



中山大學
SUN YAT-SEN UNIVERSITY

PMI

物理，机器与智能

Order parameters for emergence of consciousness?

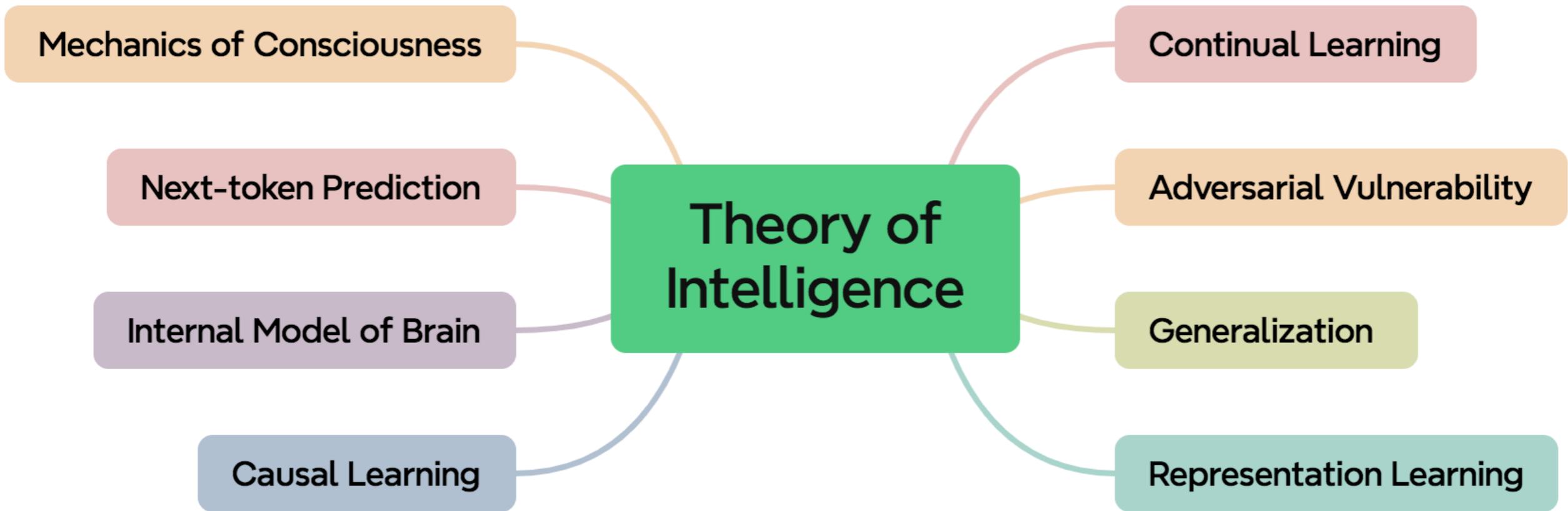
Haiping Huang
Sun Yat-sen University
Apr 28, 2024

**“Rigor of theoretical physics provides guidelines
and testable predictions for realistic applications.”
—Giulio Biroli et.al**

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Rigor helps Clarity



Presented with **xmind**

<https://arxiv.org/abs/2306.11232>

Challenges in understanding consciousness

When we perceive, think and act, there is a whirl of causation and information processing There is also an internal aspect; there is something it **feels** like to be a cognitive agent. This internal aspect is conscious experience.

Chalmers, D. J. [1997] *The Conscious Mind: In Search of a Fundamental Theory*



Subjective experience relates to feelings associated with stimuli

Arxiv: 2401.13690

“大脑用行动来检验其假设”

Buzsáki, *The brain from inside out*

Science typically studies the third person aspect, you know, I can take a brain of a mouse and poke it, put it in the scanner and record from the individual neurons, etc. But consciousness is about how the system **feels from the inside**. So that's always been the big, the big challenge.

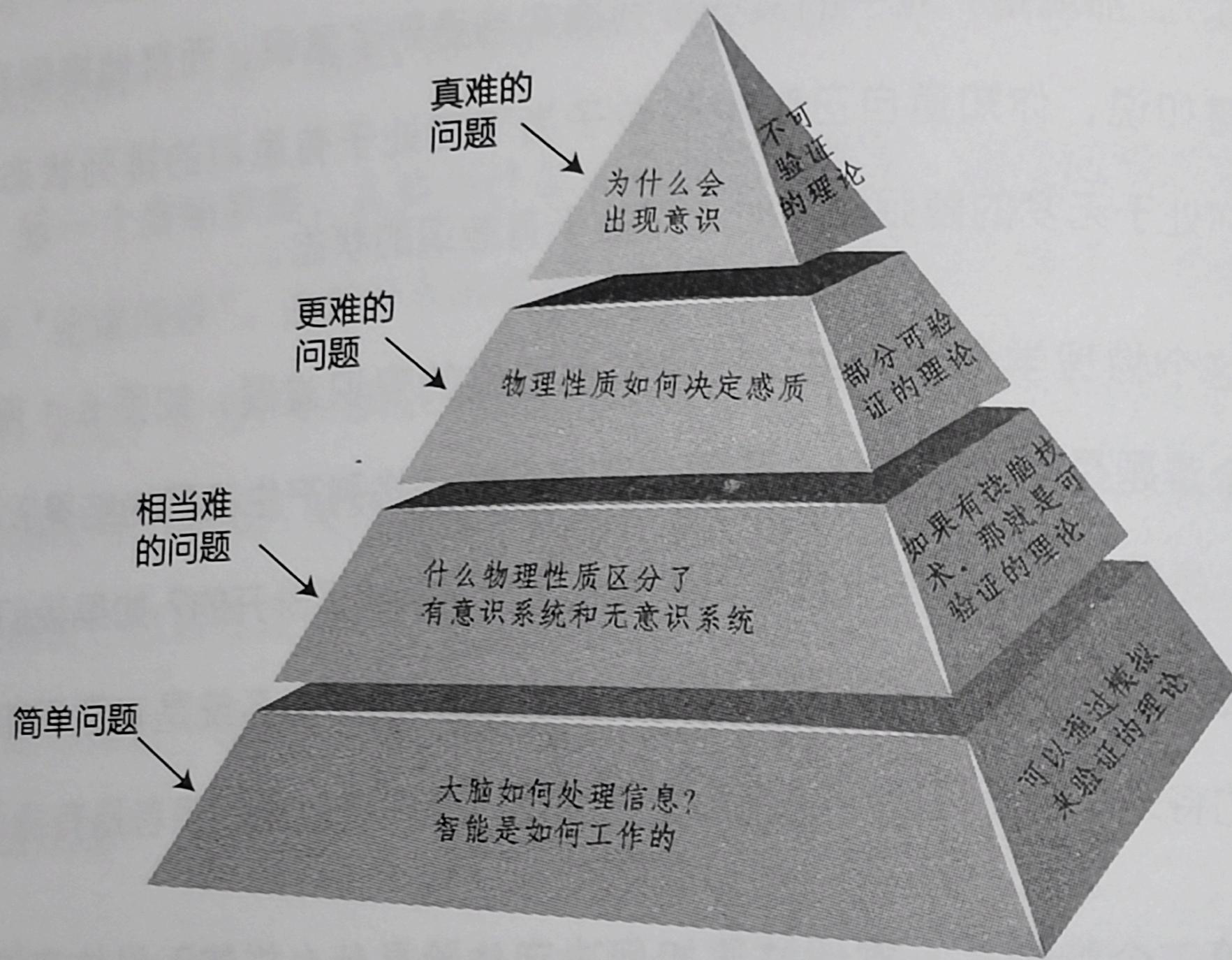
Consciousness, which we can define as experience -- your experience both of yourself and of the world around you -- is one of the hardest to pin down



