



INTRODUCTION

Can we determine the complex topology based on limited resolution measurements?

- We address this problem in the context of Magnetic Resonance (MR) measurements of porous structure such as brain tissue
- Dynamical model describing magnetization transfer in a network of pores
- The model is analyzed on a **general network** in the **spectral domain**
- The model can also describe population dynamics, agents problems in economy or directed polymers in random media
- It can also be interpreted as a Kardar–Parisi–Zhang (KPZ) equation on a graph¹

THE MODEL

Magnetization transfer in a pore network is modelled as a Stratonovich Stochastic Differential Equations (SDEs) living on a graph under a dynamical random environment. The equations are as follows:

$$\frac{dm_i(t)}{dt} = J \sum_{j \in \mathcal{G}} L_{ij} m_j(t) + g_i(t) m_i(t),$$

 $m_i(0) = m_0$, for all *i*,

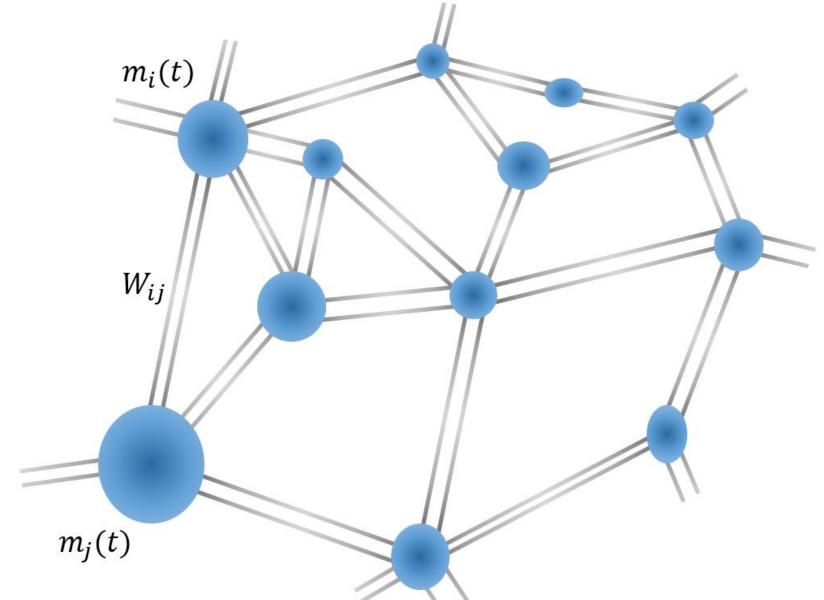
L- the graph Laplacian matrix

- the interaction strength between pores due to diffusion of molecules between pores.

 $g_i(t)$ - a multiplicative white noise,

$$\langle g_i(t) \rangle = 0$$
, and $\langle g_i(t)g_j(t') \rangle = \sigma^2 \delta_{ij} \delta(t-t')$.

 σ^2 - the change in magnetization due to the diffusion of molecules within the pore, affected by temperature, pore geometry, and the magnetic field applied on the pore.



- Analysis of the Mean field topology shows a Pareto power-law steady state distribution and the existence of a condensation/localization phase transition²
- Lattice topology is known as the time dependent Parabolic Anderson Model (PAM)³
- Tree topology was also studied⁴

Phase Transitions in Stochastic Diffusion on a General Network

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- Spectral generalization of the Janssen-De Dominicis technique⁵
- Perturbative calculation in σ^2 to all orders in the spectral domain of the graph reveals the existence of a localization phase transition using renormalization group arguments
- The existence of the phase transition is determined by what we define as the collision matrix,

