

Influences of the Circadian System on Fractal Patterns of Heart Rate Fluctuations

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Outline

- Concept of *fractal*
- Fractals in physiology (heart rate and motor activity fluctuations)
- Relevance of the circadian system to fractal physiological controls
 - Circadian rhythms of heart rate fractal properties in humans and rats
 - Effects of lesioning the master clock of the circadian system
 - Conceptual neurobiological model of fractal controls

What is fractal?

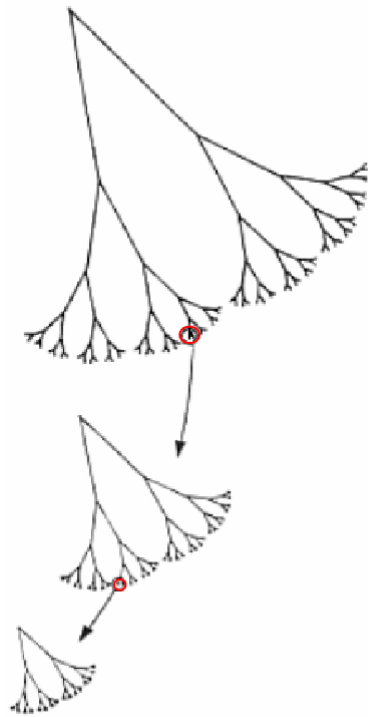
Fractal: A tree-like object or **process**, composed of sub-units (and sub-sub-units, etc) that resemble the larger scale structure

This internal look-alike property is known as *self-similarity* or *scale invariance*.

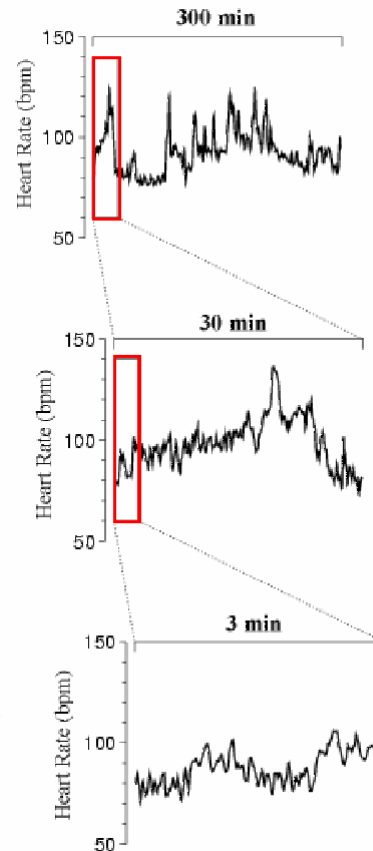


Fractals in Physiology

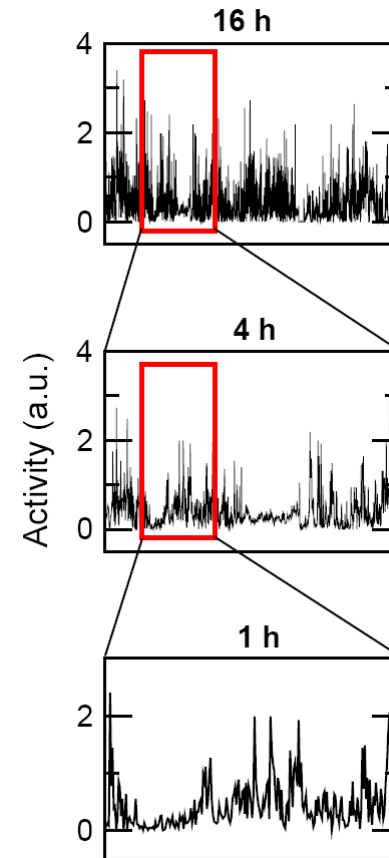
Self-similar/fractal structure



Fractal cardiac dynamics



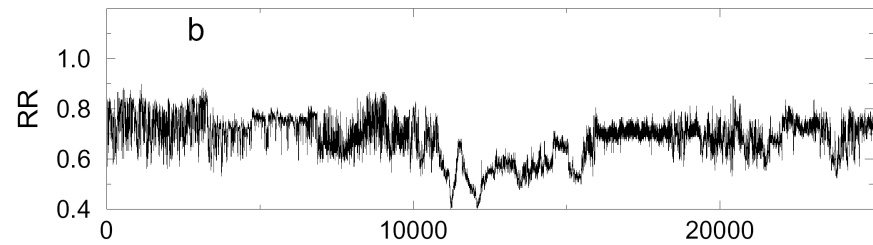
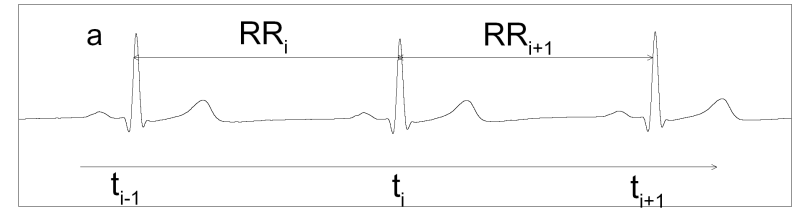
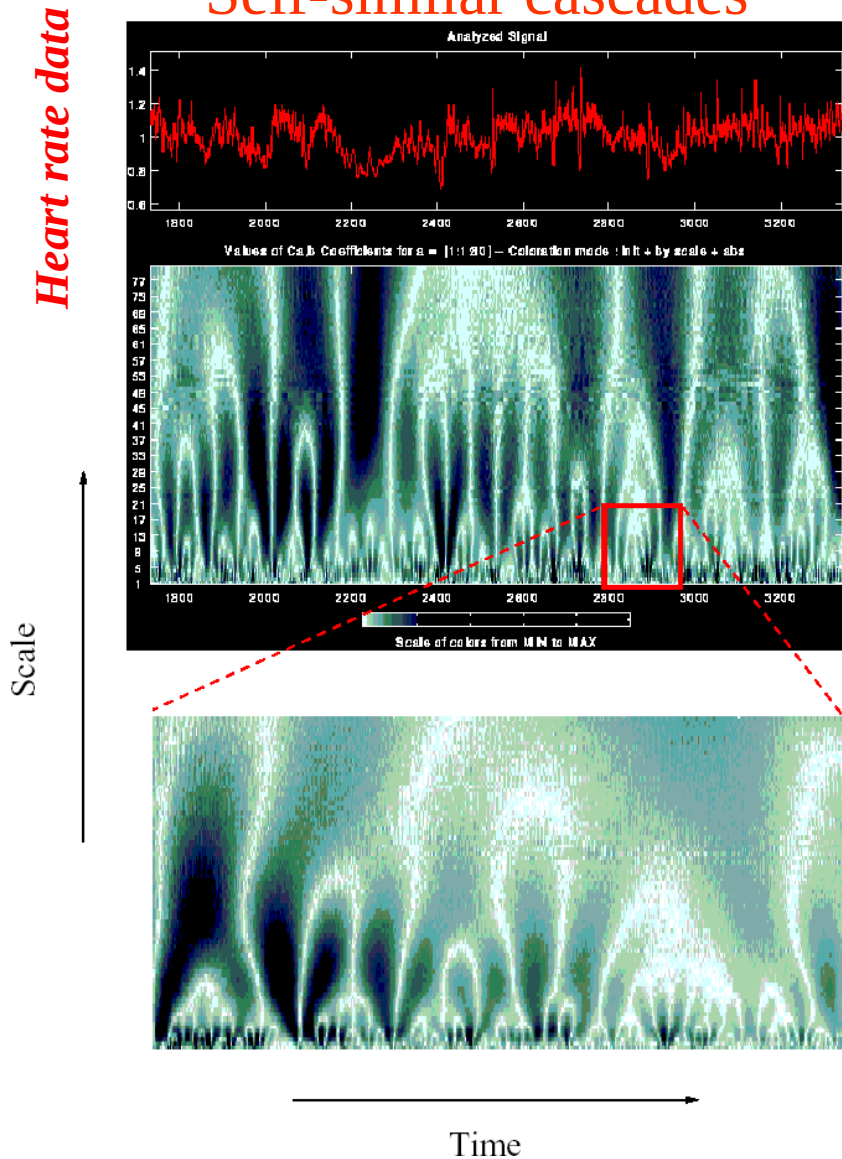
Fractal activity dynamics



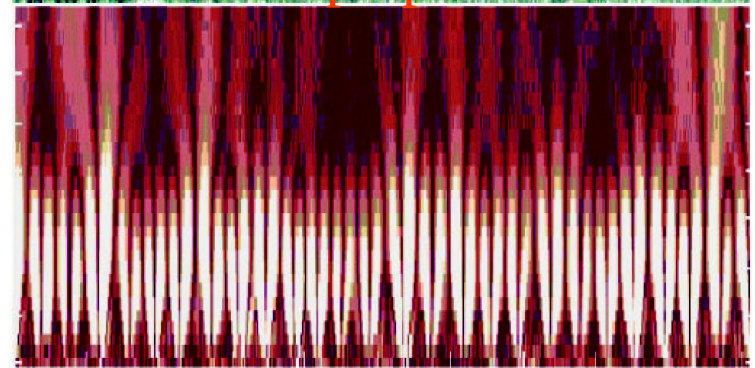
Fractal may be a property of dynamics as well as structure. Fractal fluctuations were observed in a wide range of physiological outputs.

Fractals in heart rate fluctuations

Self-similar cascades



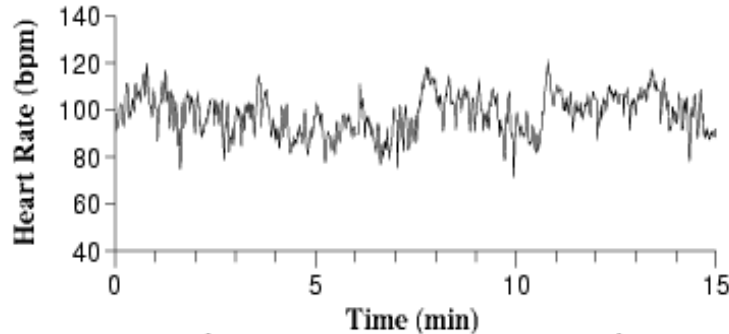
Sleep apnea



- P.Ch. Ivanov *et al.* *Nature* **383**:323 (1996).
P.Ch. Ivanov *et al.* *Physica A* **249**: 587 (1998).
P.Ch. Ivanov *et al.* *Chaos* **11**: 641 (2001).

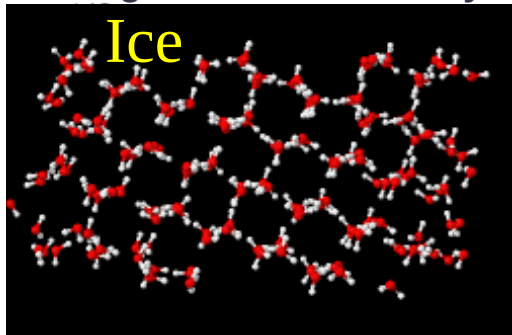
Fractal Complexity Degrades with Disease

Healthy Dynamics: Multiscale Fractal Variability

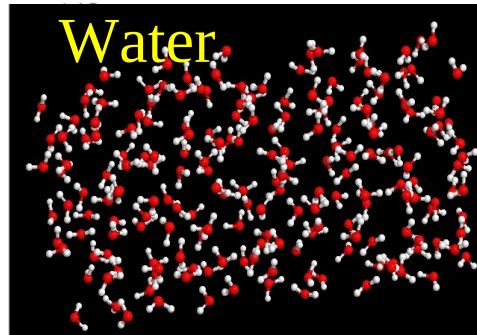


*Two Patterns of
Pathologic Breakdown*

Single Scale Periodicity



Uncorrelated Randomness



- Healthy Heart dynamics poised between too much order and total randomness
- Change or break down under pathologic conditions

Homeostasis Revisited

PHYSIOLOGICAL REVIEWS

Vol. IX

JULY, 1929

No. 3

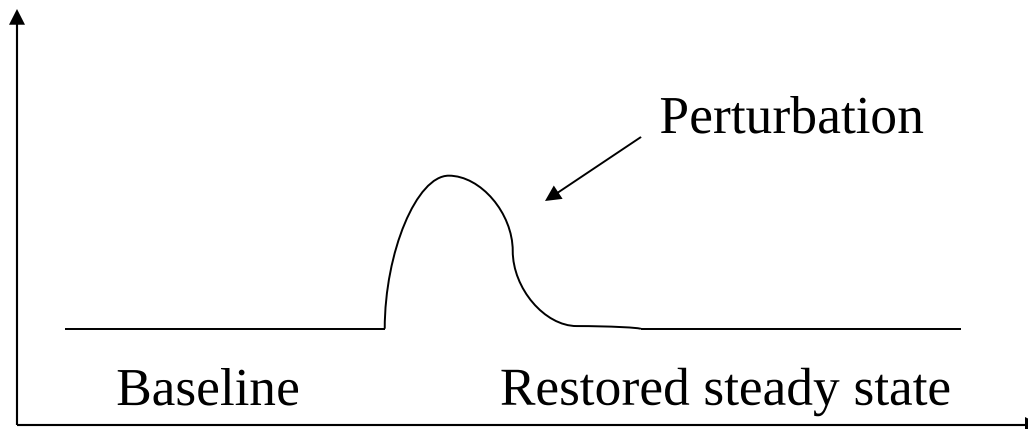
ORGANIZATION FOR PHYSIOLOGICAL HOMEOSTASIS

WALTER B. CANNON



Body as servo-mechanism type machine

- Importance of corrective mechanisms to keep variables “in bounds.”
- Underlying notion of “constant,” “single steady-state,” equilibrium-like” conditions.



...OR

Homeostasis Revisited

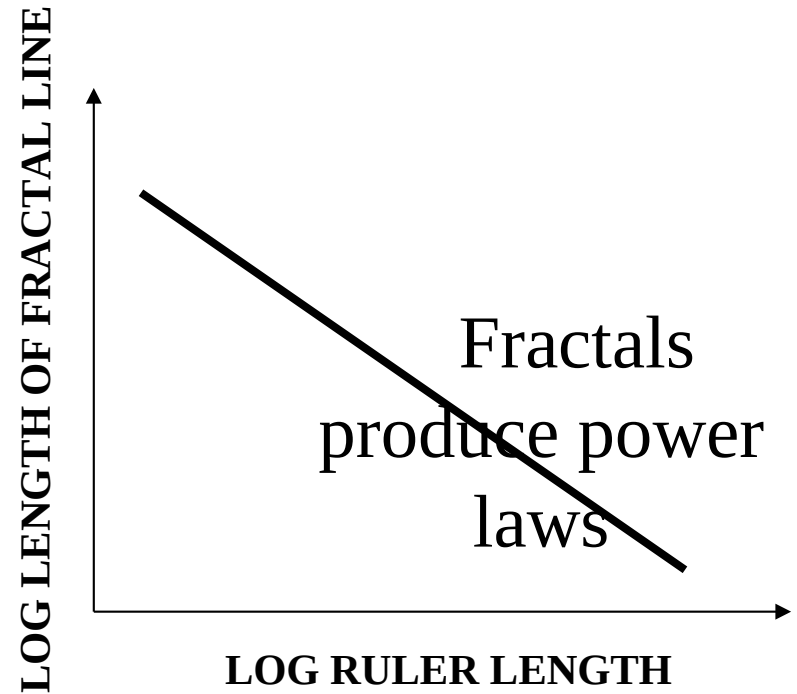
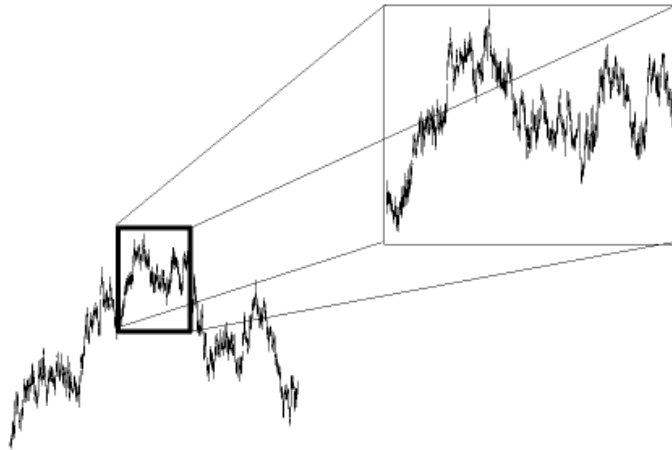
...OR

- Is complex spatio-temporal variability a *mechanism* of healthy stability?
- And, therefore, do we need fundamentally to rethink all notions of mechanisms and causality in physiology

Healthy Dynamics: An Equilibrium State?