Christof Koch

How neural interactions support conscious experience?

deep sleep, anesthesia, death



信息处理：
 有意识（高级）
 （**注意力**）
 无意识（简单）

某种对称性破缺？

图 8-1 三个彼此独立的意识难题

Life 3.0

注：对心智的理解涉及几个层次的问题。大卫·查尔默斯所谓的“简单问题”可以不提到主观体验。一些但不是全部物理系统是有意识的，这个事实提出了三个不同的问题。如果有一个理论可以回答“相当难的问题”，那它就可以用实验来检验。如果检验成功的话，我们就可以以它为基础来解决上层那些更棘手的问题。

Consciousness is supported by near-critical slow cortical electrodynamics

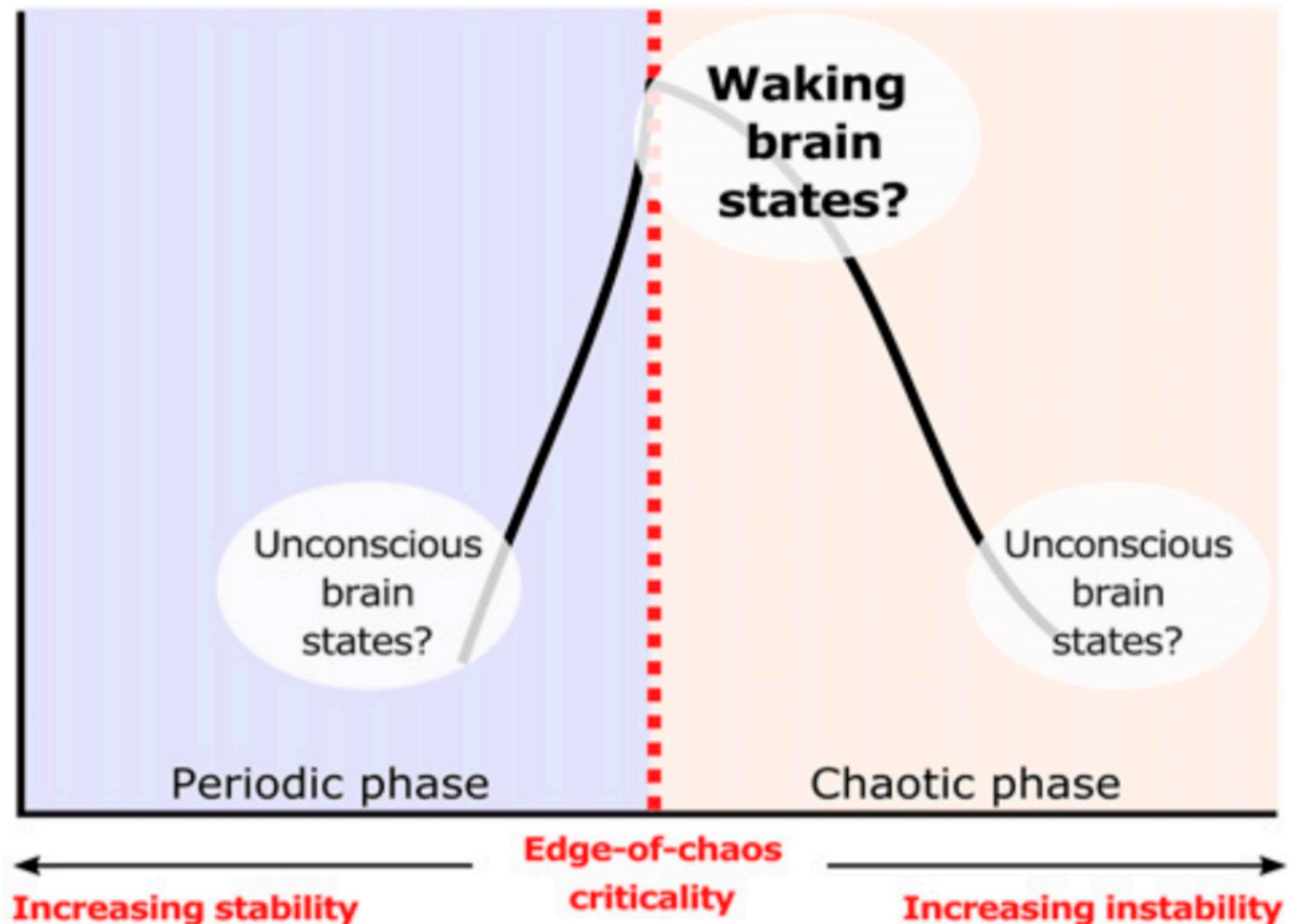
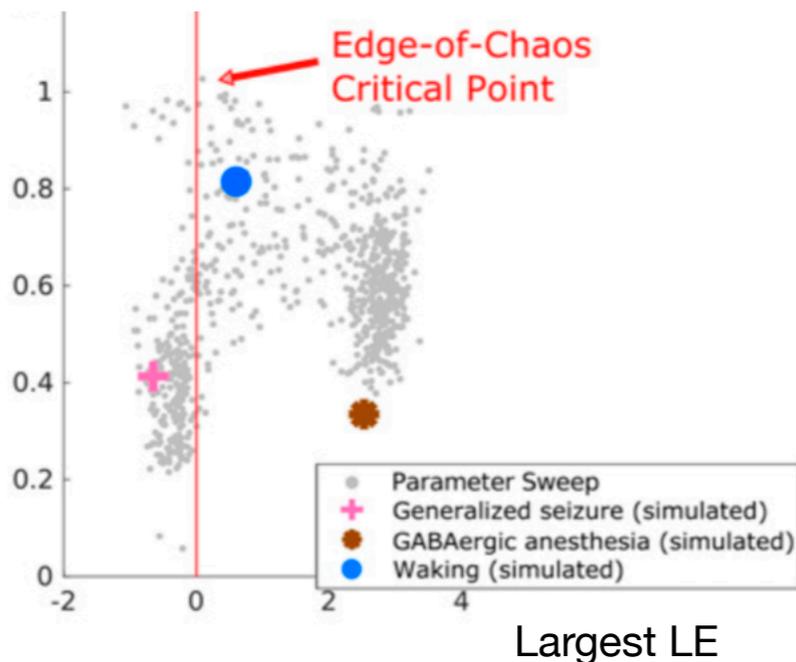
PNAS, 2022