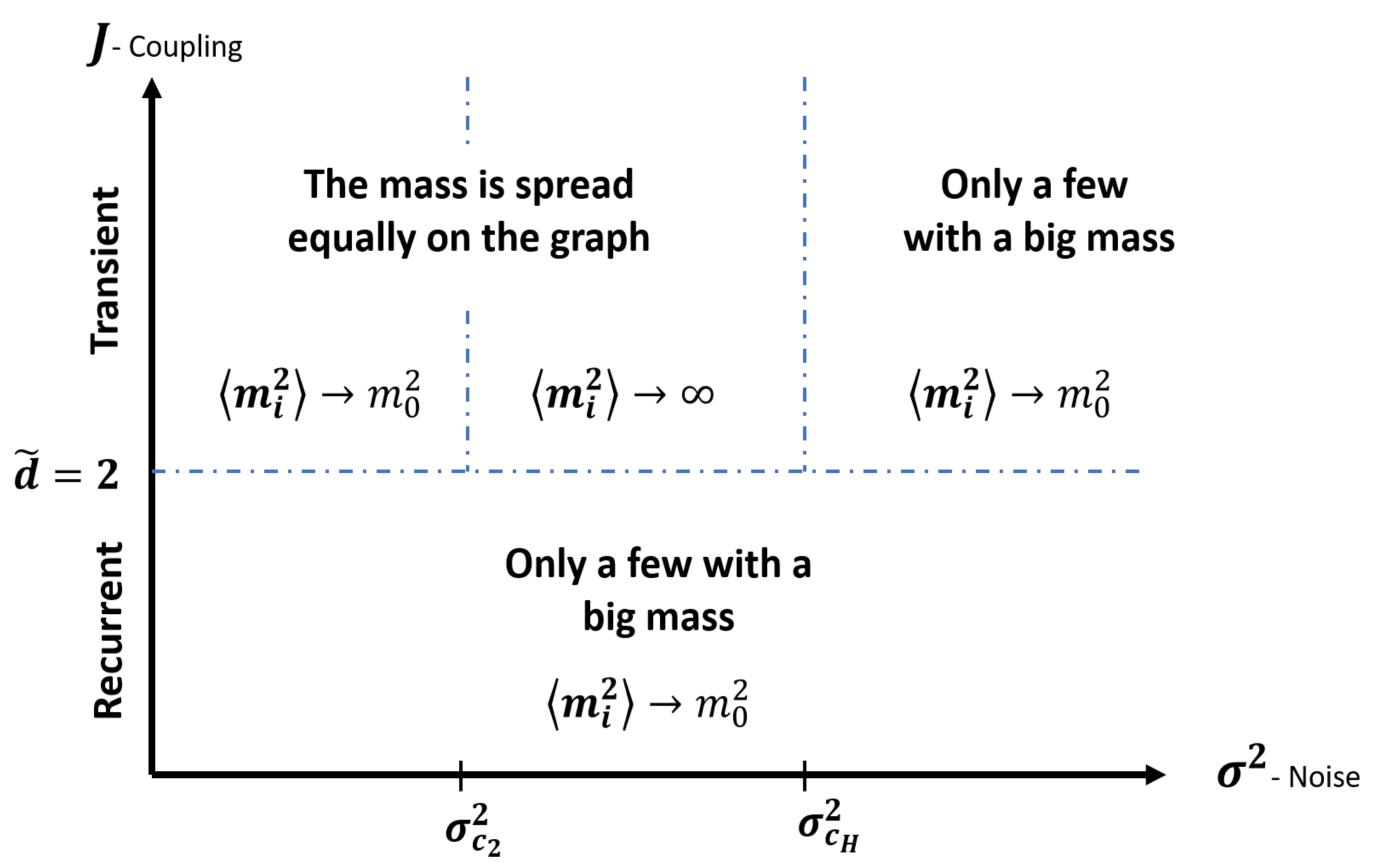
$$I_{ij}(t) = P$$

 $P_{ii}(t)$ - the transition probability from site *i* to *j* at time *t* on the graph,

• The long time limit of this matrix is directly connected to **the probability to return** to the origin and to the spectral dimension (in the symmetric case):

$$\lim_{t \to \infty} \frac{\ln(\sum_k I_{ik}(t))}{\ln(t)} = \lim_{t \to \infty}$$

- Measures of the complexity of the network
- Indicates to what extent the network is recurrent vs. transient
- ✤ A measurable parameter
- In the transient case, $\tilde{d} > 2$, there exists a localization phase transition