Another fallacy. *But* there is an equilibrium state...
...death

Fractals and Power Laws



Method for quantification of fractals

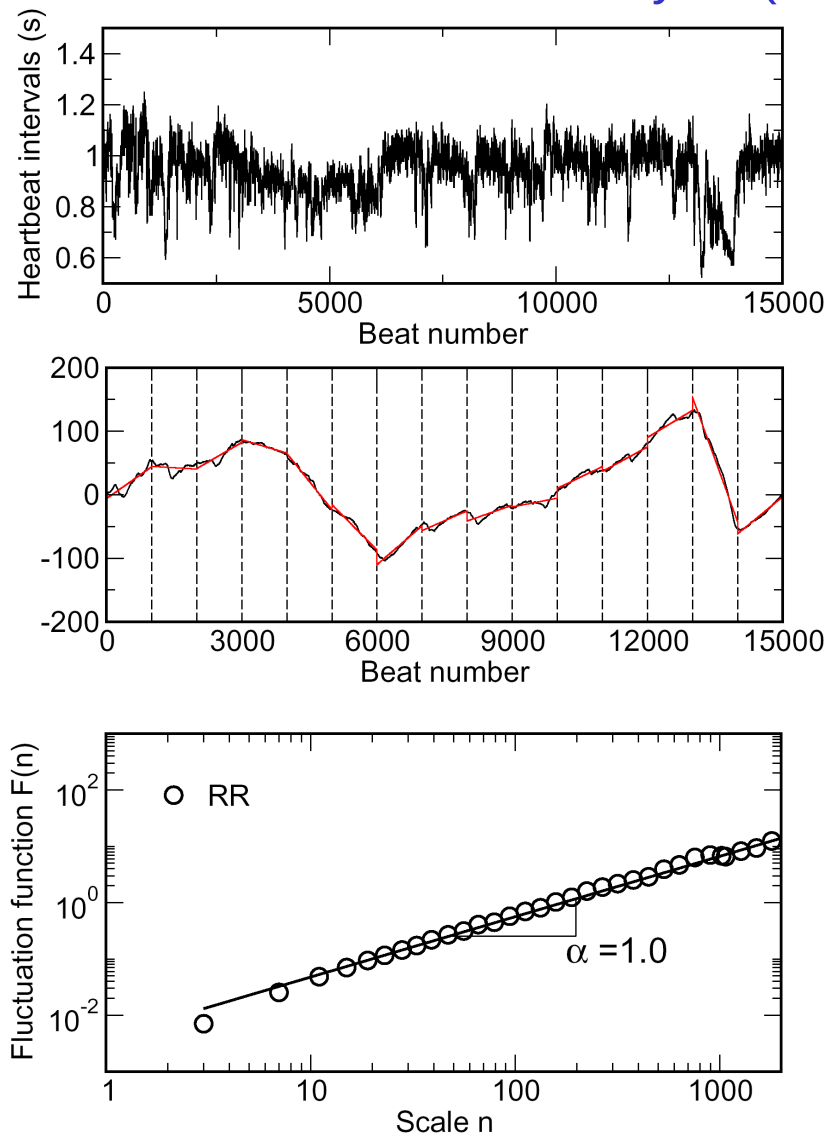
- Box counting (fractal dimension)
- Power spectral analysis
- Hurst analysis
- Root mean fluctuation analysis
- Detrended fluctuation analysis (DFA)
- Detrended moving average analysis
- Wavelet-based multifractal analysis

...

Hu K, *et al.*, *Phys. Rev. E* **64**, 011114 (2001)
Chen Z, *et al.*, *Phys. Rev. E* **65**, 041107 (2002)
Chen Z, *et al.*, *Phys. Rev. E* **71**, 011104 (2005)
Xu L, *et al.*, *Phys Rev E* **71**:851101 (2005)

Method for quantification of fractals:

Detrended fluctuation analysis (DFA)



Quantification of fluctuations at different window size

- Remove mean
- Integrate
- Detrend using polynomial functions in all windows
- Obtain average fluctuations

Log-log plot (power-law form indicates scale invariance)

Scaling exponent reveals correlation property:

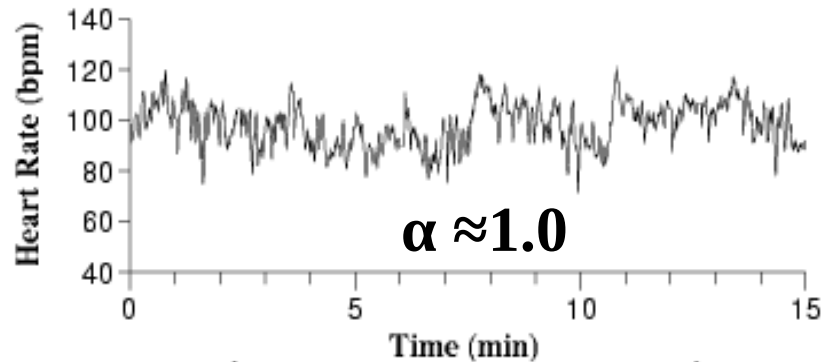
$\alpha=1.0$: the most complex fractal correlations

$\alpha=1.5$: a random walk (Brownian noise)

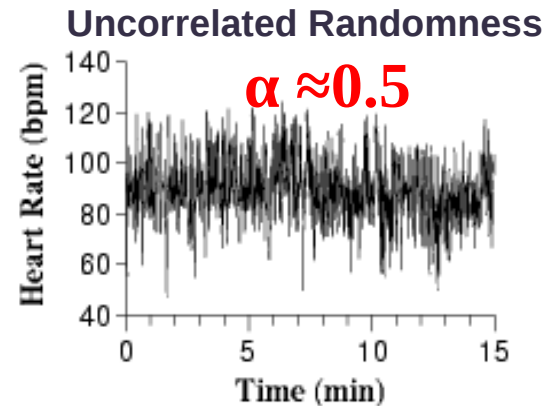
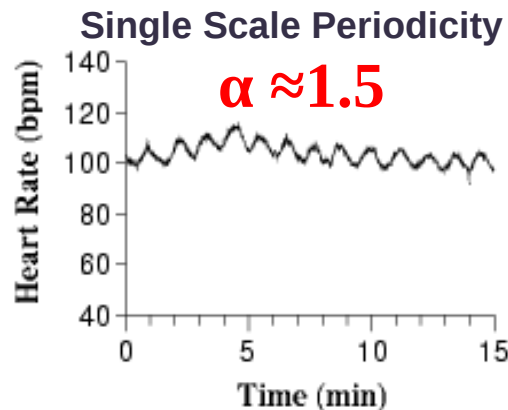
$\alpha=0.5$: a white noise

Scaling exponent as an index of fractal correlation properties

Healthy Dynamics: Multiscale Fractal Variability

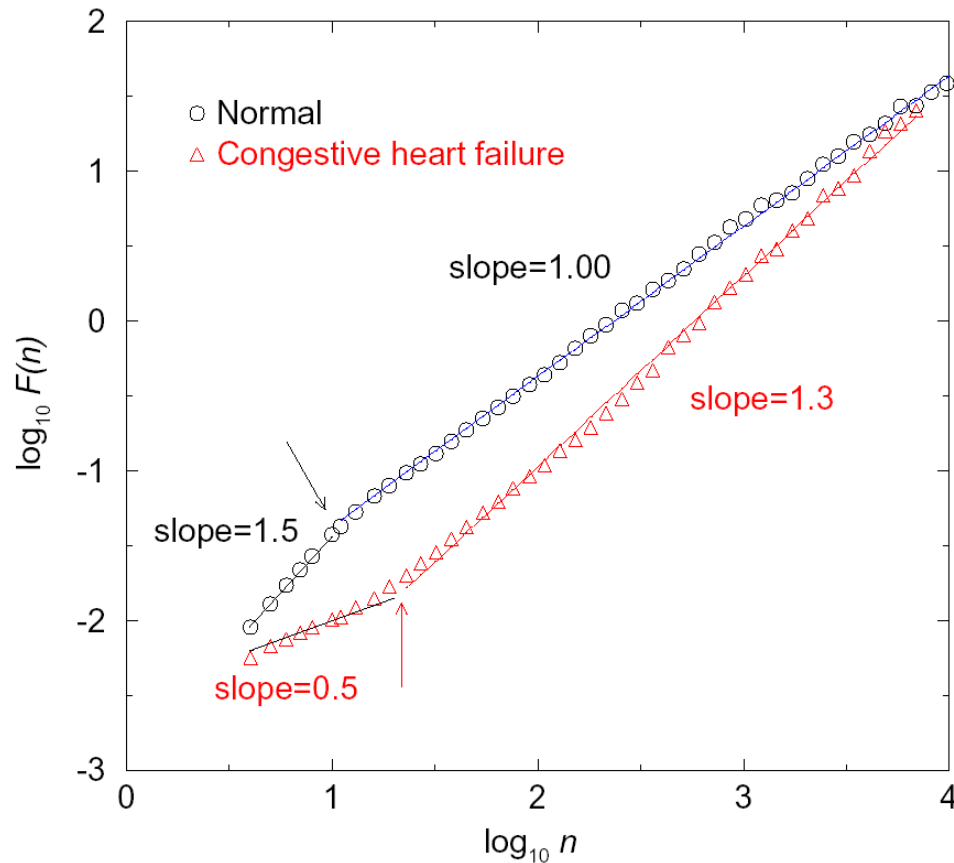


*Two Patterns of
Pathologic Breakdown*



Scaling exponent α indicates correlation properties in fluctuations.

Degraded fractal correlations under pathologic conditions



Congestive heart failure :

$\alpha \approx 1.3$ for $30 \sim 10^4$ beats

closer to a random walk.

Peng CK. *et al.*, *Chaos* **5**, 82 (1995);

Peng CK *et al.*, *J. Electrocardiol.* **28**, 59 (1996).

Turcott RG, Teich MC, *Ann. Of Biomed. Eng.* **24**, 269 (1996).

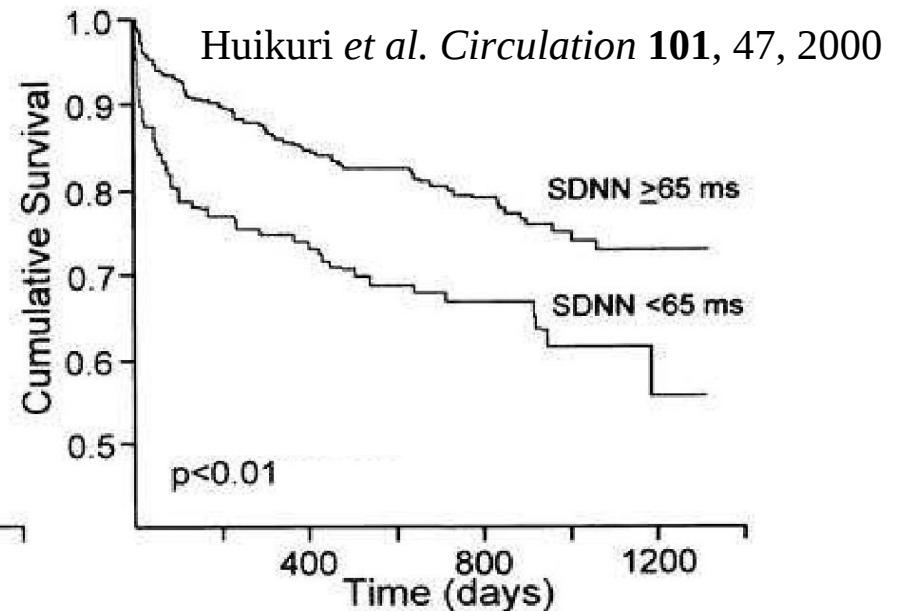
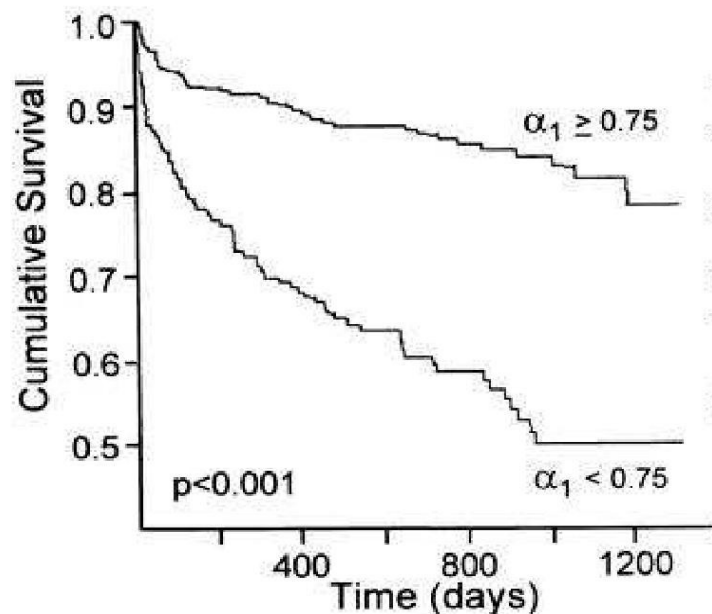
Ho KKL. *et al.*, *Circulation* **96**, 842 (1997).

Mietus JE *et al.*, *Comp. Cardiol.* **27**, 753 (2000).

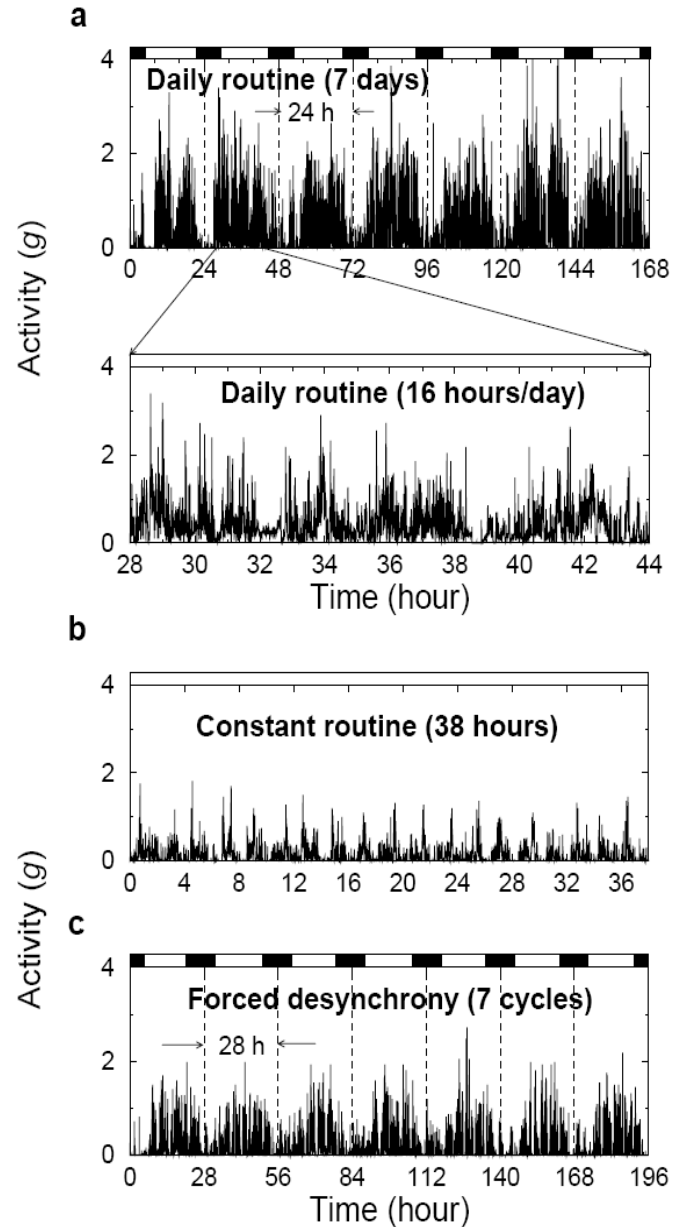
Clinical Utility of DFA:

Fractal cardiac dynamics (DFA) can also predict

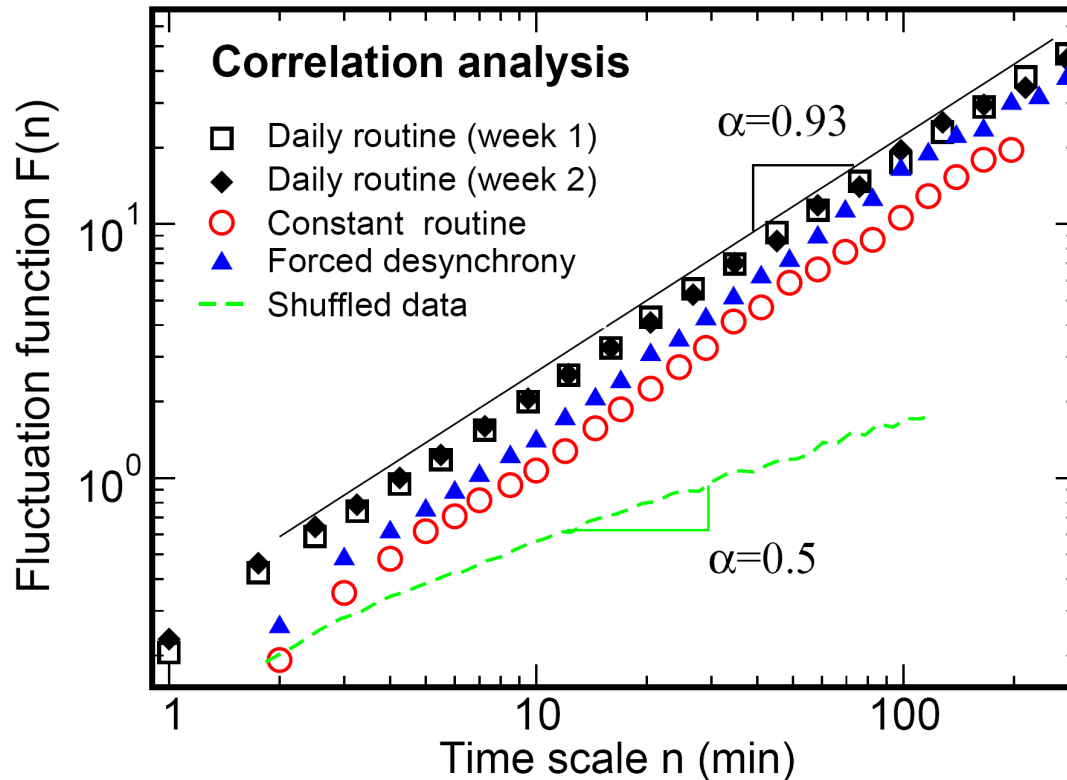
- survival rate of patients after myocardial infarction (Huikuri *et al. Circulation* **101**, 47, 2000).
- mortality of patients after stroke (Makikallio AM *et al. Neurology* **62**, 1822-1826, 2004)
- successful defibrillation in patients with out-of-hospital cardiac arrest (Lin LY *et al. Resuscitation* **81**, 297-301, 2010)