Daniel Toker^{a,1}, Ioannis Pappas^{b,c,d}, Janna D. Lendner^{b,e}, Joel Frohlich^a, Diego M. Mateos^{f,g,h}, Suresh Muthukumaraswamyⁱ, Robin Carhart-Harris^{j,k}, Michelle Paff^l, Paul M. Vespa^m, Martin M. Monti^{a,m}, Friedrich T. Sommer^{b,n}, Robert T. Knight^{b,c}, and Mark D'Esposito^{b,c}

Lempel-Ziv Complexity

The amount of non-redundant information in a signal

Information-richness of cortical electrodynamics



Metaphor (toy model) of brain dynamics

Synaptic currents

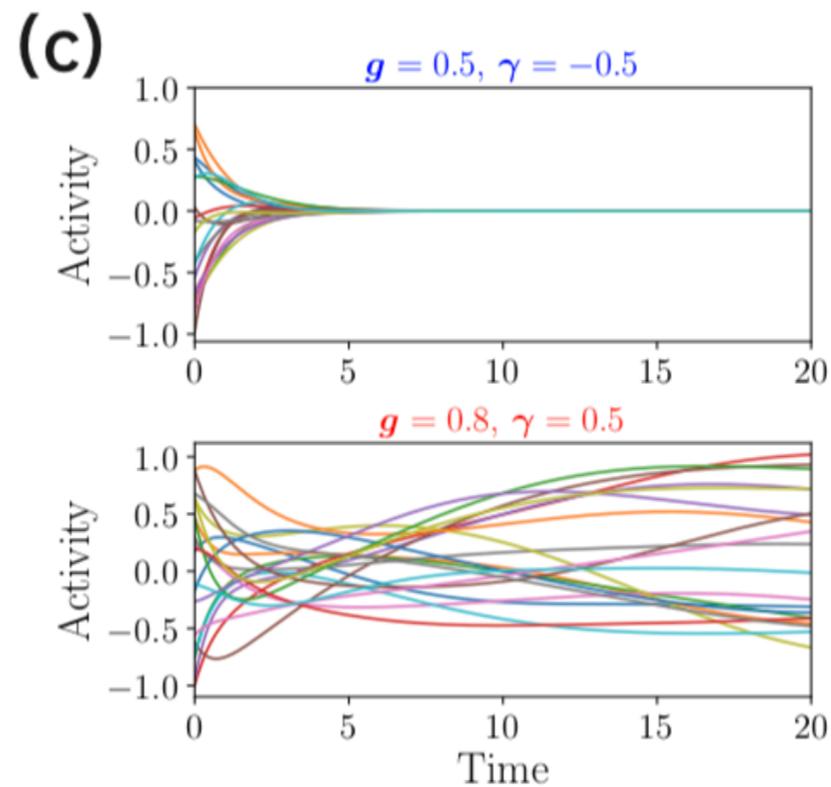
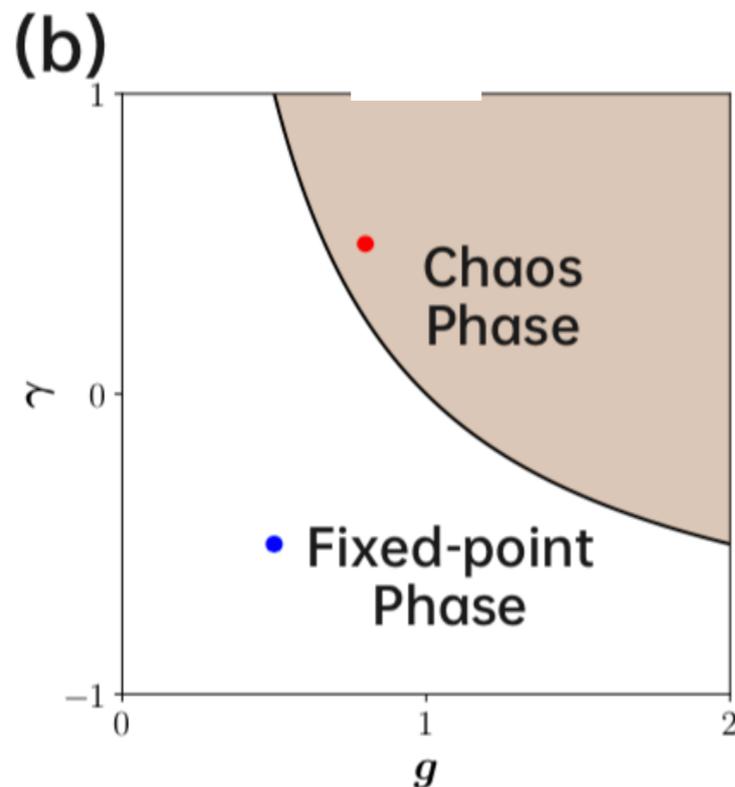
$$\frac{dx_i}{dt} = -x_i + \sum_{j=1}^N J_{ij} \phi(x_j),$$

