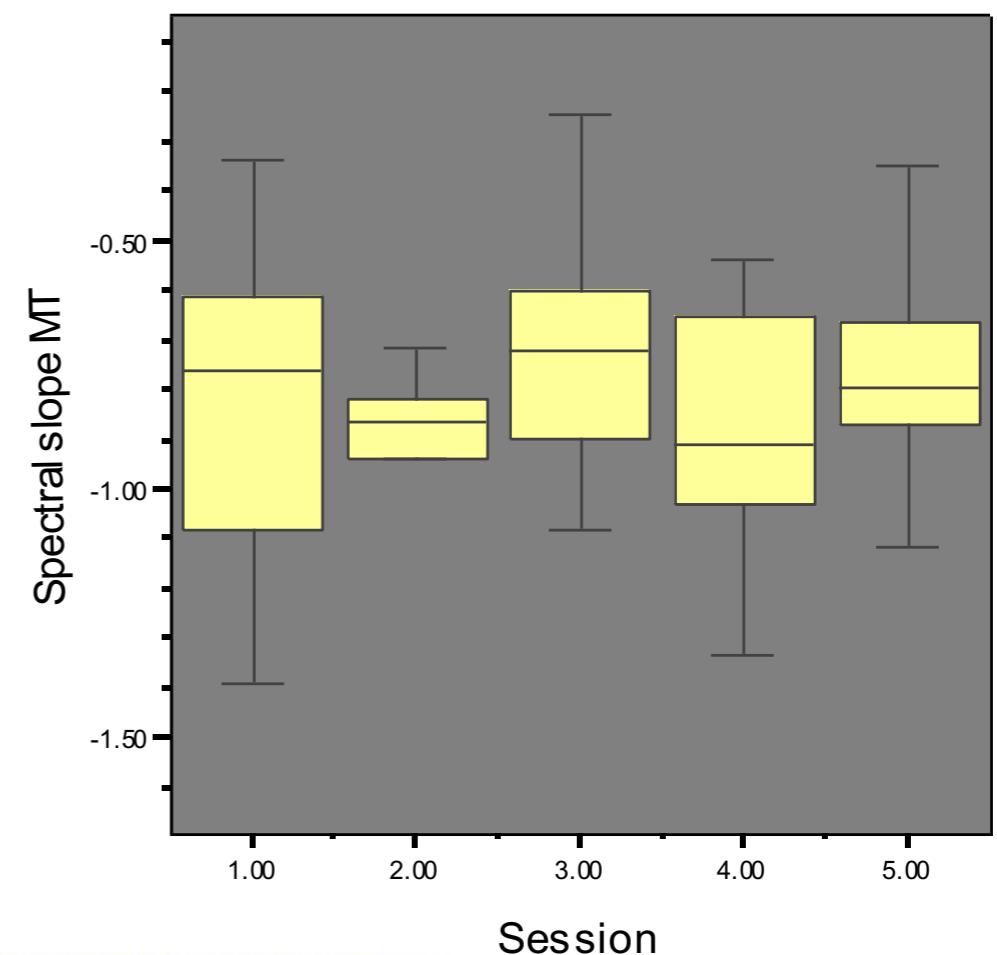
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## No Motor Learning



- 5 blocks x 1100 trials
- Non-dominant hand



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# *Speed-Accuracy Trade Off*

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- Not very accurate
  - Speed vs. Accuracy:  $r = -.77^{**}$
- Recruits slow-timescale dynamics of the fractal pattern to the measurement of speed (Movement Time)
  - $r = .62^*$
  - Faster participants show more 1/f scaling in their MT dynamics
- Does not recruit slow-timescale dynamics of the fractal pattern to the measurement of accuracy (Line Length)
  - $r = -.58^*$
  - Faster participants show reduced 1/f scaling in their LL dynamics



- Not very fast
  - Speed vs. Accuracy:  $r = -.77^{**}$
- Recruits slow-timescale dynamics of the fractal pattern to the measurement of accuracy (Line Length)
  - $r = .45^*$
  - More accurate participants show more 1/f scaling in their LL dynamics
- Does not recruit slow-timescale dynamics of the fractal pattern to the measurement of speed (Movement Time)
  - $r = -.71^*$
  - Accurate participants show reduced 1/f scaling in their MT dynamics



		<b>Harmonicity</b>	<b>Movement time</b>	<b>Accuracy</b>
Difficult condition	FD MT	-0.61 **	0.52 *	0.70 **
	SampEn MT	-0.66 **	0.45 *	0.74 **
	FD MA	0.50 *	-0.45 *	-0.48 *
	SampEn MA	0.75 **	-0.64 **	-0.74 **
Easy condition	FD MT	-0.13	0.00	0.15
	SampEn MT	-0.12	0.03	0.05
	FD MA	-0.31	0.33	0.30
	SampEn MA	0.03	-0.08	0.33

\*\* $p < 0.01$ , \* $p < 0.05$ , one-tailed.

## Conclusion

- The coordination of perception and action emerges in a similar way as the coordination of neural and bodily processes
- Many diverse systems perform optimally when interacting across many scales
- Conjecture: These fractal fluctuations indicate healthy, adaptive and stable, well-coordinated systems