

# Spectral renormalization group

---

Ayşe Erzan

Eser Aygün

Istanbul Technical University

- Landau free energy expansion and “Field theoretic” RG a la Wilson
- Graph Laplacian, eigenvalue distribution, eigenvectors
- Integration and rescaling on deterministic hierarchical graphs
- Random graphs: Gaussian theory, replicas
- Critical behaviour

# Landau free energy expansion in the order parameter

---

$$H = \int_{\Omega} d\mathbf{x} \left\{ \frac{1}{2} \left[ r_o \psi^2(\mathbf{x}) - \psi(\mathbf{x}) \nabla^2 \psi \right] + u_o \psi^4(\mathbf{x}) \right\}$$

$$H_0 = \frac{1}{2} \int_0^{\Lambda} \frac{k^{d-1} dk}{(2\pi)^d} [r_o + k^2] \hat{\psi}^*(\mathbf{k}) \hat{\psi}(\mathbf{k})$$

$$H_{\text{int}} = u_o \int_0^{\Lambda} d\mathbf{k}_1 \dots d\mathbf{k}_4 \hat{\psi}(\mathbf{k}_1) \dots \hat{\psi}(\mathbf{k}_4) \delta\left(\sum_{i=1}^4 \mathbf{k}_i\right)$$

where  $\hat{\psi}(\mathbf{k}) = \text{F.T.} \psi(\mathbf{x})$

# Integrate out the small wavelength (large $k$ ) fluctuations

---

$$Z = \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} \prod_{0 \leq |\mathbf{k}| \leq \Lambda/b} d\hat{\psi}_{\mathbf{k}}^{\text{lower}} \prod_{\Lambda/b < |\mathbf{k}| \leq \Lambda} d\hat{\psi}_{\mathbf{k}}^{\text{upper}} e^{-H}$$

The density of states  $\sim k^{d-1}$  is scale free!

Rescale to restore original Hamiltonian  $\rightarrow$  renormalized coupling constants  $r$  and  $u$

Try similar scheme on networks?

# The Graph Laplacian

---

- The analogue of  $(-1 \times)$  the Laplacian on arbitrary networks is given by

$$L = D - A$$

- $A$  is the **adjacency matrix**

$$D_{ij} = \delta_{ij}d_i$$

$d_i$  is the degree of the  $i$ th node;  $i = 1, \dots, N$

The normalized graph Laplacian is defined to be

$$\tilde{L} = D^{-1}L = I - D^{-1}A$$

where  $I$  is the identity matrix.

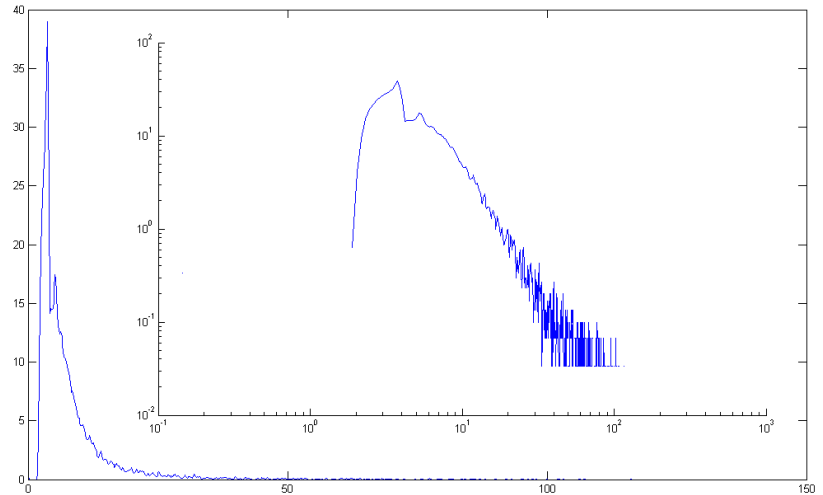
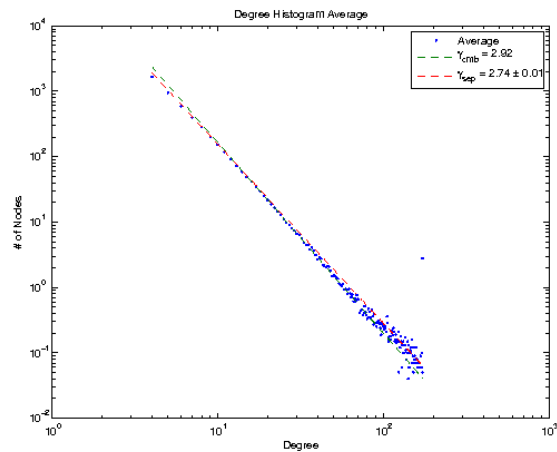
# Properties of the eigenvectors and eigenvalues