Human motor activity fluctuations



Fractal correlations in human motor activity fluctuations



- Strong correlations ($\alpha > 0.5$ and close to 1)
- Stable over a wide range of time scales
- Independent of scheduled events and mean activity levels

Human studies

1. Young control :

2. Elderly control:



Age-matched

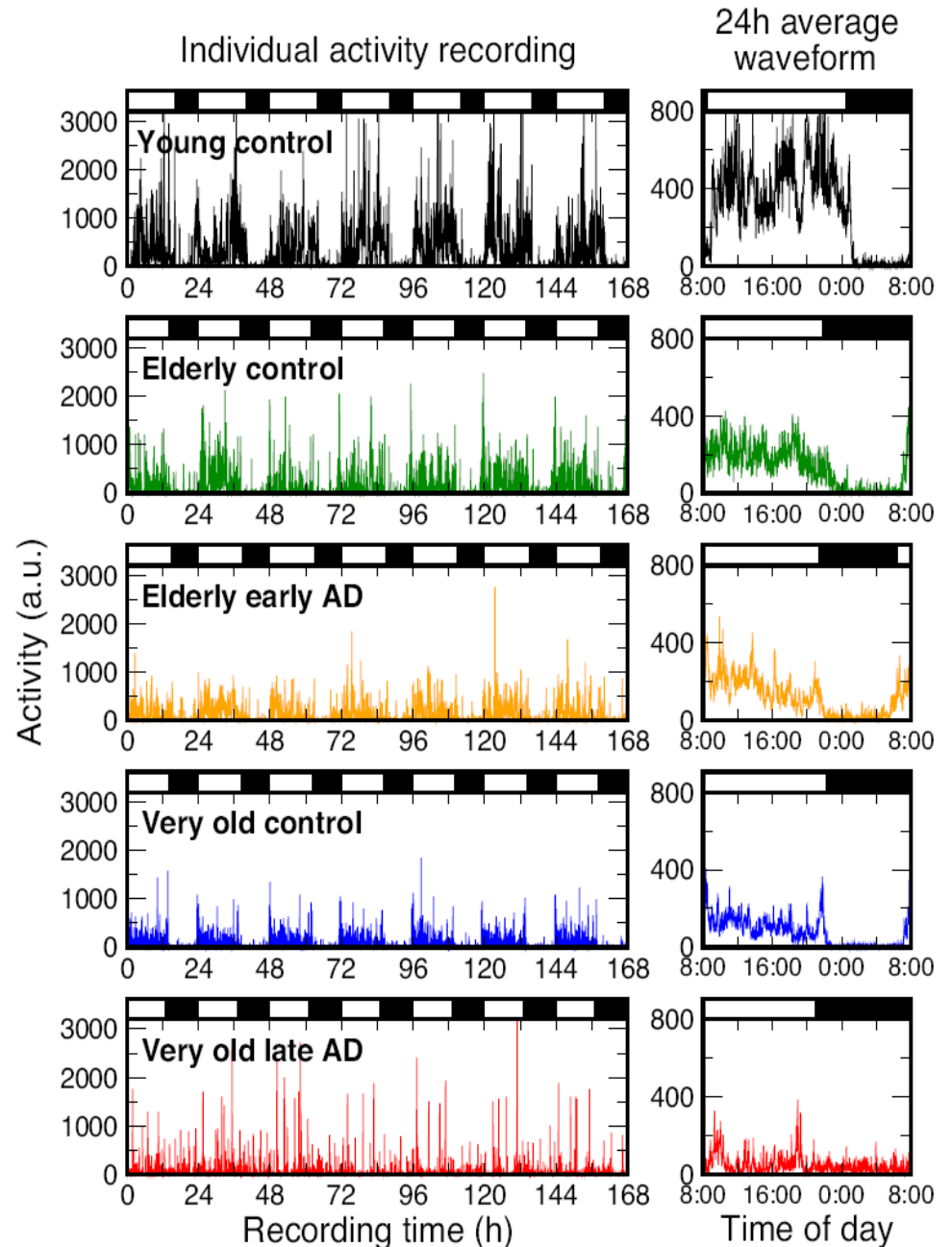
3. Elderly Early-stage AD:

4. Very old control:

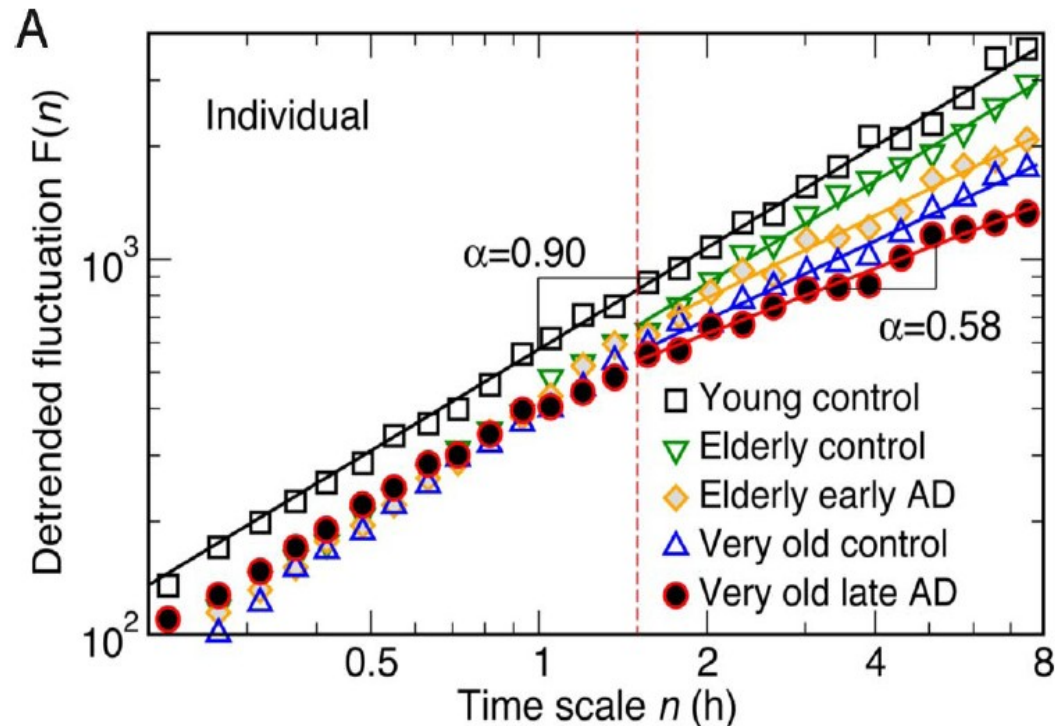


Age-matched

5. Very old Late-stage AD:

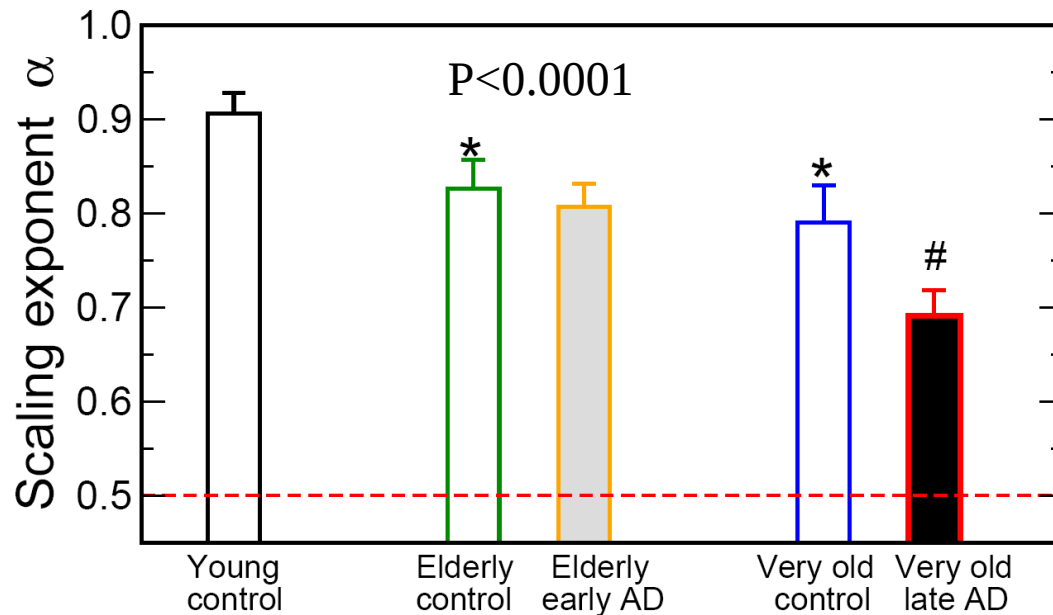


Altered activity fractal fluctuations with aging and AD



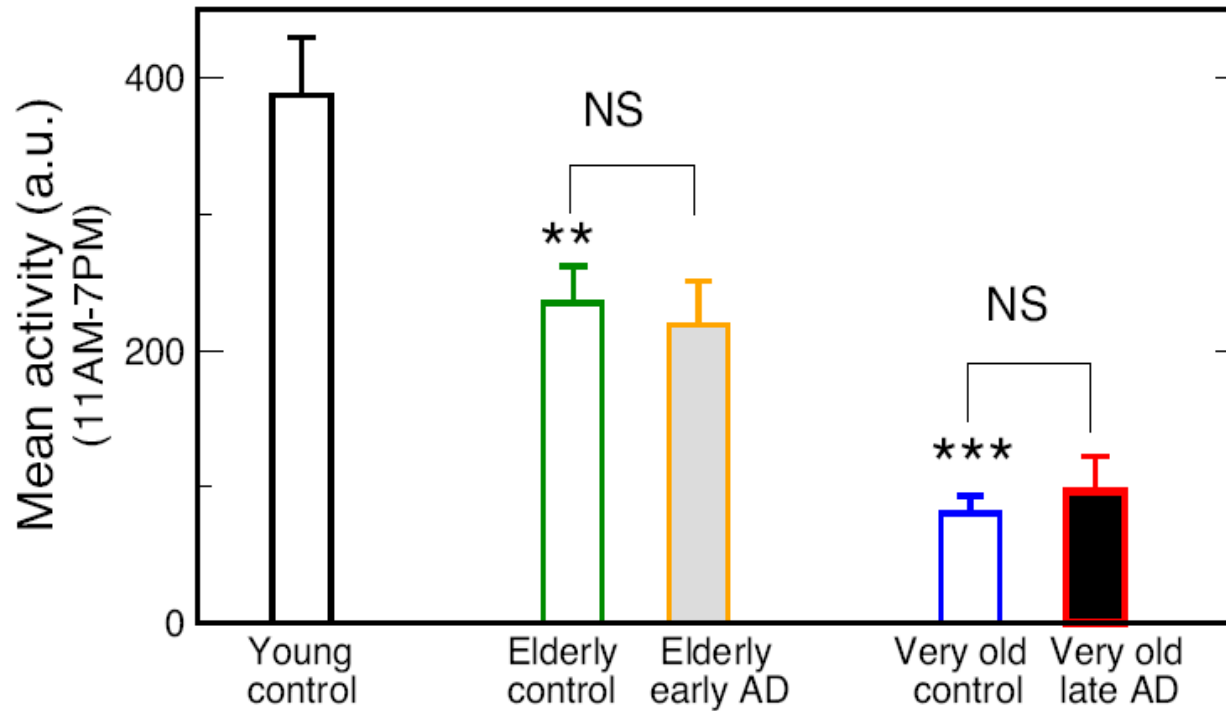
- Robust scale-invariant patterns at time scales from minutes to 10 h in young control
- Similar patterns in other groups at small time scales (<1 h)
- Reduction in fractal correlations at $>\sim 2$ h with aging and AD

Effect of aging and AD on fractal activity correlations



- There was significant group effect ($p < 0.0001$)
- The elderly control and very old control subjects had smaller α values than young controls, as indicated by * ($p < 0.05$).
- Early-stage AD subjects and age- and living condition-matched elderly controls showed no significant difference.
- Late-stage AD subjects has significantly smaller α than the age- and living condition-matched very old controls, as indicated by # ($p < 0.05$).

Effect of aging and AD on the scale-invariant activity regulation



Why is it healthy to be fractal?

- Healthy function requires capability to cope with the unpredictable
- Scale-invariant systems generate broad repertoire of responses & long-range correlations → adaptability & plasticity
- Absence of characteristic time scale helps prevent mode-locking (pathologic resonances)

Neurobiological circuitry unknown and meaningful models lacking!

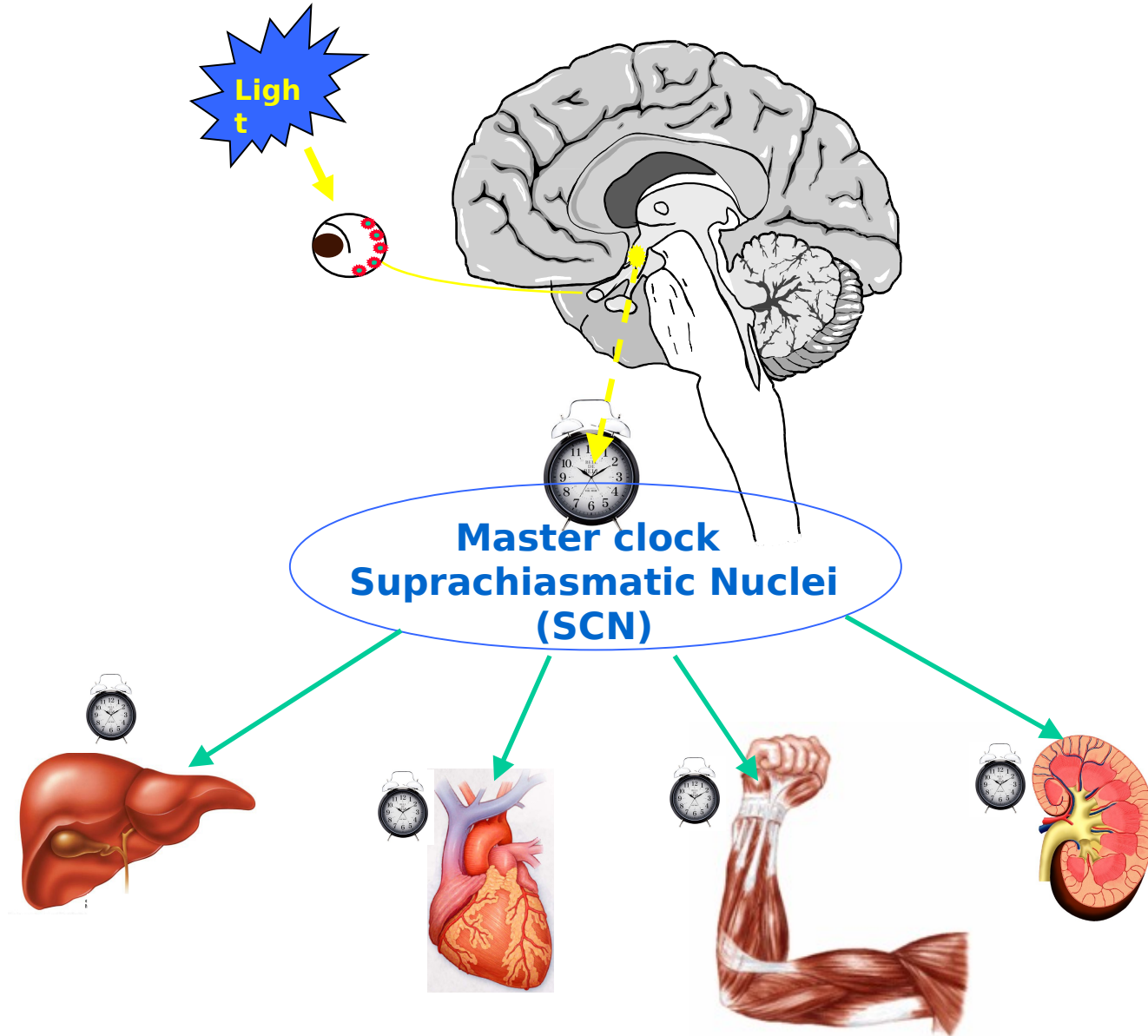
“Understanding scale invariance (fractal) is one of the top five great ideas in biology ...”