Connectivity

$$\langle (J_{ij})^2 \rangle = \frac{g^2}{N},$$
$$\langle J_{ij} J_{ji} \rangle = \frac{g^2}{N} \gamma,$$

Characteristics:

Asymmetric, correlated synapses, non-linear dynamics [Non-gradient dynamics]



Equilibrium limit

$$\frac{dx_i(t)}{dt} = -x_i(t) + g \sum_{j=1}^N J_{ij} x_j(t) + \sigma \xi_i(t), \quad J_{ij} = J_{ji}$$

$$\mathcal{H}(\mathbf{x}) = -\frac{1}{2} \sum_i x_i^2 - \frac{1}{2} g \sum_{i \neq j} J_{ij} x_i x_j, \quad F = -\nabla_x \mathcal{H}(\mathbf{x})$$

Symmetric, Linear dynamics:

$$P(\mathbf{x}) \sim \exp\left(-\frac{\mathcal{H}(\mathbf{x})}{T}\right), \quad T = \sigma^2 / 2$$

Lyapunov Function decreasing over dynamics

More complex cases

$$\frac{d\mathbf{x}}{dt} = F(\mathbf{x}) + \zeta$$

Probability conservation

$$\frac{\partial P}{\partial t} + \nabla \cdot \mathbf{J} = 0,$$

Steady state: $P_{ss}(x) = e^{-U(\mathbf{x})}$ \longrightarrow $\mathbf{J}_{ss} = \mathbf{F}P_{ss} - \nabla \cdot (\mathbf{D}P_{ss})$

Non-equilibrium potential

Curl force

$$\mathbf{F} = \mathbf{J}_{ss}/P_{ss} - \mathbf{D}\nabla U + \nabla \cdot \mathbf{D}$$

Divergent free flux: $\nabla \cdot \mathbf{J}_{ss} = 0$

Diffusion matrix of stochastic noise
e.g., background activity

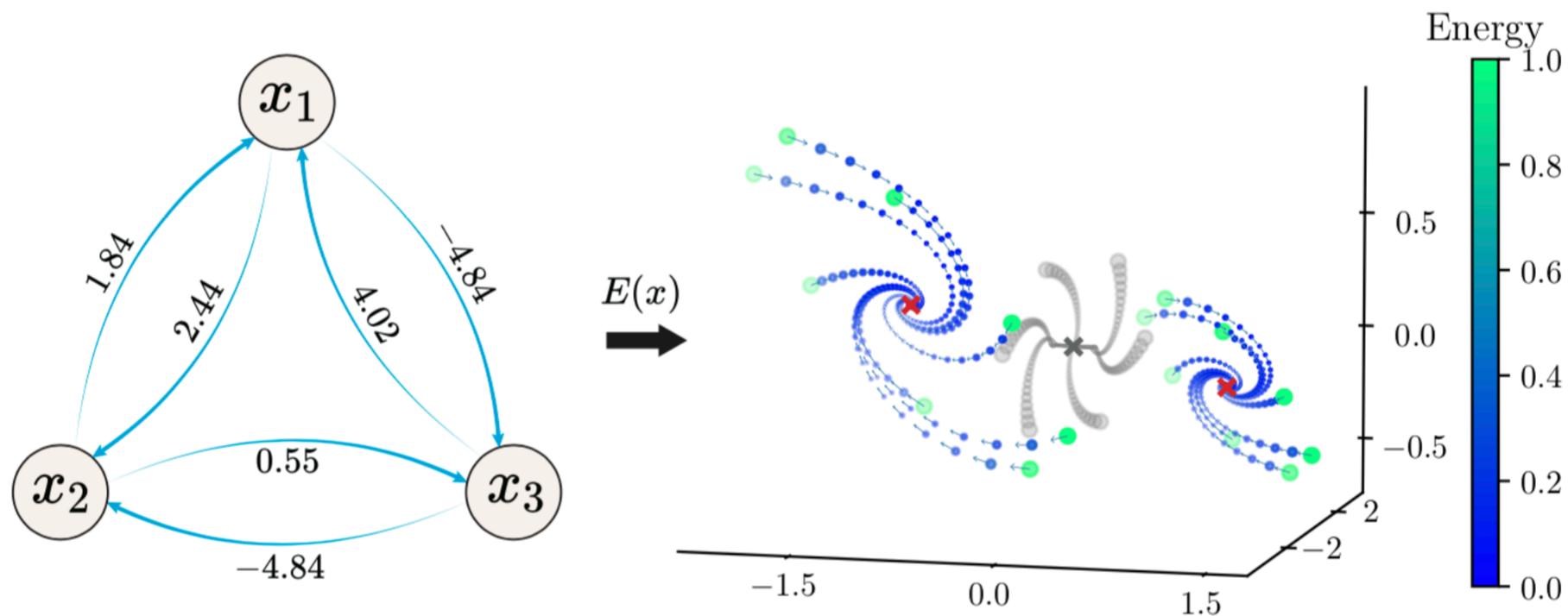
But: hard to get $U(x)$, and steady state probability generally unknown

Intuitive illustration

The phase space is partitioned into different regions of different speeds

(quasi-potential) of the dynamics,

$$E(\mathbf{x}) = \frac{1}{2} \sum_i \left(-x_i + \sum_{j=1}^N J_{ij} \phi(x_j) \right)^2 + \eta \|\mathbf{x}\|^2, \quad (4)$$



Rationale underlying the intuition

According to Eq. (4), we can write down the following stochastic gradient dynamics (or Langevin dynamics)

$$\frac{d\mathbf{x}}{dt} = -\nabla_{\mathbf{x}}E(\mathbf{x}) + \sqrt{2T}\boldsymbol{\epsilon}, \quad (5)$$

where $\boldsymbol{\epsilon}(t)$ is a time-dependent white noise, whose statistics is given by $\langle \epsilon_i(t) \rangle = 0$, $\langle \epsilon_i(t)\epsilon_j(t') \rangle = \delta_{ij}\delta(t-t')$. The temperature T is used to tune the energy level, playing the same role as in the equilibrium Boltzmann measure. We can write the gradient (force) in Eq. (5) in the component wise,

$$F_i \equiv -x_i + h_i - \phi'(x_i) \sum_{j:j \neq i} J_{ji}(h_j - x_j), \quad (6)$$

Impact on neighbors

The force is not a gradient of a potential, but approaches the true dynamics in the steady state limit!