---

- $\sum_j L_{ij} = 0$
- Eigenvectors  $L \mathbf{u}_\lambda = \lambda \mathbf{u}_\lambda$  orthogonal  $\mathbf{u}_\lambda^\dagger \mathbf{u}_{\lambda'} = \delta_{\lambda, \lambda'}$
- $\lambda_0 = 0$ ,  $\lambda_1 > 0$  (connected NW),  $0 < \lambda_1 \leq \lambda_2 \dots \lambda_N \equiv \Lambda$
- $u_{\lambda_0}(i) = \text{const.}$
- $\sum_i u_\lambda(i) = 0$  for  $\lambda \neq 0$
- In general,  $u_{\lambda_k}(i) u_{\lambda_l}(i) \dots u_{\lambda_m}(i) \neq u_{\lambda_n}(i)$  for some  $n$ .
- Eigenvectors of the normalized Laplacian  $\mathbf{v}_\mu^\dagger D \mathbf{v}_{\mu'} = \delta_{\mu, \mu'}$
- Eigenvalues  $\mu_0 = 0$ ,  $0 < \mu_1 \leq \mu_2 \dots \mu_N \leq 2$ .

# Degree distribution and Laplacian eigenvalue spectra for a “scale free” network

---

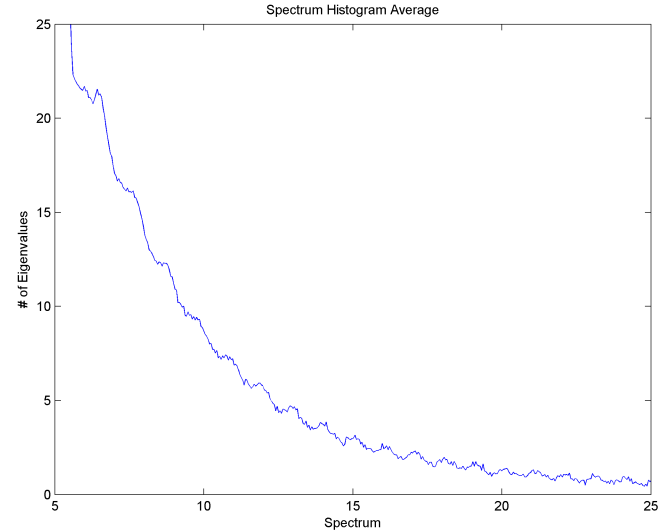
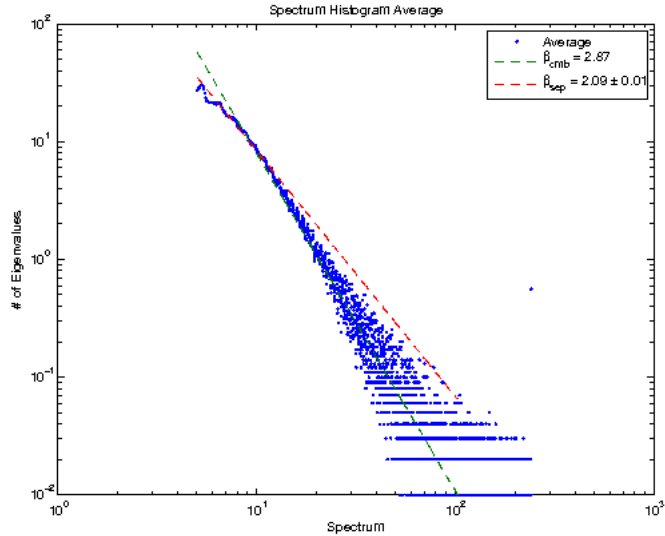


Preferential attachment(BA) model

$m_0 = 5, m = 5, N = 5000, 100$  realizations.

The Laplace spectrum is not scale free; exponential crosses over to power law

# Tail end of the Laplace spectrum



a) superposition, 100 realizations    b) averaged to display fine structure.