Sir Paul Nurse,
(winner of the 2001 Nobel Prize in Physiology or Medicine)

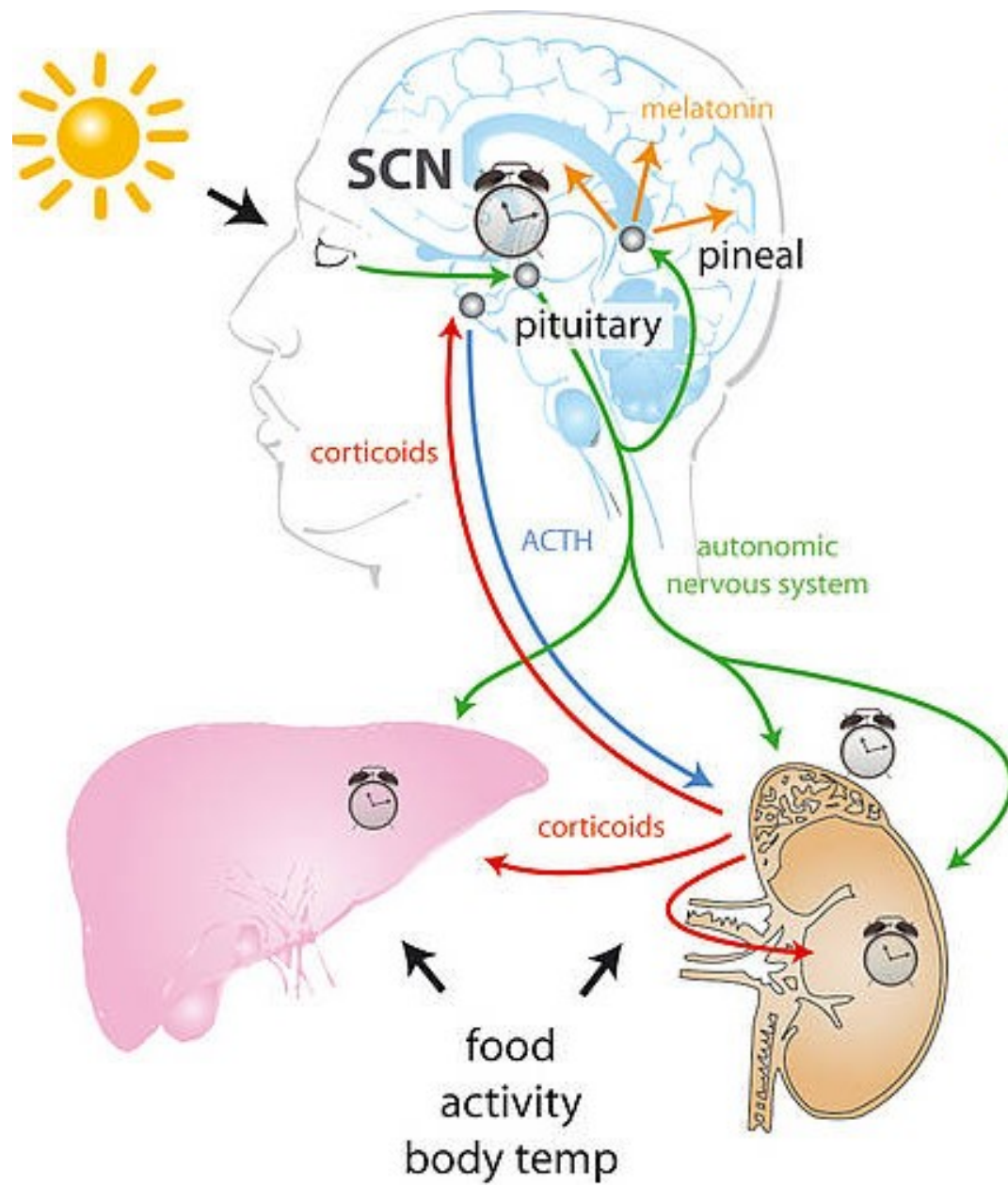


http://estream.med.harvard.edu:8080/ramgen/Content/CustomVideo/Leaders_in_Biomedicine/C_03202007105445.rm

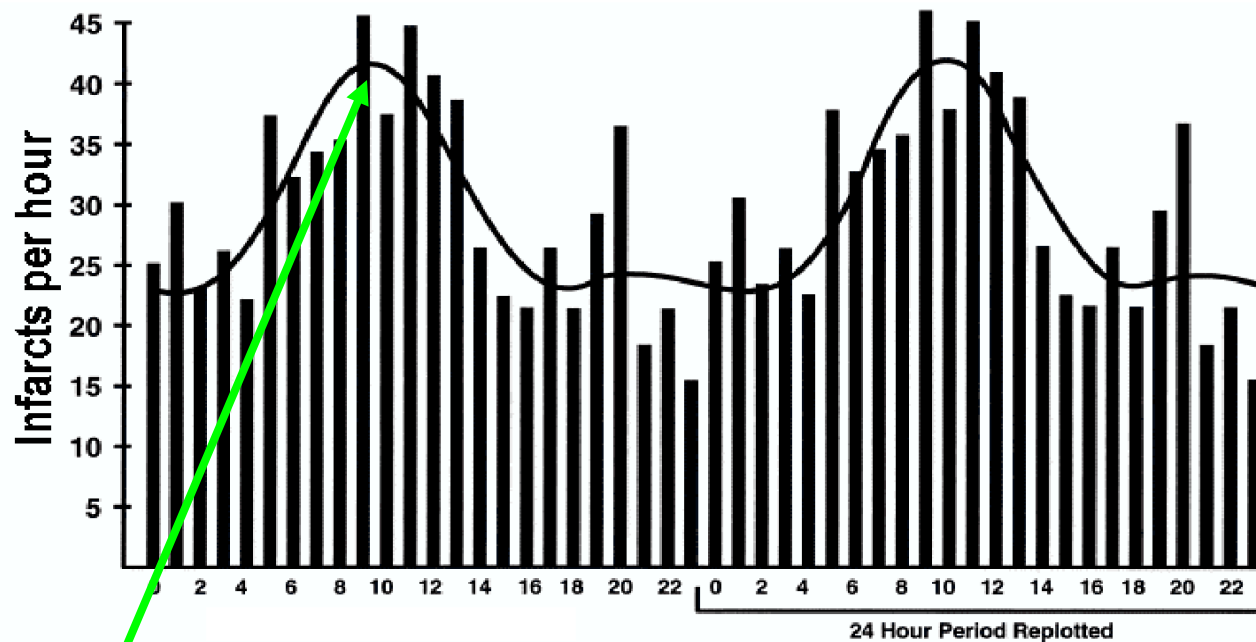
Effects of the circadian system on physiological functions



Coordinated rhythms of ~24 h in Physiological functions



Daily pattern of cardiac risk



Muller JE, et al., *N. Eng. J. Med.* **313**, 1315 -1322 (1985).

Highest cardiac vulnerability/heart attacks at 9-11 AM

Internal physiologic factors

External activity factors

How to separate these two factors ?

Experimental methods to distinguish the effects of intrinsic circadian system from behaviors

Data was collected in two complementary protocols performed in dim light:

Behavior cycles were uniformly distributed across all circadian phases

Forced desynchrony

6:00 12:00 18:00



5 healthy subjects; Sleep period delayed 4 hours each day; Same scheduled behavior in 28-h sleep/wake cycles.

Behavior was kept constant

Constant routine

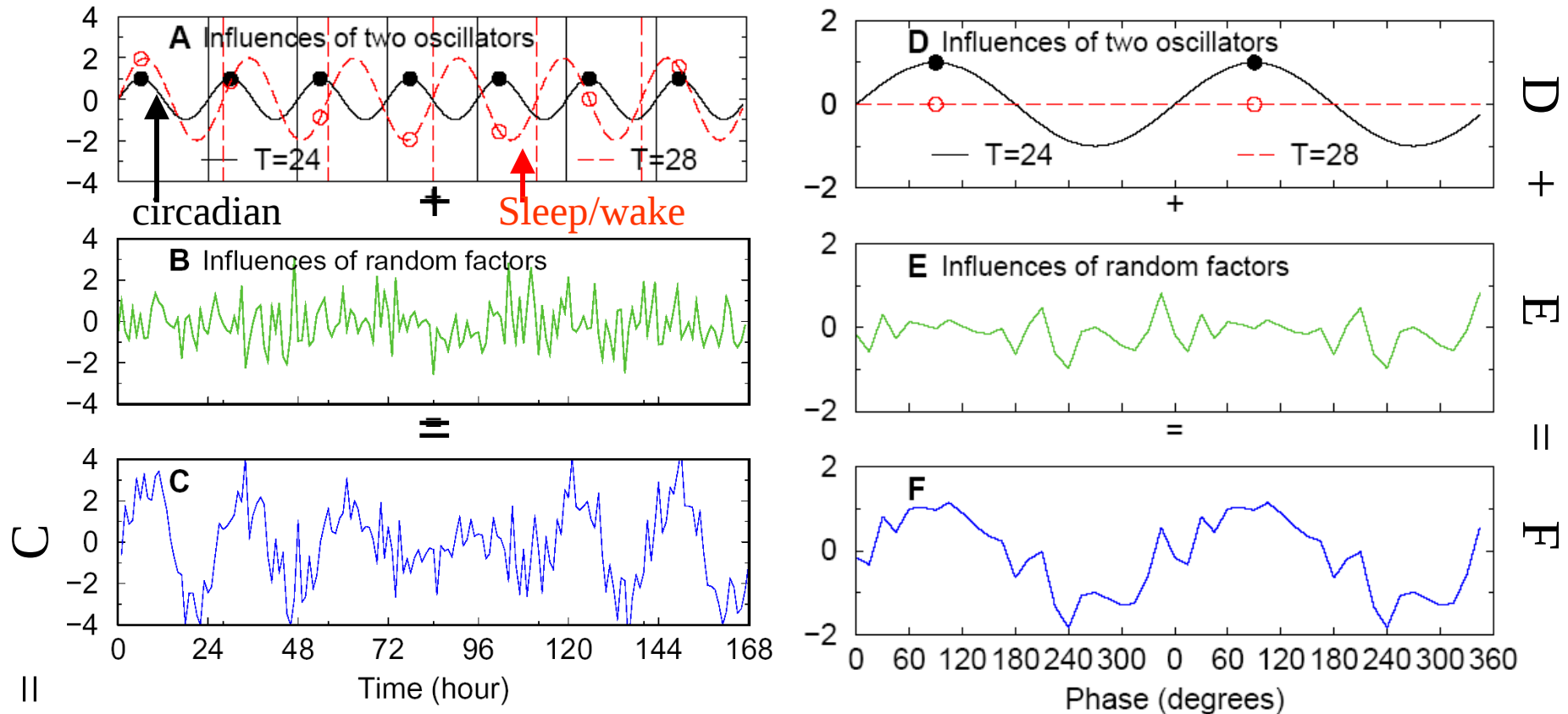
6:00 12:00 18:00



9 healthy subjects; Semi-recumbent and awake for 38 hours; identical snacks provided every 2 h.

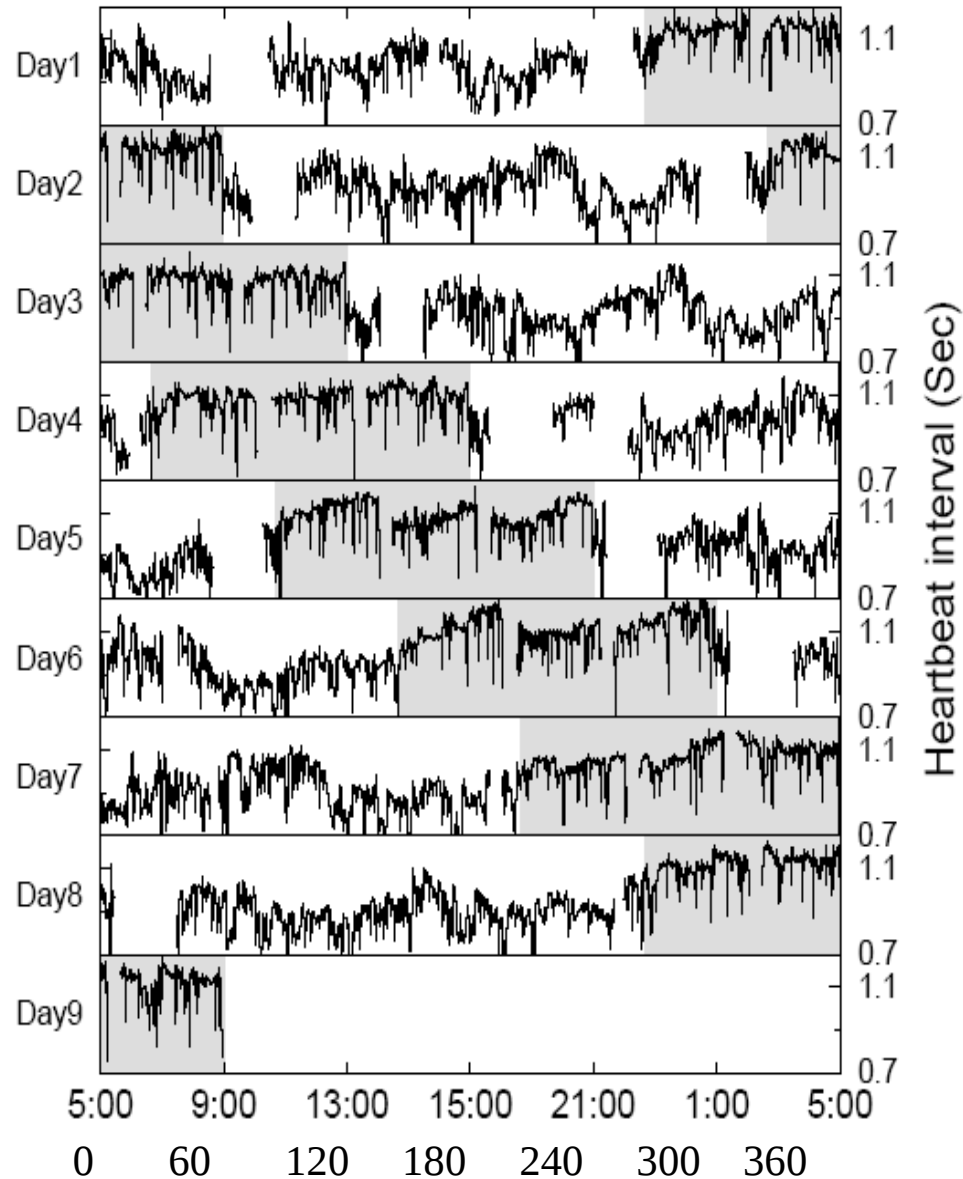
Czeisler CA *et al.*, *Science* **284**, 2177-2181 (1999).
Dijk DJ. & Czeisler CA, *Neurosci. Lett.* **166**, 63-68 (1994).

Schematic illustration of methods of assessing endogenous circadian rhythm

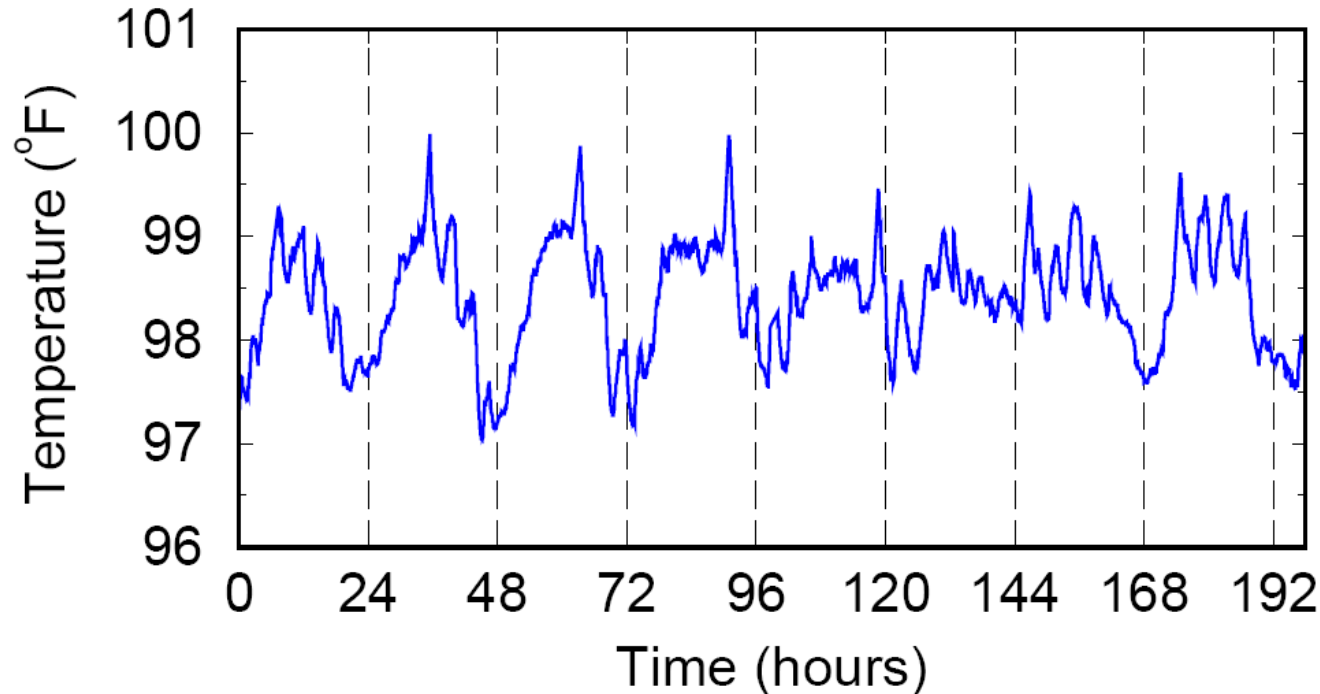


Need sensitive marker to determine the circadian phases.

Forced desynchrony protocol: a heartbeat recording



Estimation of internal circadian phases



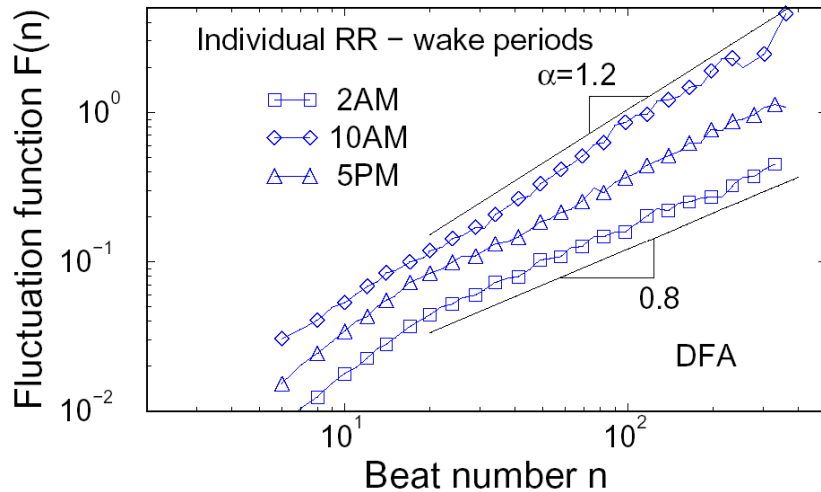
- Body temperature as the marker;
- Phase = 0 for the minimum of body temperature.