Therefore:

!!We can formulate the steady state behavior as the canonical ensemble of stationary fixed points!!

$$P(\mathbf{x}) = \frac{1}{Z} e^{-\beta E(\mathbf{x})},$$

Order parameters for non-gradient dynamics

Free energy for non-equilibrium steady state/NESS

$$-\beta f = \lim_{n \rightarrow 0} \frac{\ln \langle Z^n \rangle}{Nn}$$

Quenched disorder over J

$$Z^n = \int d\mathbf{x} e^{-\beta \left(\frac{1}{2} \sum_{ia} \left(-x_i^a + \sum_j J_{ij} \phi(x_j^a) \right)^2 + \eta \sum_a \|\mathbf{x}^a\|^2 \right)}$$

Constraining norm of activity

Replicated dynamics states

Order parameters

$$Q^{ab} = \frac{1}{N} \sum_i \phi(x_i^a) \phi(x_i^b),$$

Activity level of the dynamic system

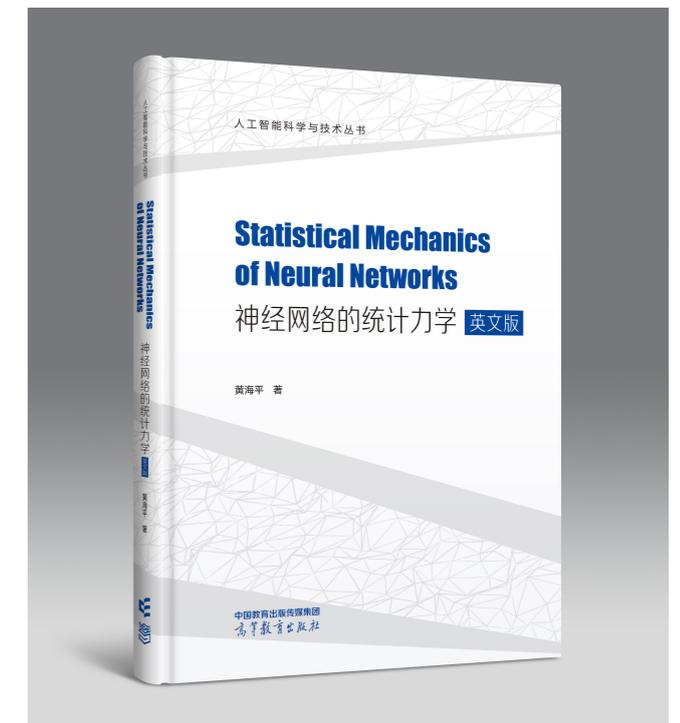
$$R^{ab} = \frac{1}{N} \sum_i \hat{x}_i^a \phi(x_i^b),$$

Response property of the dynamic system

A response field inspired by DMFT [2305.08459]

Technical note:

Q and R emerges naturally from disorder average!



Theory summary

$$-\beta f = \frac{1}{2}Q\hat{Q} - q\hat{q} + R\hat{R} - r\hat{r} - \ln \sigma + \frac{1}{2}\beta g^2\gamma(r^2 - R^2) + \int (DuDv) \ln I, \quad (11)$$

where $\sigma \equiv \sqrt{1 + g^2\beta(q - Q)}$, and \hat{Q} , \hat{q} , \hat{R} and \hat{r} are conjugated order parameters. The integral $I \equiv \int dx e^{\mathcal{H}(x)}$, where