Self similar curve

$$f(z) \sim f_0(z) + \sum_{n=0}^N \frac{1}{a^n} \sum_m^M g[b^n(z - m/b^n)]$$

# Expand the order parameter in terms of the eigenvectors of the Laplacian

---

Consider just a Gaussian model

$$H_0 = \frac{1}{2} \sum_i^N \psi(i) [r_o + L] \psi(i) = \frac{1}{2} \int_0^\Lambda \rho(\lambda) d\lambda \hat{\psi}(\lambda) [r_o + \lambda] \hat{\psi}(\lambda)$$

- $\hat{\psi}(\lambda) = \frac{1}{N} \sum_i^N \psi(i) u_\lambda(i)$
- $\rho(\lambda) = \sum_i^N \delta(\lambda - \lambda_i)$



# Naive renormalization

---

## Gaussian theory

Integrate out the coefficients  $\hat{\psi}_\lambda$  for  $\lambda > \Lambda/s$

Assume  $\rho(s\lambda) \sim s^{-\beta} \rho(\lambda)$

$\beta$  depends on the degree distribution

Fix the renormalization factors by requiring the coefficient of the  $\lambda$  term in the Hamiltonian to remain constant

$r = sr_0$  giving  $\nu' = 1$  for any  $\beta$  - correct within Gaussian or MF theory.

f.p. again Gaussian to first order in  $u_0$ .

$\rho(\lambda)$  is not scale free  
“rescaling” is not trivial!

---

- Introduce replicas and average over the realizations of the stochastic network
- Consider hierarchical networks obtained by successive decorations for which the spectrum can be computed iteratively.

# Averaging out over replicas

---

$$Z^n = \int_{-\infty}^{\infty} \prod_{\alpha=1}^n \prod_{\lambda} d\hat{\psi}_{\lambda}^{(\alpha)} e^{-\sum_{\alpha} H_0^{\alpha}}$$

$$\sum_{\alpha} H_0^{\alpha} = \frac{1}{2} \int_0^{\Lambda} \sum_i^N \sum_{\alpha}^n \hat{\psi}_z^{(\alpha)}(r_0 + z) \hat{\psi}_z^{(\alpha)} \delta(z - \lambda_i)$$

Assuming that a limiting distribution  $\rho(\lambda)$  exists, and approximating the distribution over different realizations of the spectrum by  $\mathcal{P}(\{\lambda_i\}) \prod_i d\lambda_i = \prod_i p(\lambda_i) d\lambda_i$

$$\langle Z^n \rangle = \int_0^{\infty} \prod_i p(\lambda_i) d\lambda_i Z^n$$

# Cumulant expansion for effective Hamiltonian

---

Evaluating the expectation value by means of a cumulant expansion up to second order gives, for the effective Hamiltonian

$$\int_0^{\Lambda_{\max}} dz \rho(z) [r_0 + z] \sum_{\alpha} [\hat{\psi}^{(\alpha)}(z)]^2 +$$
$$\int_0^{\Lambda_{\max}} dz \rho(z) z^2 \sum_{\alpha, \beta} [\hat{\psi}^{(\alpha)}(z)]^2 [\hat{\psi}^{(\beta)}(z)]^2$$

Note, however that there is no small parameter in which to expand the fourth order term.

# Restoring the integrated out part of the spectrum

---

$\lambda' = T_s(\lambda)$ , a nonlinear transformation parametrized by the scale factor  $s$  calls for an initial  $\lambda$ -dependent temperature like coupling constant,  $r(\lambda)$ .

Under graph decoration, **Matrix extension transformation** yields all the new eigenvalues obtained in terms of the existing ones

The spectrum of the *normalized* Laplacian is given by the preimages of the *decimation* transformation  $\lambda = R(\lambda')$ , and converges to the Julia set of  $R^{-1}$  as  $N \rightarrow \infty$ .

See:

- N. Bajorin, T. Chen, ..A. Teplyaev, *J. Phys.***A** (2008)
- Z. Zhang et al., *PRE* **80**, (2009)

# Conclusions?

---

The Julia set is chaotic. Thus there is no smooth way in which to rescale existing eigenvalues to restore the integrated-out ones.

The  $\lambda$ -dependent temperature a way to understand the critical regions with power law behaviour of correlations obtained by Real Space RG on scale free networks.