Dijk DJ & Czeisler CA. *Neurosci. Lett.* **166**, 63–68 (1994).

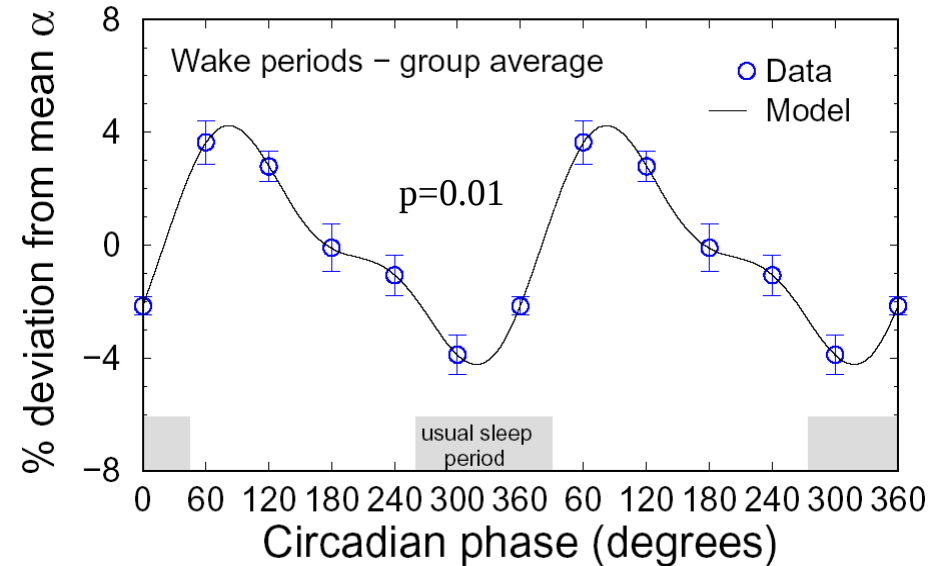
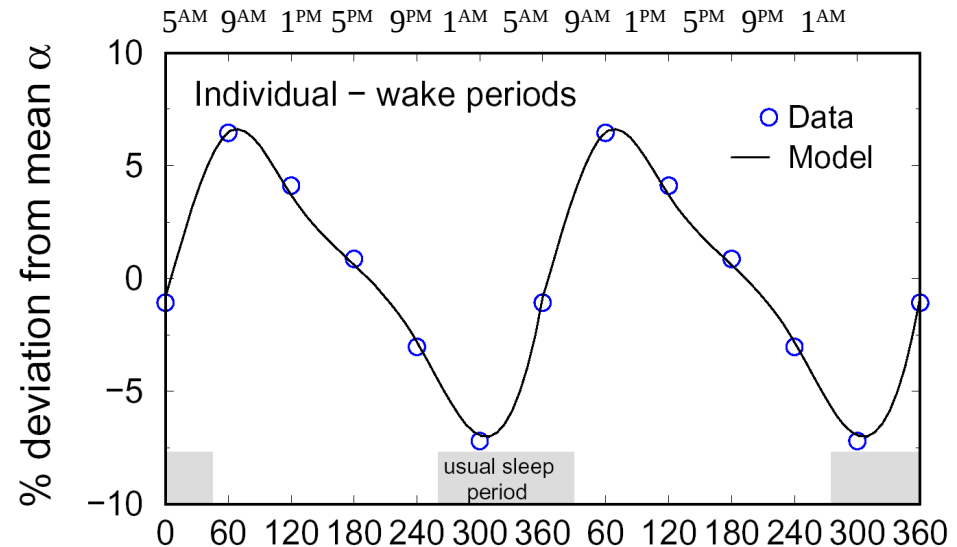
Brown EN & Czeisler CA. *J. Biol. Rhythms*; **7**, 177–202 (1992)

Khalsa SB, et al. *J. Biol. Rhythms* **15**, 524–530 (2000).

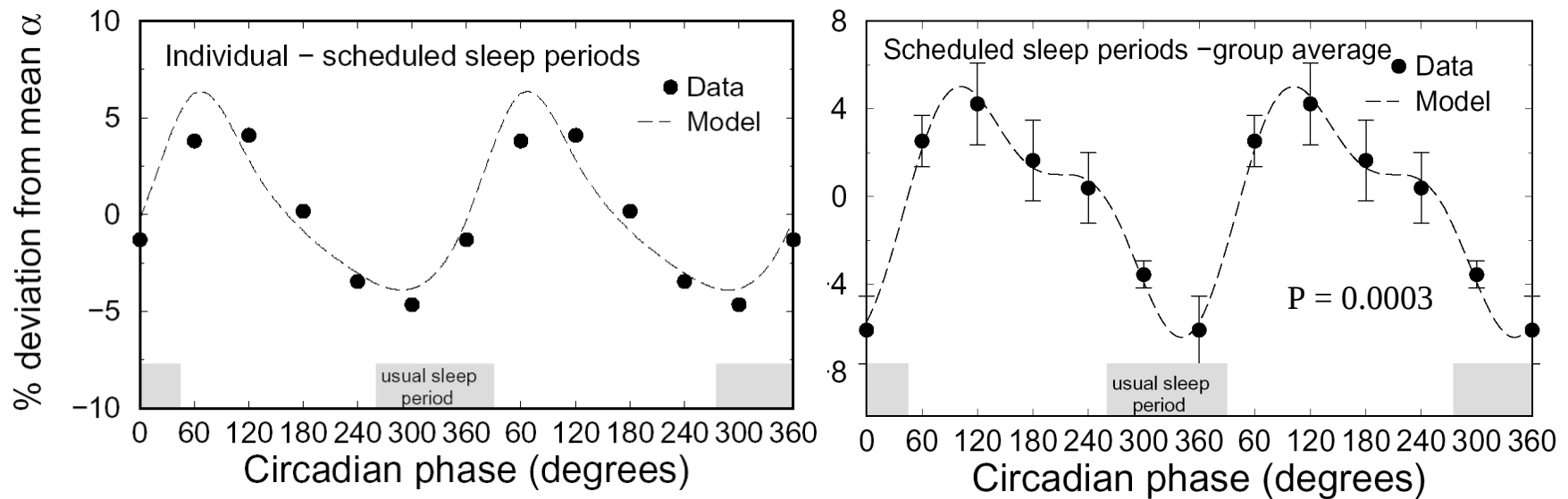
Circadian rhythms in cardiac fractal correlations: wakeful periods



- Fractal: $F(n) \sim n^\alpha$
- Circadian rhythms in α
- A sharp peak at 60-90 degrees (corresponding to 9-11 AM);
- A minimum at 300-330 degrees (corresponding to 1-4 AM).

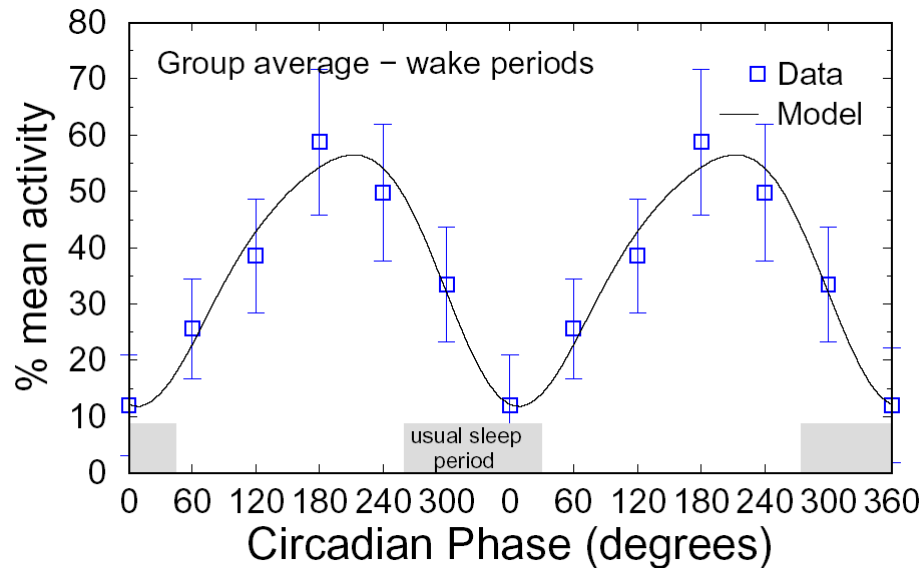
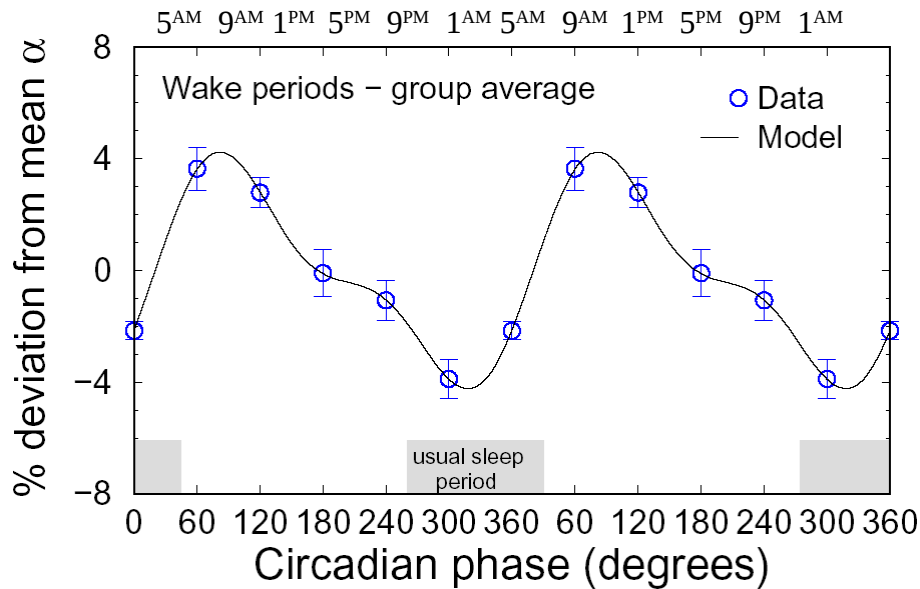


Circadian rhythms in cardiac fractal correlations: sleep periods



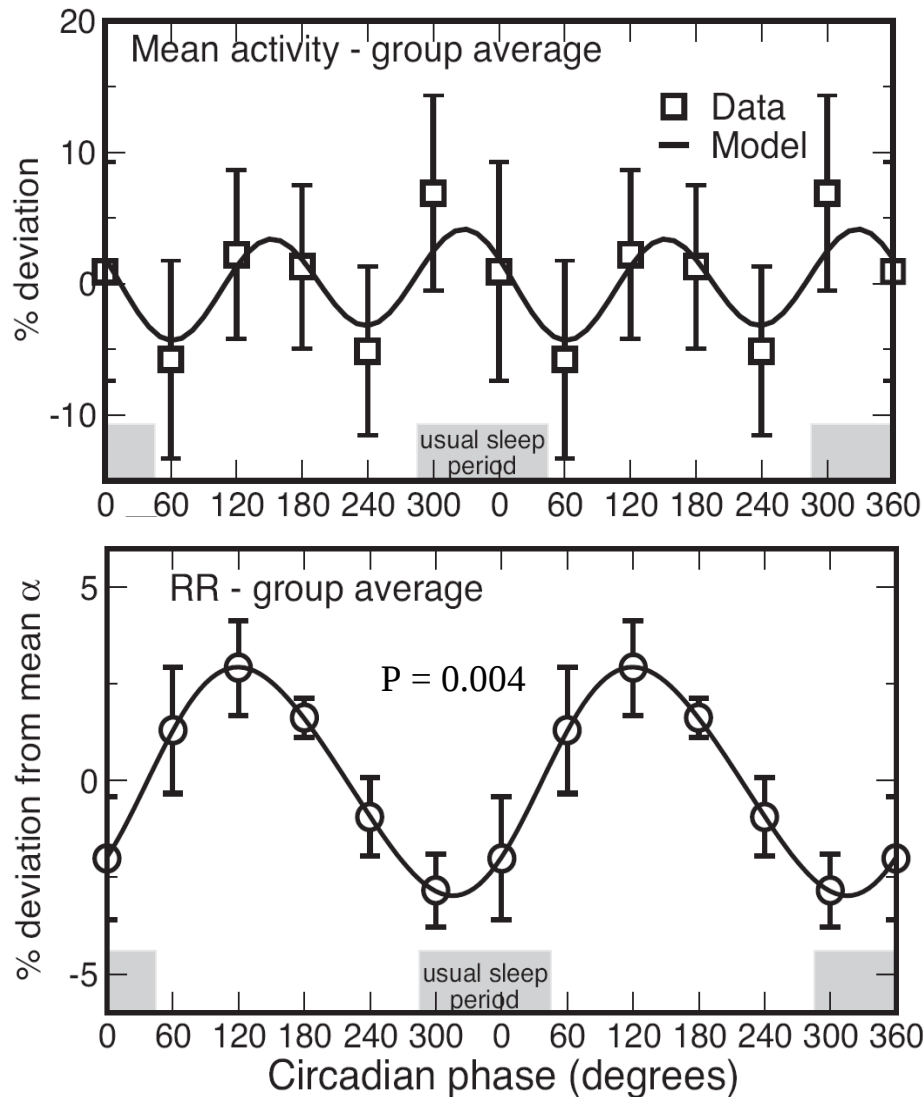
Consistent and significant circadian rhythms during sleep periods.

Circadian rhythm in motor activity



- Significant circadian rhythms in mean activity.
- The circadian pattern in α with a peak at 9-11 AM is **very different** from the pattern in activity with a peak at 5-9 PM.

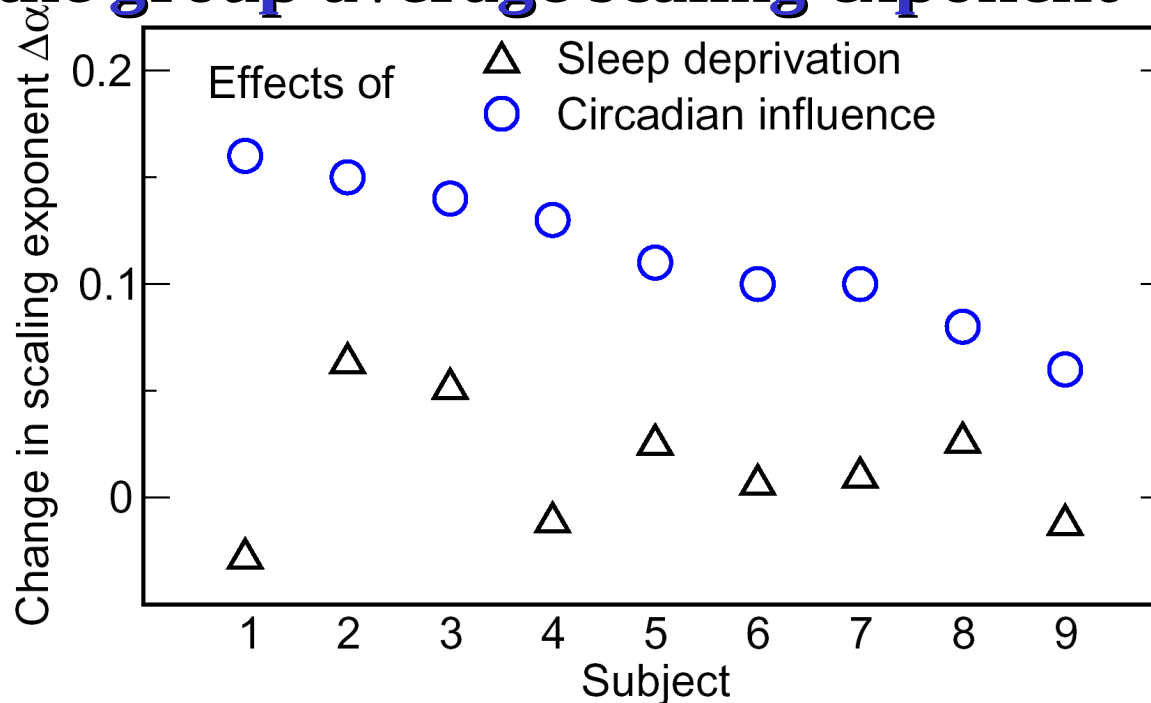
Consistent circadian rhythms in cardiac fractal correlations during the constant routine protocol