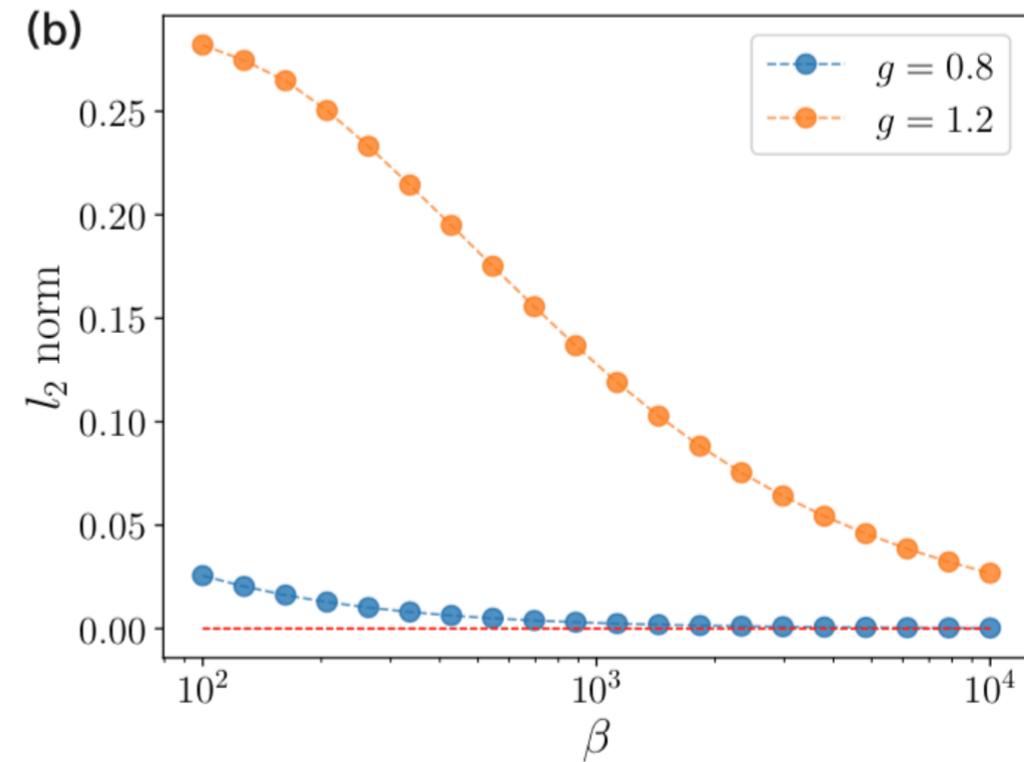
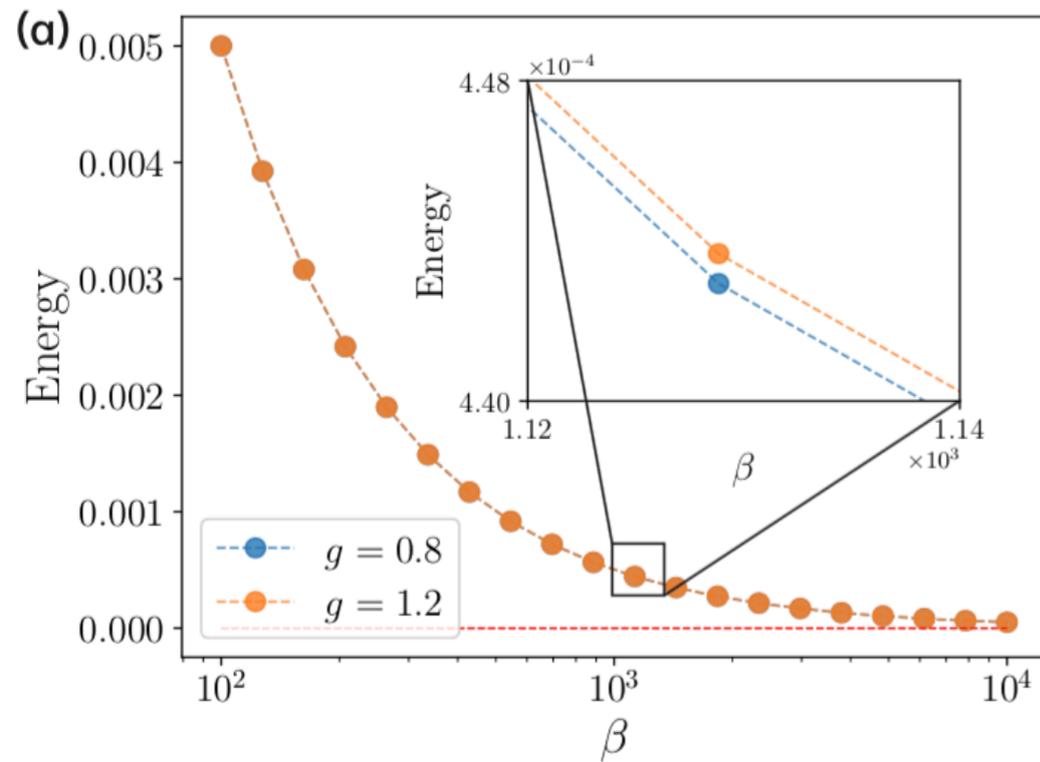
$$\begin{aligned} q &= [\langle \phi^2 \rangle], \\ Q &= [\langle \phi \rangle^2], \\ \hat{q} &= -\frac{gk}{2} + \frac{g^2k^2Q}{2} + \frac{gk}{2\sigma^2}(\hat{r} - \hat{R})f(1, 1, -2)[\langle \phi^2 \rangle] \\ &\quad + \frac{gk}{\sigma^2}(\hat{r} - \hat{R})f(0, -1, 1)[\langle \phi \rangle^2] + \frac{k^2}{2}(1 - 2gkQ)[\langle x^2 \rangle] \\ &\quad + \frac{gk\sqrt{\beta}}{\sigma^2}f(0, 1, -2)[\langle x \rangle \langle \phi \rangle] + \frac{gk\sqrt{\beta}}{\sigma^2}f(-1, 0, 2)[\langle x\phi \rangle] \\ &\quad + gk^3Q[\langle x \rangle^2], \\ \hat{Q} &= g^2k^2Q + \frac{2gk}{\sigma^2}(\hat{r} - \hat{R})f(0, 1, -1)[\langle \phi^2 \rangle] \\ &\quad + \frac{gk}{\sigma^2}(\hat{r} - \hat{R})f(1, -3, 2)[\langle \phi \rangle^2] - 2gk^3Q[\langle x^2 \rangle] \\ &\quad - \frac{2gk\sqrt{\beta}}{\sigma^2}f(1, -2, 2)[\langle x \rangle \langle \phi \rangle] + \frac{2gk\sqrt{\beta}}{\sigma^2}f(0, -1, 2)[\langle x\phi \rangle] \\ &\quad + k^2(1 + 2gkQ)[\langle x \rangle^2], \\ r &= -\frac{1}{\sigma^2}f(1, 0, -1)[\langle \phi^2 \rangle] + \frac{1}{\sigma^2}f(0, 1, -1)[\langle \phi \rangle^2] \\ &\quad + \frac{\sqrt{\beta}}{\sigma^2}(1 - gkQ)[\langle \phi x \rangle] + \frac{\sqrt{\beta}}{\sigma^2}gkQ[\langle \phi \rangle \langle x \rangle], \\ R &= -\frac{1}{\sigma^2}f(0, 1, -1)[\langle \phi^2 \rangle] - \frac{1}{\sigma^2}f(1, -2, 1)[\langle \phi \rangle^2] \\ &\quad - \frac{\sqrt{\beta}}{\sigma^2}gkQ[\langle x\phi \rangle] + \frac{\sqrt{\beta}}{\sigma^2}(1 + gkQ)[\langle x \rangle \langle \phi \rangle], \\ \hat{r} &= \beta g^2\gamma r, \\ \hat{R} &= \beta g^2\gamma R, \end{aligned} \quad \begin{aligned} \mathcal{H}(x) &\equiv -\beta\eta x^2 + \frac{1}{2}(2\hat{q} - \hat{Q})\phi^2(x) \\ &\quad + \left(\sqrt{\frac{g^2\beta Q\hat{Q} - \hat{R}^2}{g^2\beta Q}}u + \frac{\hat{R}}{g\sqrt{\beta Q}}v \right)\phi(x) \\ &\quad - \frac{1}{2\sigma^2} \left(g\sqrt{\beta Q}v + (\hat{r} - \hat{R})\phi(x) - \sqrt{\beta}x \right)^2. \end{aligned} \quad (12)$$

Single variable effective Hamiltonian

Replica symmetry Ansatz:

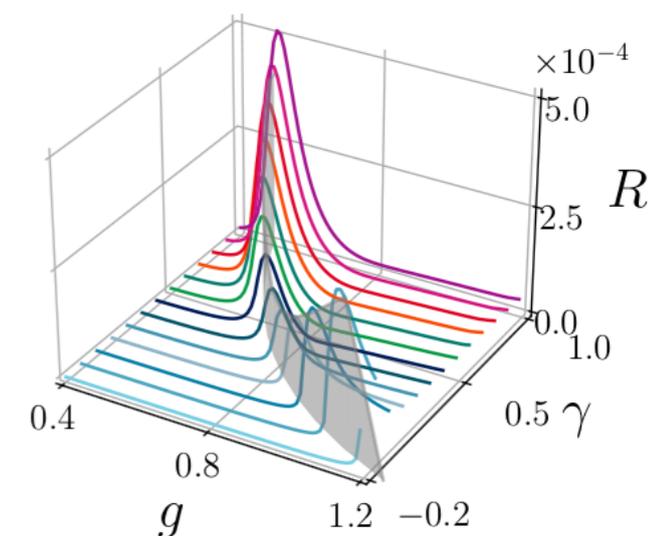
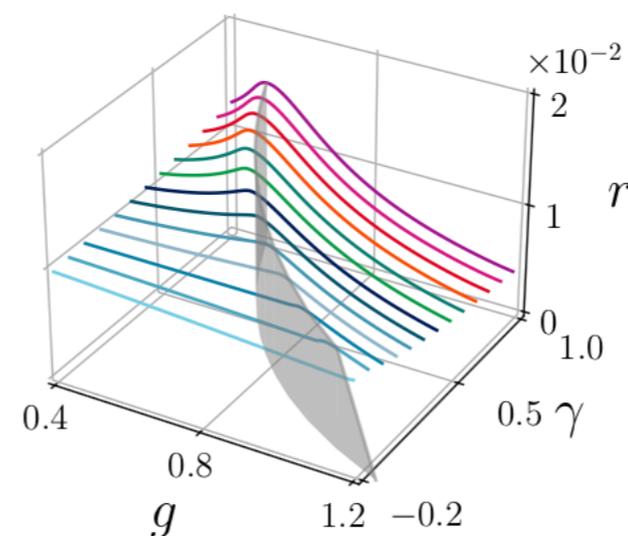
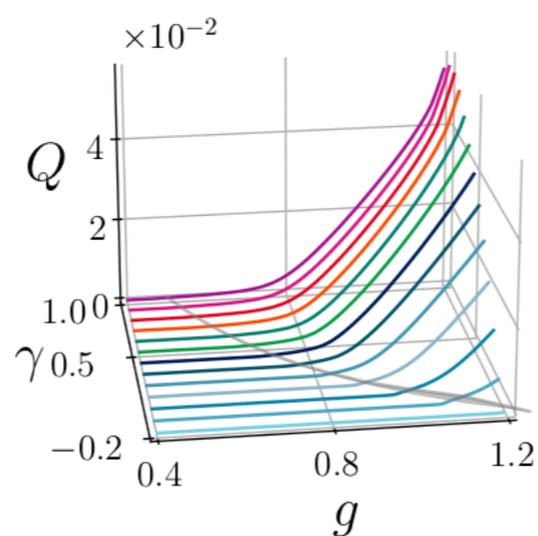
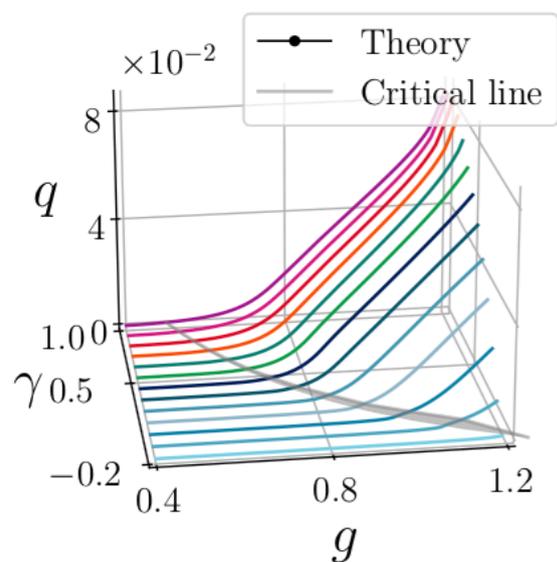
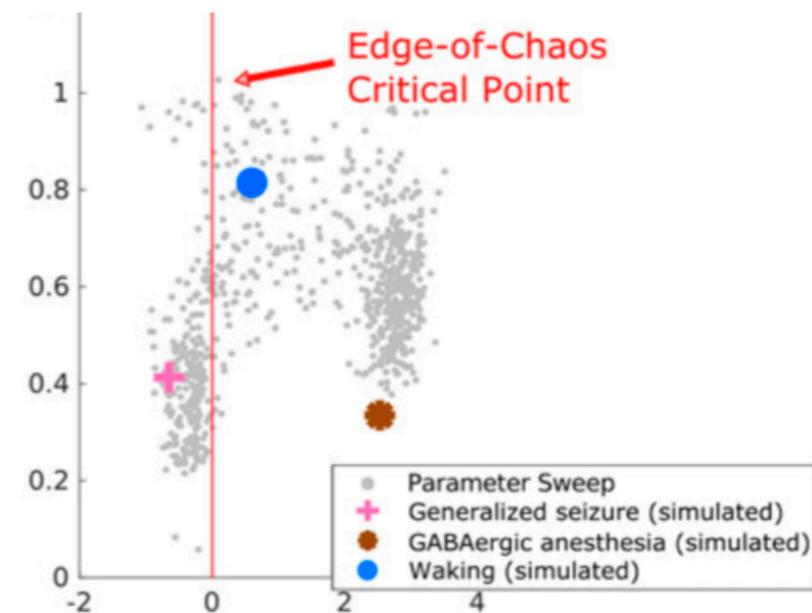
$$Q^{ab} = q\delta_{ab} + Q(1 - \delta_{ab}) \quad \text{and} \quad R^{ab} = r\delta_{ab} + R(1 - \delta_{ab})$$