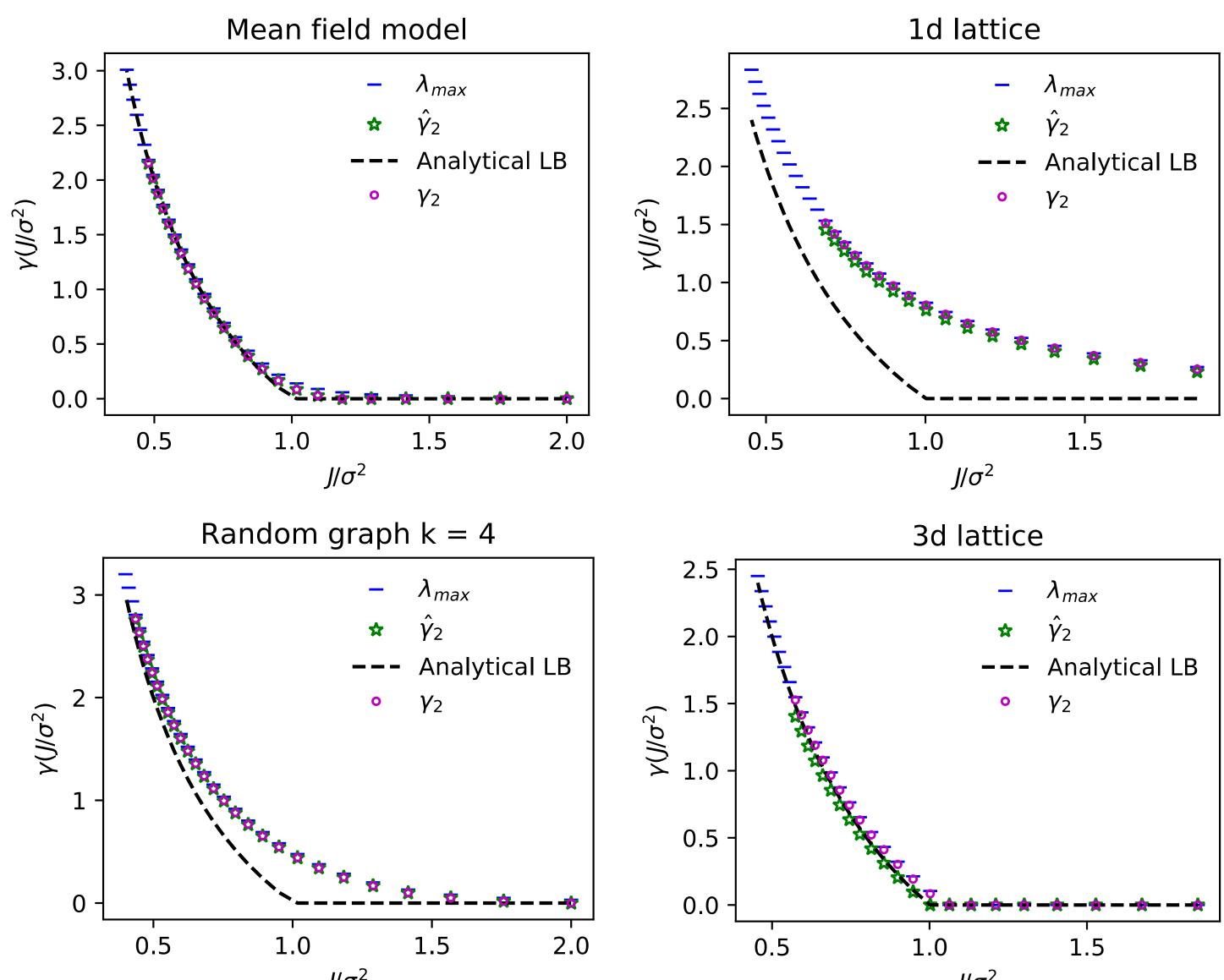
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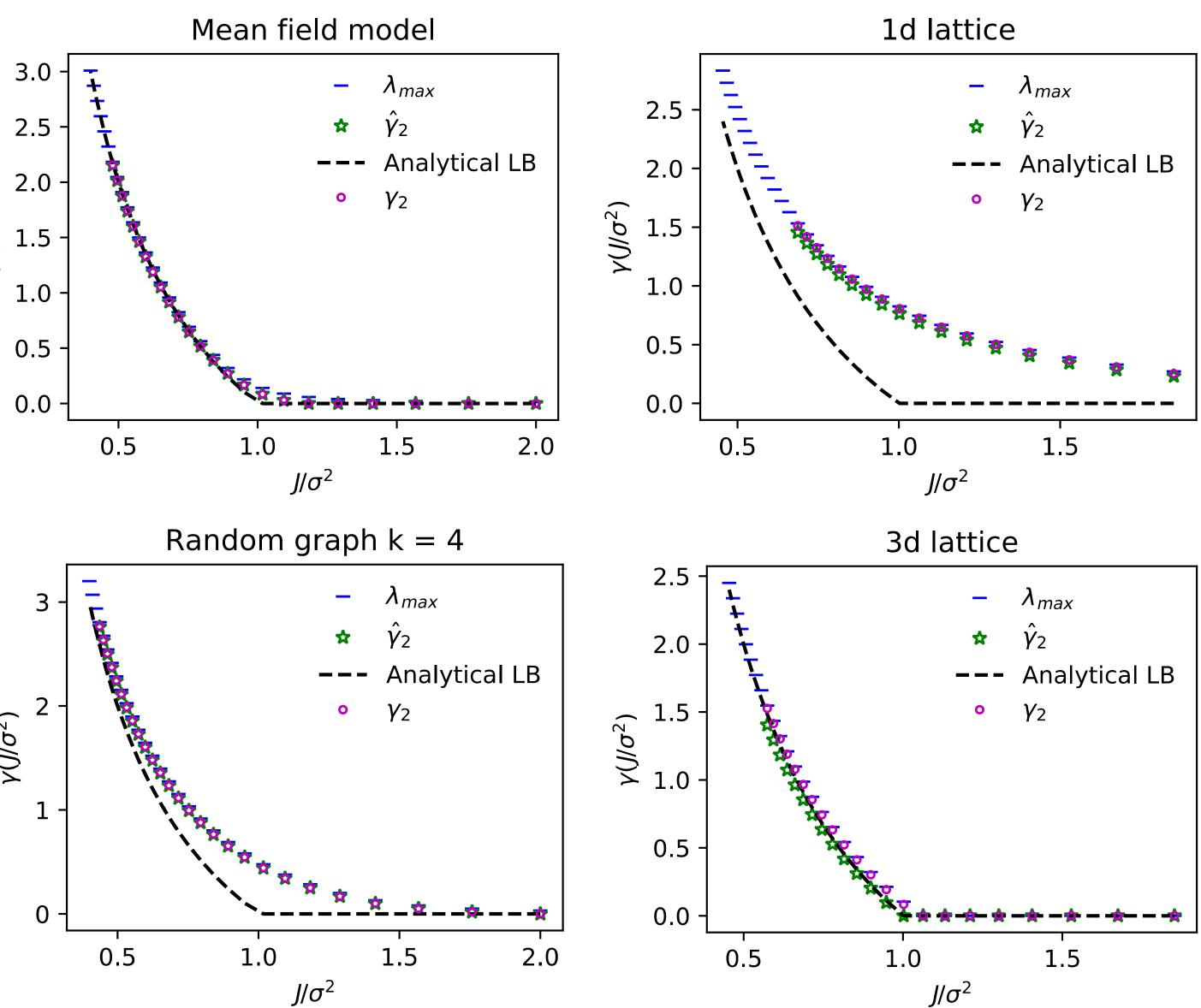
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SUMMARY OF RESULTS

$$_{ij}^{2}(t),$$

 $\frac{\ln(P_{ii}(t))}{=} = \frac{\tilde{d}}{\tilde{d}}$ ln(t)





 $\hat{\gamma}_2$ - sample Lyapunov exponents for a general graph Analytical LB -

where $\langle k \rangle$ is the average degree of the graph

CONCLUSIONS

- dimension and the graph Lyapunov exponents
- We show the existence of a localization phase transition
- problems
- We propose to classify tissues and brain areas, represented as a pore network, based on their spectral dimension
- This may lead to new experiments and observable parameters revealing exciting structural properties of our brain as well as improvements in diagnostic of disease
- noise⁶



- γ_2 moment Lyapunov exponents for a general graph
- $\lambda_{\rm max}$ the largest eigenvalue of the second moment equations

$$\hat{\gamma}_2 \ge 2\sigma^2 - 2J\langle k \rangle,$$

• The complexity of the network can be extracted by measuring the spectral

• Using our **novel spectral technique** one can analyze other complex structured

• Interesting results for time dependent random coupling and spatially varying