- Abolished circadian rhythms in mean activity

- Circadian rhythms in cardiac fractal correlations persist

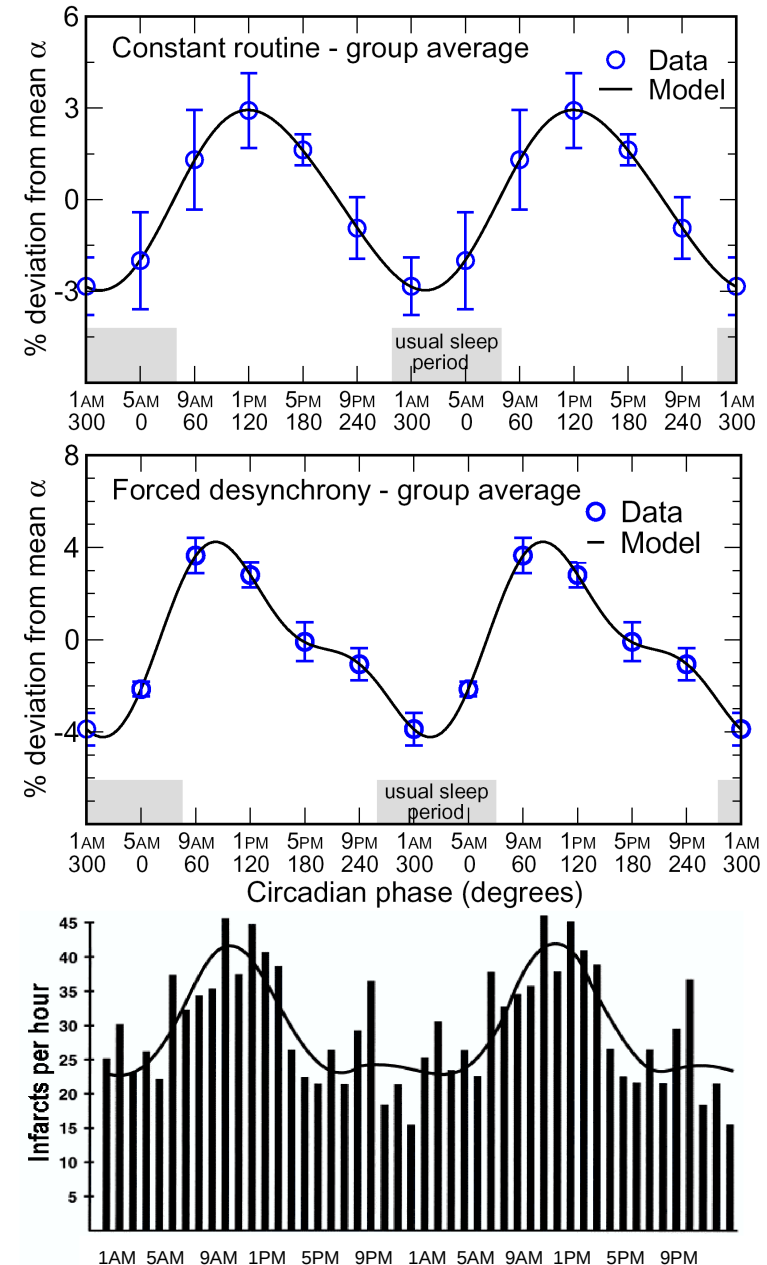
No significant influence of acute sleep deprivation on the group average scaling exponent



- The changes of the scaling exponent along the linear trend in 12 hours were much smaller than the peak to trough amplitude of circadian rhythms.
- Linear trends of the scaling exponent ($P < 0.05$) were significant only in 3 out of 9 subjects.
- No significant changes in the group average of the scaling exponent throughout the constant routine protocol.

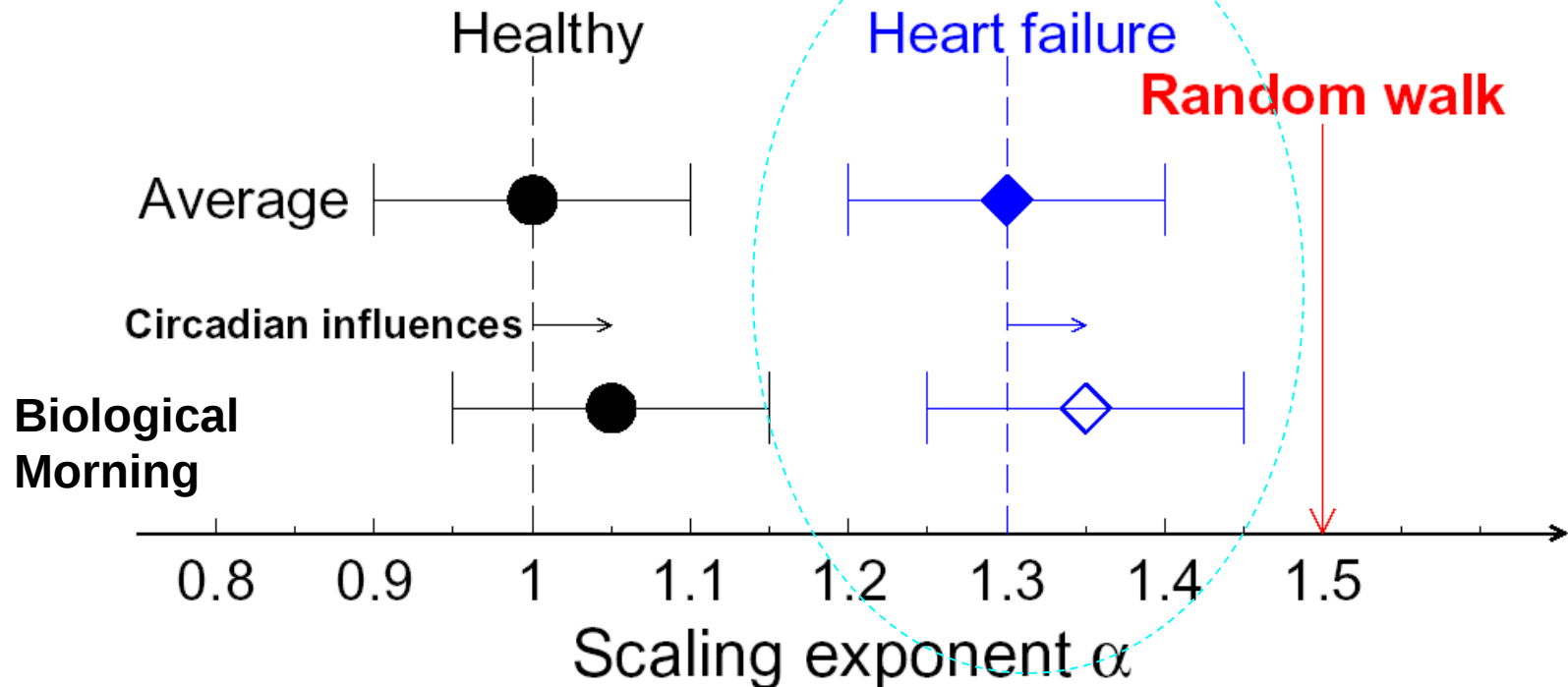
Circadian rhythms in fractal heartbeat correlations

- Similar circadian rhythms in α during both protocols
- Maximum α during biological morning
- Circadian peak of α close to the time window of highest cardiac risk observed in epidemiological studies.



Muller *et al.*, *N. Eng. J. Med.* **313**, 1315 (1985).

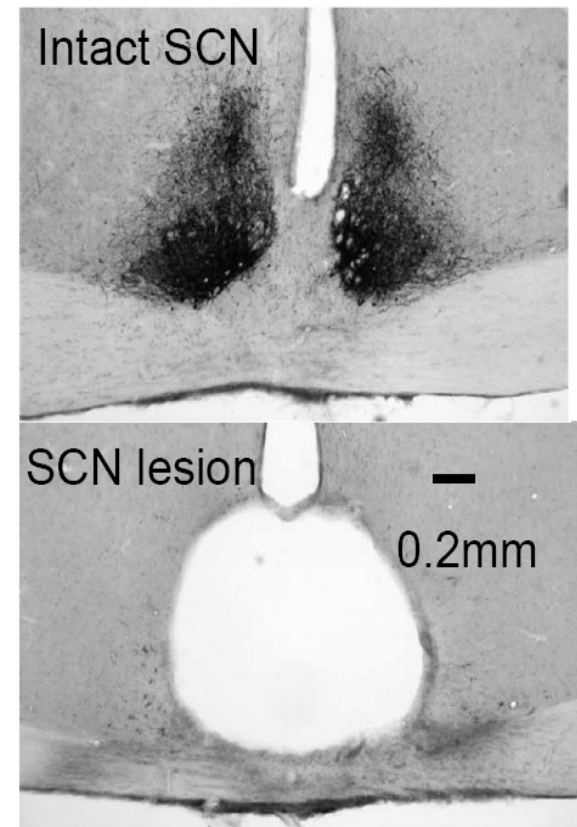
Hypothesis: Relevance to cardiac risk



A similarly circadian effect in subjects with congestive heart failure may bring the value of their exponent α even closer to the critical value of 1.5 at specific time window during biological morning.

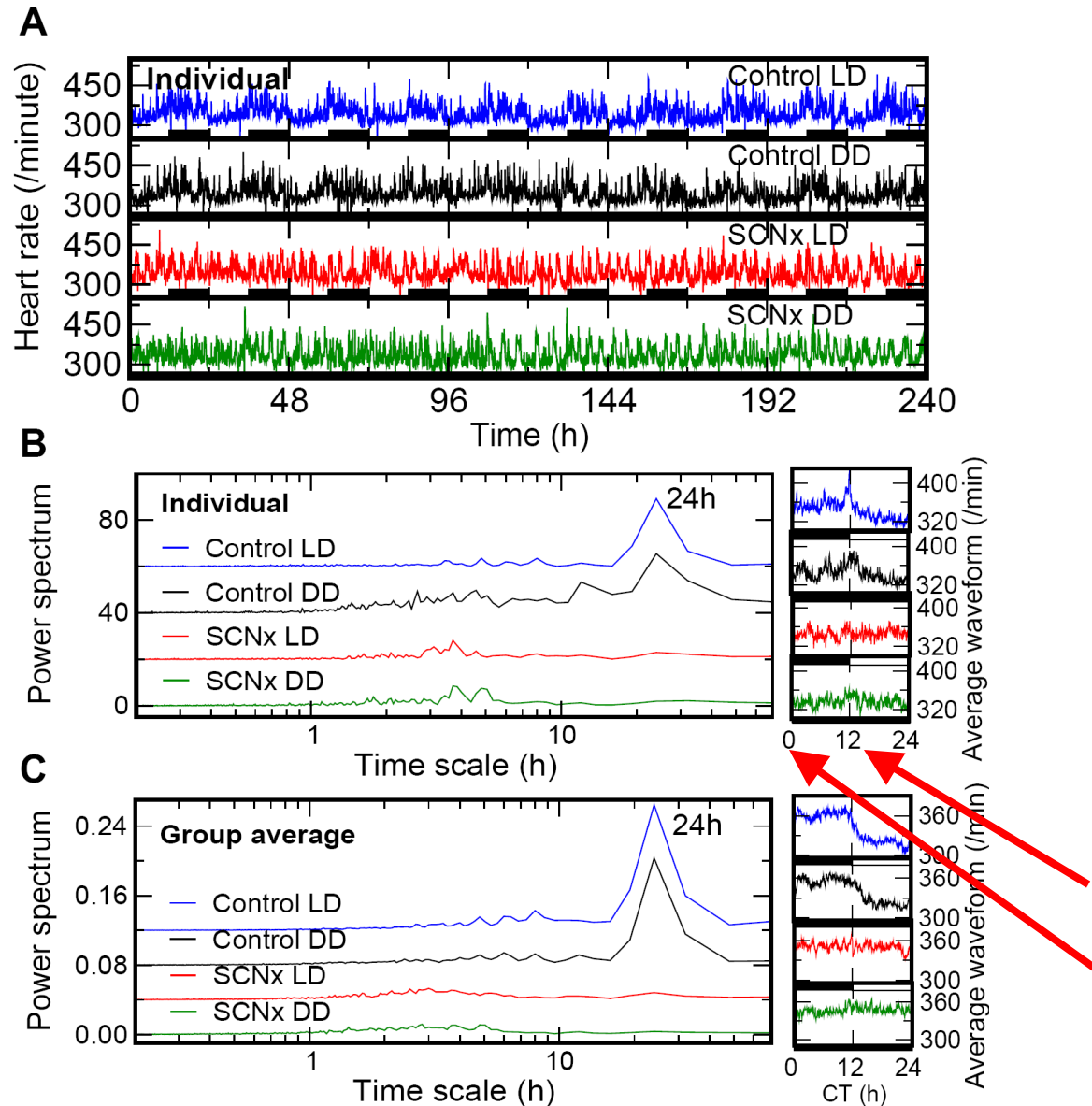
Animal study

- 8 control and 6 SCN-lesion (SCNx) Wistar rats
- Under 10-day 12-hour:12-hour light/dark condition (LD) followed by 10-day constant dark condition (DD)
- Movements cross the space in a cage 39x38x38 cm
- One data point every 4 minutes (10 sec sampling at 500Hz)
- Core body temperature as circadian phase marker. Mean circadian period of control rats (24.1h) was used for SCNx rats

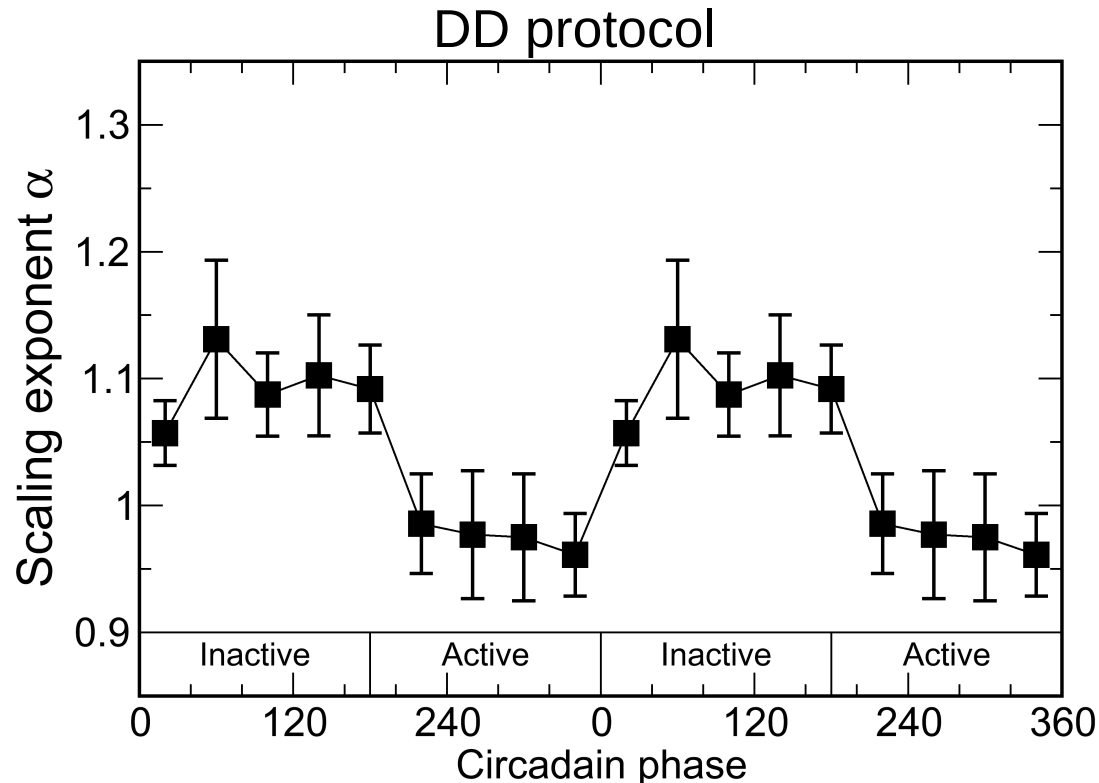


Scheer et al., Neuroscience 2005; 132, 465-477.

Circadian rhythms are abolished by SCN lesion



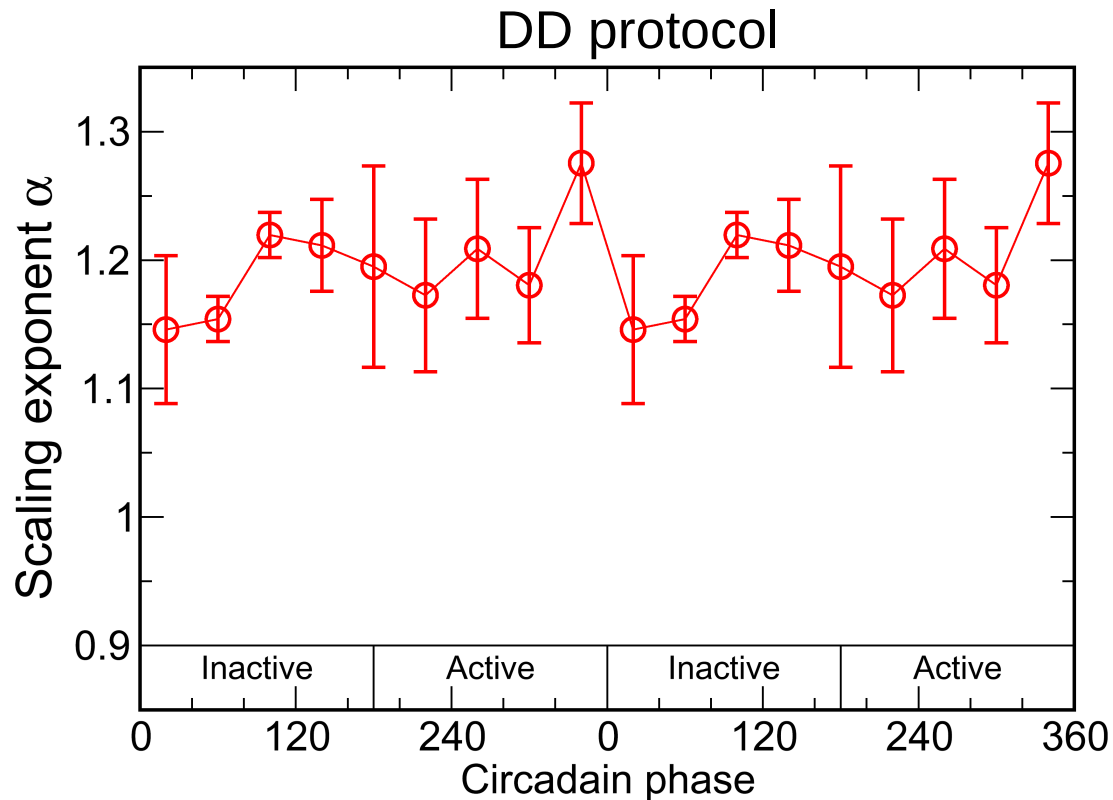
Circadian rhythms in cardiac fractal correlations



In control rats:

- scale exponent $\alpha=1.03\pm0.04$ (Mean \pm SE)
- a significant circadian rhythm ($p=0.0006$)
- peak during the biological day (inactive phase for rats)
- peak to trough amplitude $\sim 18\%$ of the mean.