Result I: energy



Result II: order parameters

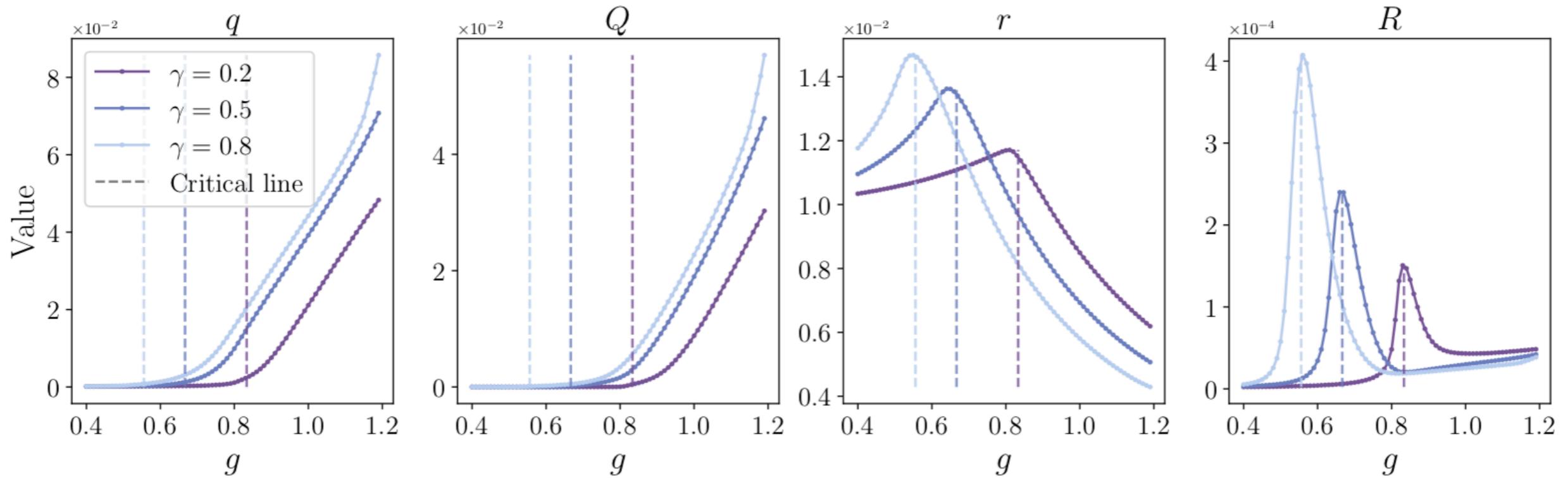
Continuous transition to chaos!



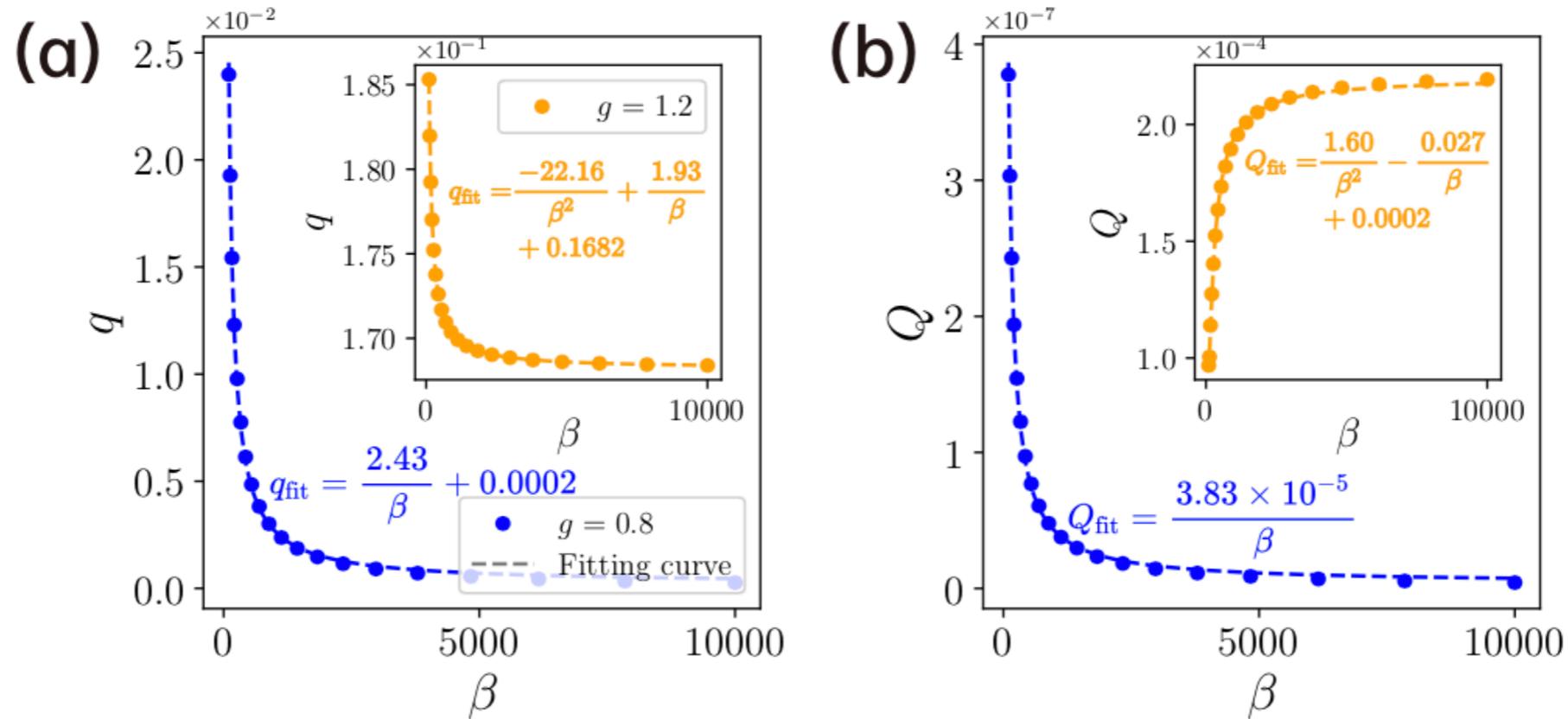
Response R & r peaked at the transition point

!!The steady dynamics is more responsive at the edge of chaos!!

Result II: order parameters



Result III: scaling behavior



$$q - Q \propto \beta^{-1}$$

$$R - r \propto \beta^{-1/2}$$

$$r \propto \beta^{1/2}$$