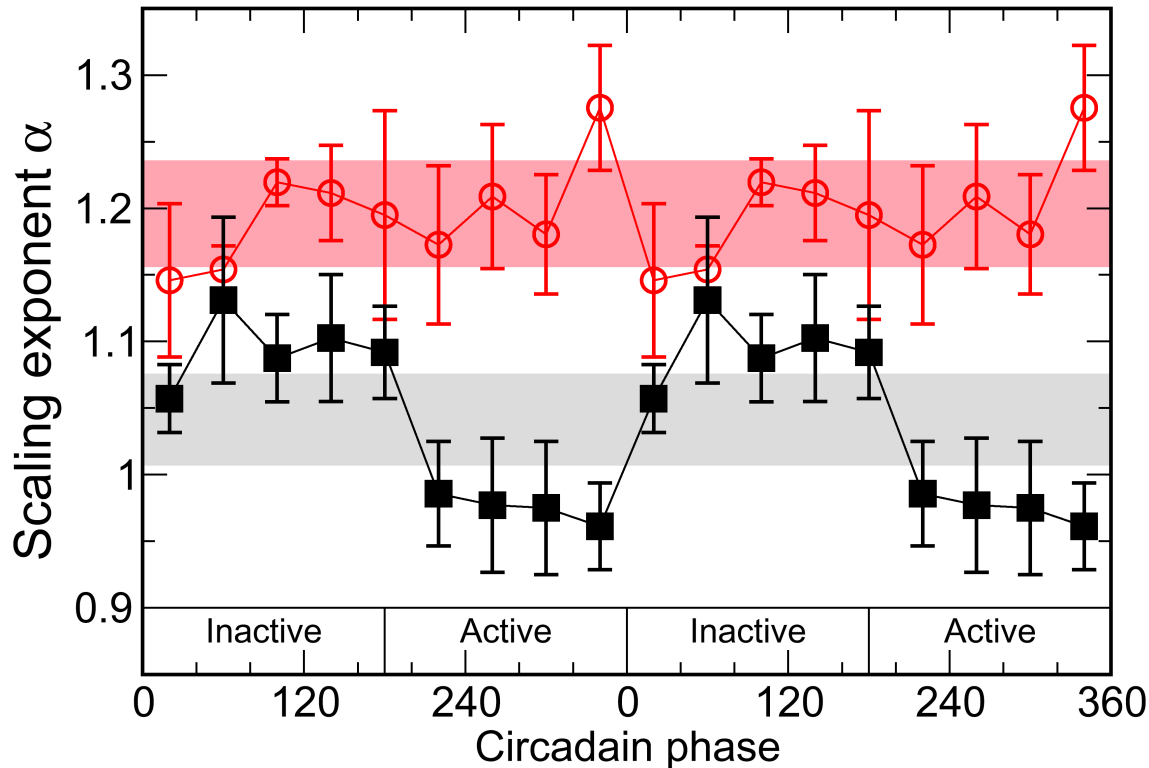
Effect of SCN_x on fractal heart rate fluctuations: abolished circadian rhythm



SCN_x rats showed no circadian rhythm in cardiac fractal correlations ($p > 0.2$).

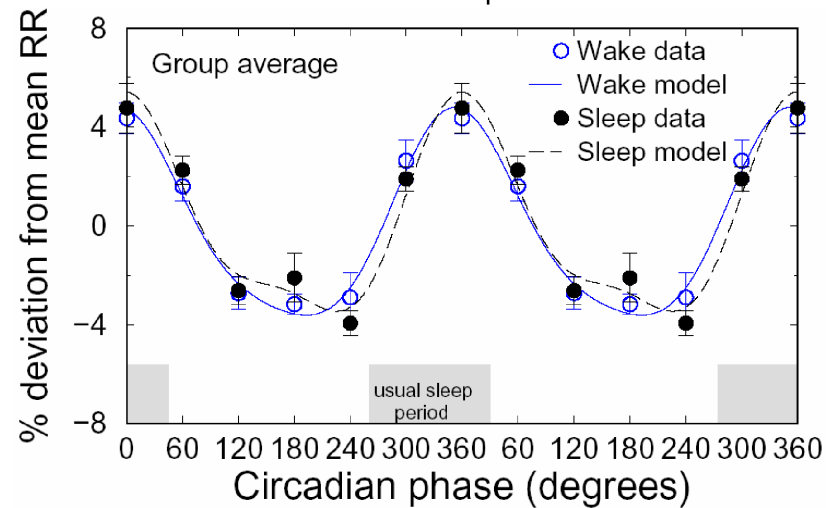
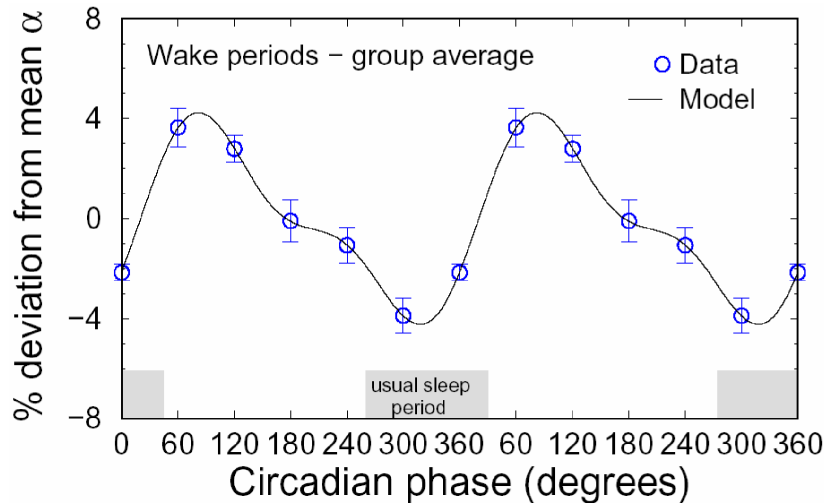
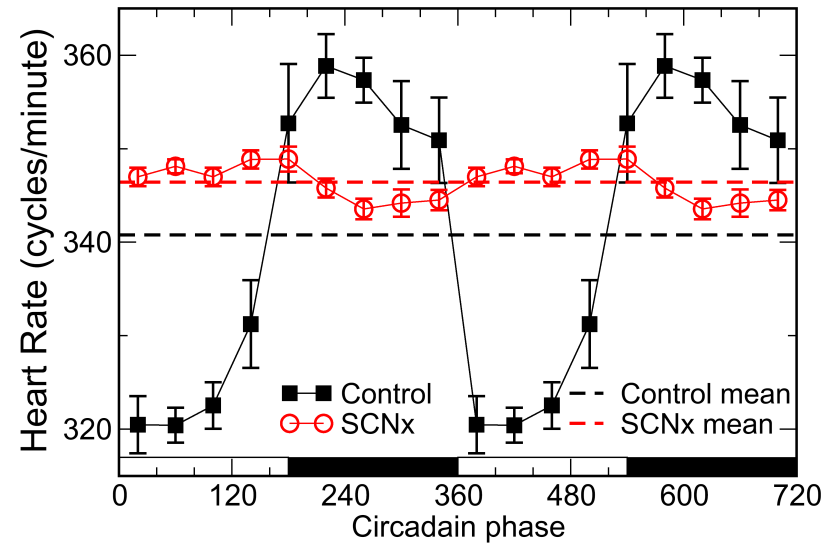
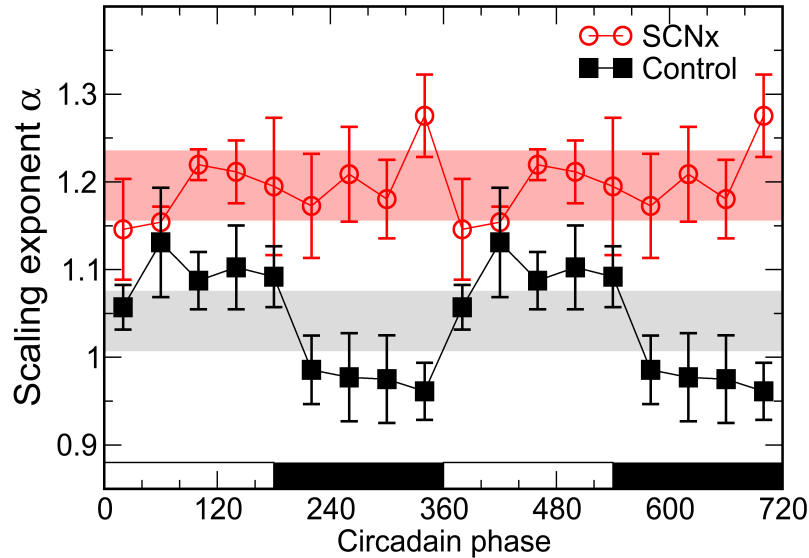
Effect of SCN_x on fractal heart rate fluctuations: increased scaling exponent

DD protocol

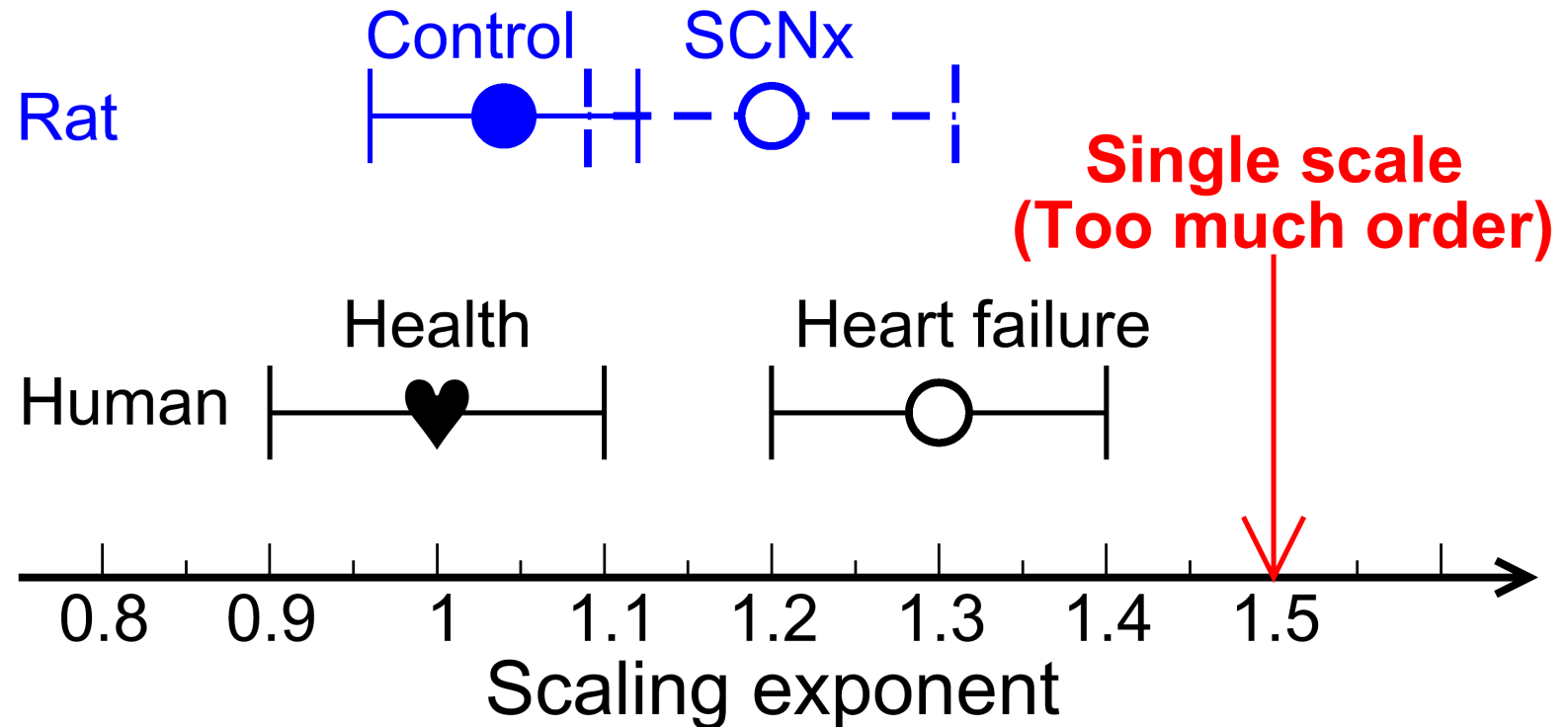


- A significant increases in the scaling exponent in SCN_x: $\alpha=1.2\pm0.04$ (Mean \pm SE) ($p=0.01$).
- Interaction between group and circadian phase ($p=0.04$)

Fractal patterns of heart rate fluctuations are independent of mean heart rate



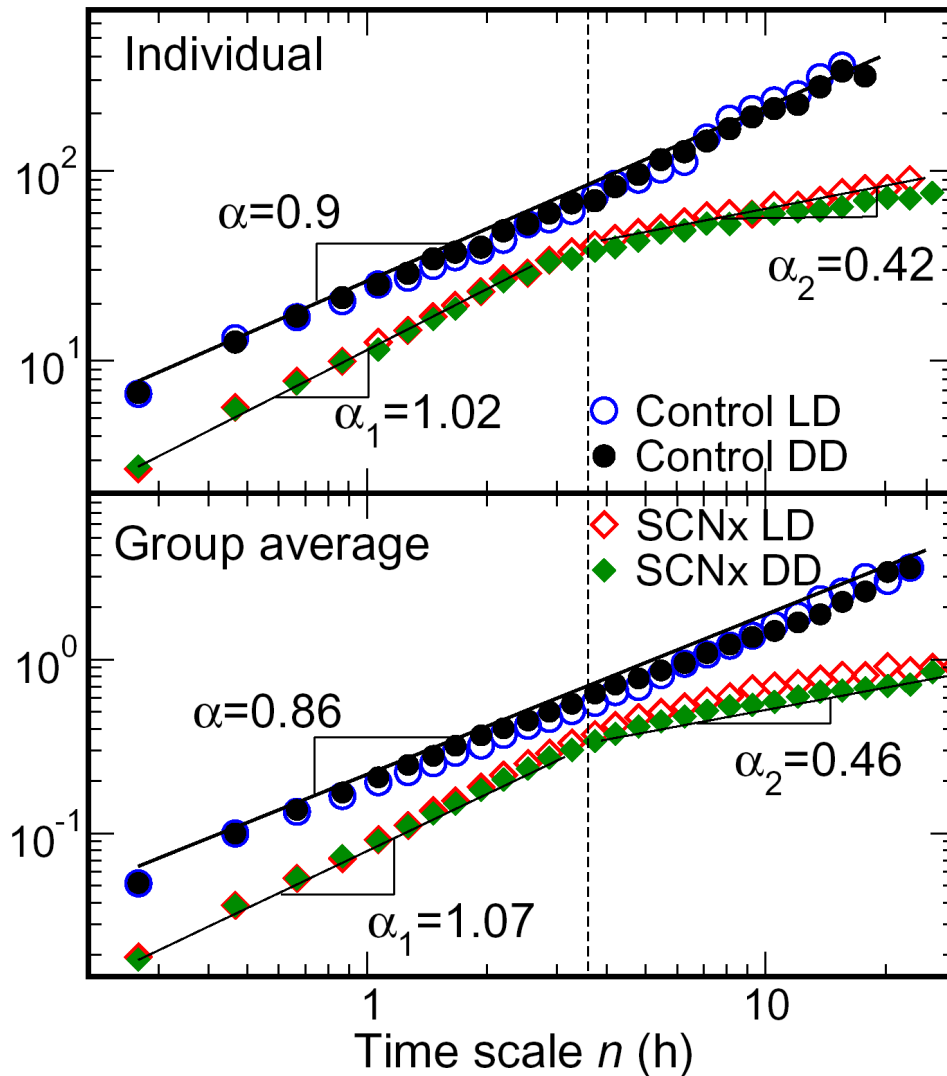
Effect of SCN lesion on fractal heart rate fluctuations



**The above results are focused on the time
scales < 1 hours**

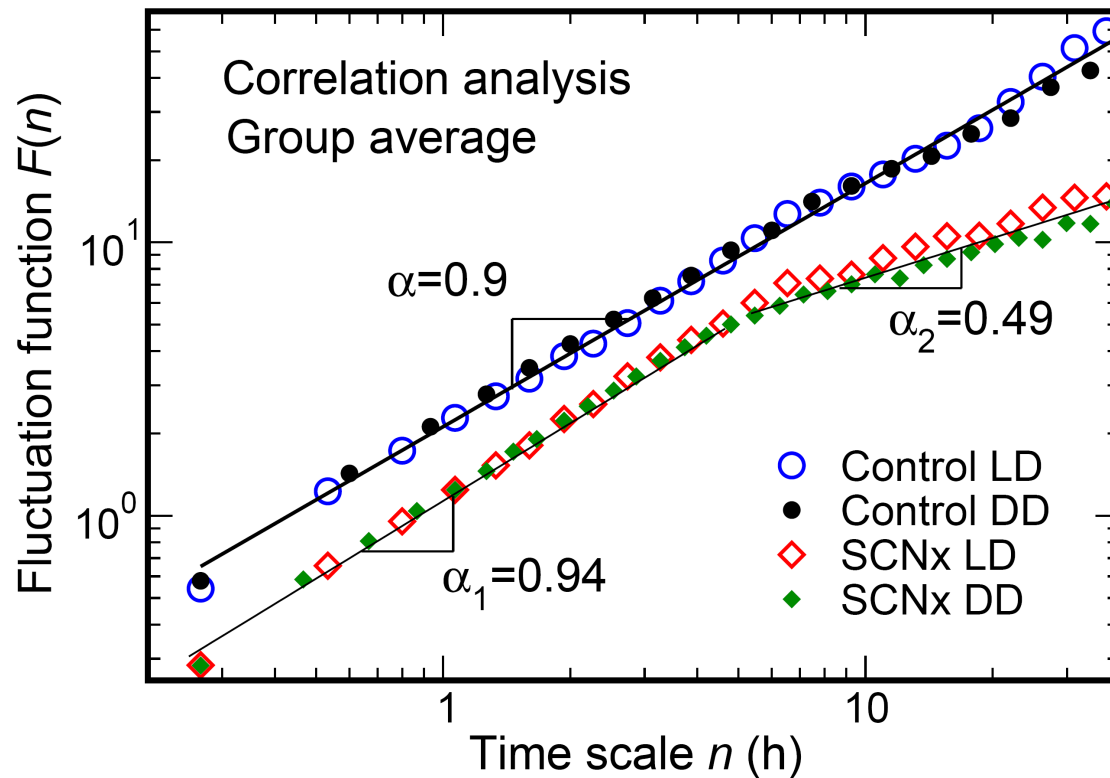
**What about fractal patterns at larger time
scales?**

Effect of SCN-lesion on fractal heart rate correlations at a wide range of time scales



- Correlations completely break down at >4 hours
- Correlations persist but with a significant change at < 4 hours
- Independent of light influences

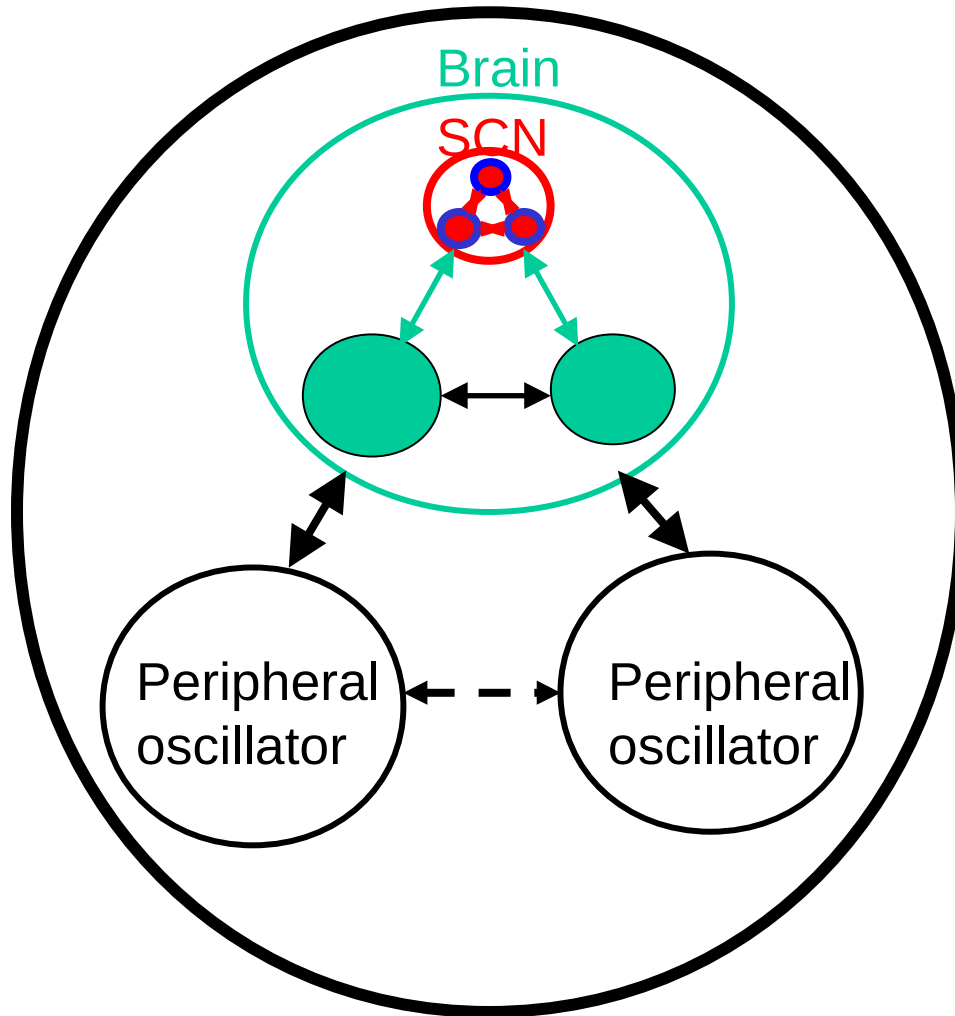
Effect of SCN-lesion on fractal activity correlations at a wide range of time scales



- Scale invariant correlations in control rats from minutes to at least 24 hours
- White-noise behavior at time scales >4 hours in SCN-lesioned rats
- Independent of light influences

Control Mechanisms for Scale-invariant Regulation

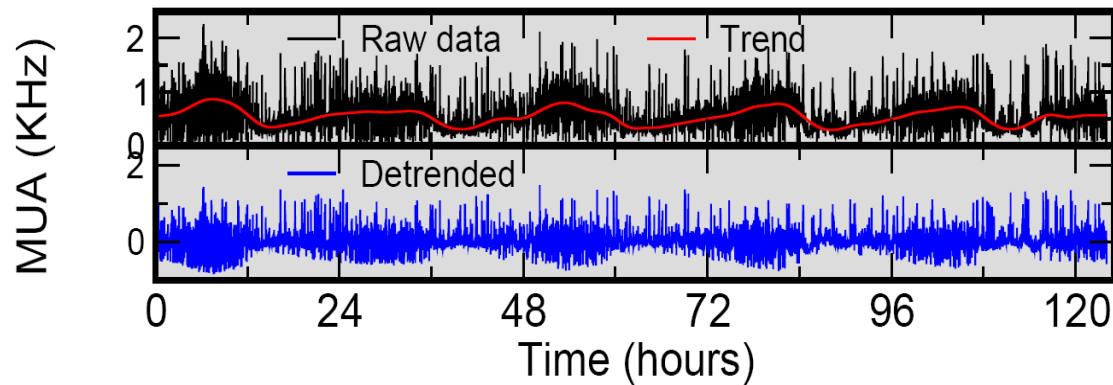
Integrative physiological system



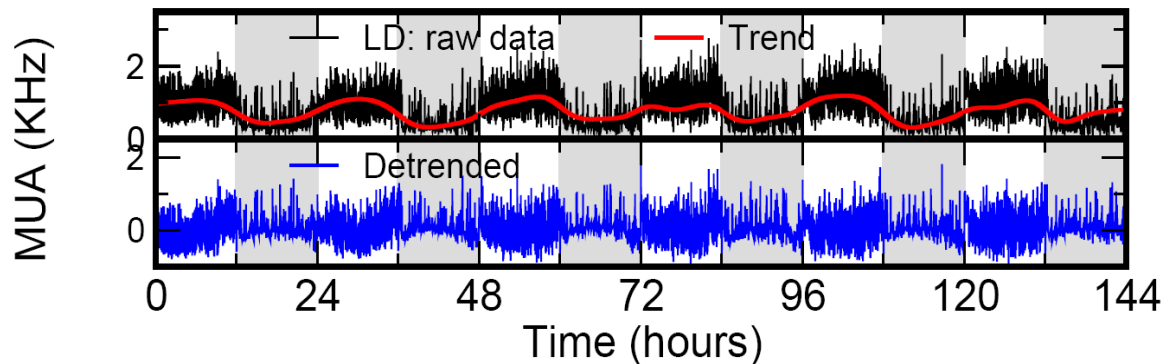
- The SCN is a major node in the network of fractal regulations.

The multi-scale influences of the SCN may be through neuronal interactions within the SCN; or between the SCN and other neuronal nodes

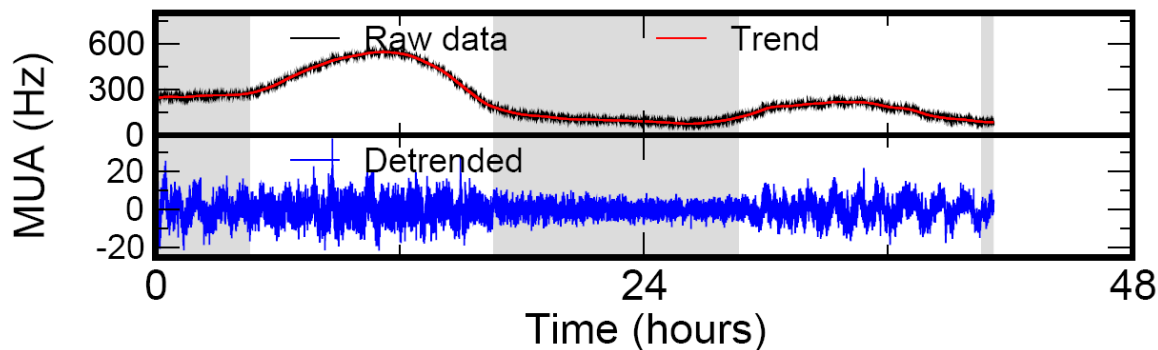
Multiunit activity of the SCN



In vivo:
constant dark

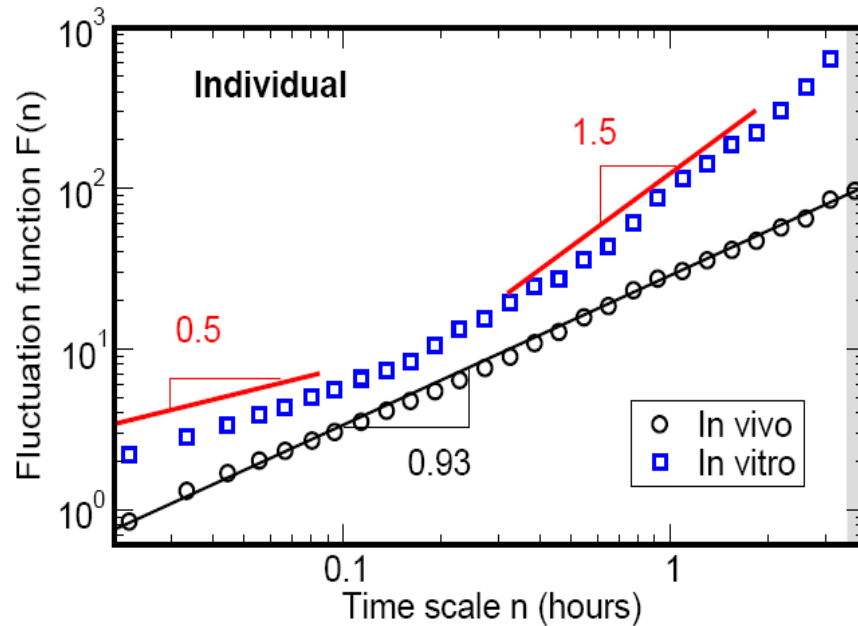


In vivo:
Light-Dark
(12h:12h)

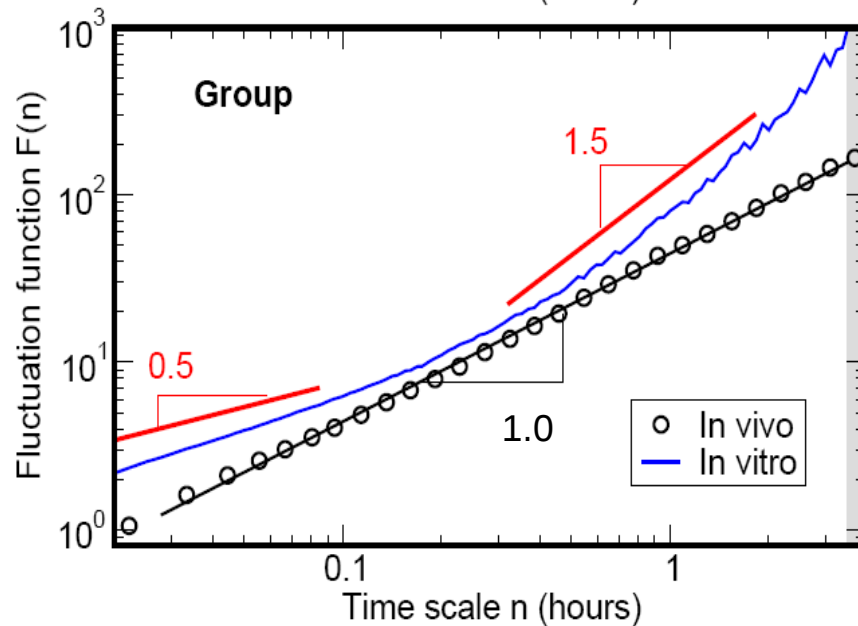


In vitro

Fractal correlations in SCN MUA

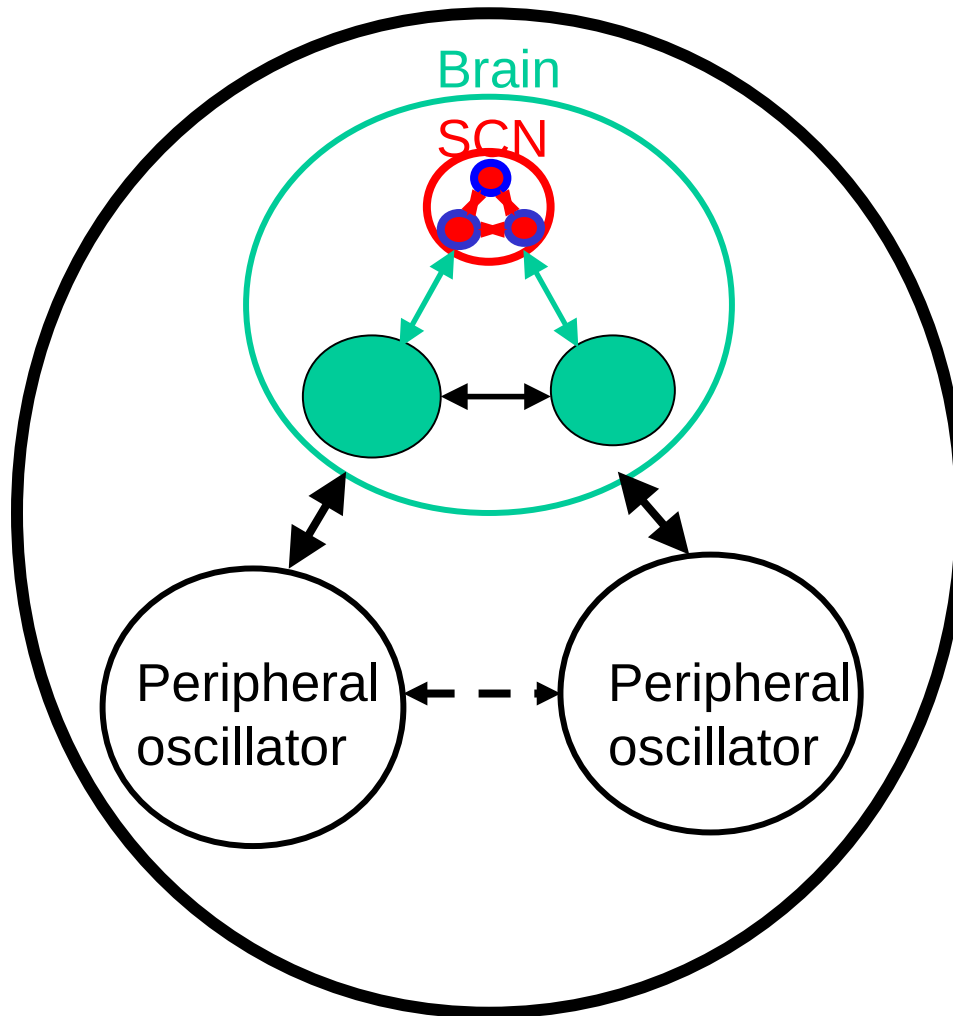


Fractal correlations
in vivo at time scales
from ~ 1 min to 3h



Abolished fractal
patterns in vitro

Control Mechanisms for Scale-invariant Regulation



Summary

- **Healthy cardiac dynamics show fractal regulatory patterns**
- **There is an endogenous circadian rhythm in fractal cardiac control at time scales $< \sim 1$ h**
 - Independent of behavioral and environment influences
 - With larger scaling exponent during biological day and smaller scaling exponent during biological night
 - Similar in diurnal and nocturnal mammals
- **Removing the SCN (the master clock) abolishes the fractal cardiac correlations at large time scales ($> \sim 4$ h)**
 - The SCN functions beyond a simple circadian pacemaker
 - There is a network responsible for the fractal cardiac control and the SCN is one major node in the network
- **The SCN activity also display fractal patterns**
 - Due to the interactions between the SCN and other neuronal nodes

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