

Imaginary replica analysis of loopy regular random graphs

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5th International Workshop on
Innovative Algorithms for Big Data

Kyoto, Nov 1st 2019

based on joint work with
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Introduction

- Tailoring random graphs
- Loopy random graph ensembles

Imaginary replica approach

- Analysis of the generating function
- Limit of tree-like graphs
- Spectra of loopy graphs

Exactly solvable toy model

- Definitions and generating function
- Phenomenology
- Lessons from the toy model

Return to the imaginary replica method

- Replica analysis in two leading orders
- Tests of the theory
- Applications
- Beyond the region of validity
- Other ensembles

Summary

Tailoring random graphs

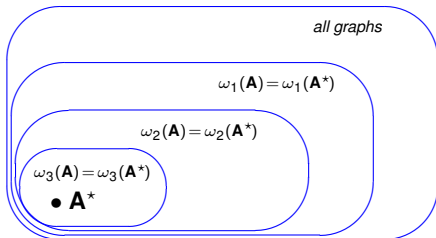
stat mech of process on complex network \mathbf{A}^* ,
use *random graph* \mathbf{A} as proxy

- ▶ max entropy ensembles,
constrained by values
of observables $\omega_1(\mathbf{A}) \dots \omega_L(\mathbf{A})$

$$\text{hard constraints: } p(\mathbf{A}) \propto \prod_{\ell \leq L} \delta_{\omega_\ell(\mathbf{A}), \omega_\ell(\mathbf{A}^*)}$$

$$\text{soft constraints: } p(\mathbf{A}) \propto e^{\sum_{\ell=1}^L \hat{\omega}_\ell \omega_\ell(\mathbf{A})}, \quad \langle \omega_\ell(\mathbf{A}) \rangle = \omega_\ell(\mathbf{A}^*) \quad \forall \ell$$

- ▶ larger $L \rightarrow$
better approximation



The problem

- ▶ nearly all real-world networks:
(physics, biology, economics,
computer science, ...)
 - *sparse graphs*,
 - *many short loops*
- ▶ max entropy graph ensembles
with prescribed degree stats:
 - *sparse graphs*,
 - *locally tree-like*
- ▶ realistic tailoring of graphs
requires enforcing short loops

but

most available analysis methods,
e.g. replicas, GFA, cavity, BP, ...

assume graphs are locally tree-like



Loopy random graph ensembles

- ▶ *simplest model*
(Strauss '86, Jonasson '99)

nondirected graphs

maximum entropy ensemble,
constrain average degree $\langle k \rangle$
and density of **triangles** $\langle m \rangle$

$$p(\mathbf{A}) \propto e^{\alpha \sum_{ij} A_{ij} + \beta \sum_{ijk} A_{ij} A_{jk} A_{ki}}$$

generating
function:

$$\phi(\alpha, \beta) = \frac{1}{N} \log \sum_{\mathbf{A}} e^{\alpha \sum_{ij} A_{ij} + \beta \sum_{ijk} A_{ij} A_{jk} A_{ki}}$$

$$\langle k \rangle = \partial \phi / \partial \alpha$$

$$\langle m \rangle = \partial \phi / \partial \beta$$

- tricky in finite connectivity regime (sum over graphs?)
- pathological behaviour ...

better loopy graph models?

- maximum entropy
- hard constrained degrees
- soft constrained nrs of short loops

▶ $\text{Tr}(\mathbf{A}^\ell) = N \int d\mu \mu^\ell \varrho(\mu|\mathbf{A}),$

$\varrho(\mu|\mathbf{A})$: spectrum of \mathbf{A}

$$p(\mathbf{A}) \propto e^{N \int d\mu \hat{\varrho}(\mu) \varrho(\mu|\mathbf{A})} \prod_i \delta_{k_i, \sum_j A_{ij}}$$

generating function:

$$\phi[\hat{\varrho}] = \frac{1}{N} \log \sum_{\mathbf{A}} e^{N \int d\mu \hat{\varrho}(\mu) \varrho(\mu|\mathbf{A})} \prod_i \delta_{k_i, \sum_j A_{ij}} \quad \varrho(\mu) = \delta\phi / \delta\hat{\varrho}(\mu)$$

- ▶ *simplest example*

$k_i = q$ (regular random graphs)

$\hat{\varrho}(\mu) = \alpha\mu^3$ (constrain nr of triangles)

$$p(\mathbf{A}) \propto e^{\alpha \text{Tr}(\mathbf{A}^3)} \prod_i \delta_{q, \sum_j A_{ij}}, \quad \phi(\alpha) = \frac{1}{N} \log \sum_{\mathbf{A}} e^{\alpha \text{Tr}(\mathbf{A}^3)} \prod_i \delta_{q, \sum_j A_{ij}}$$

Making calculations feasible

$$\text{graph ensemble : } p(\mathbf{A}) = \frac{1}{Z[\hat{\rho}]} e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})} \prod_i \delta_{k_i, \sum_j A_{ij}}$$

$$\text{generating function : } \Phi[\hat{\rho}] = \frac{1}{N} \log \sum_{\mathbf{A}} e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})} \prod_i \delta_{k_i, \sum_j A_{ij}}$$

- ▶ derive via Edwards-Jones spectrum formula:

$$e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})} = \lim_{\varepsilon, \Delta \downarrow 0} \lim_{n_\mu \rightarrow \frac{i\Delta}{\pi} \frac{d}{d\mu} \hat{\rho}(\mu)} \lim_{m_\mu \rightarrow -n_\mu} \prod_{\mu} \left[Z(\mu + i\varepsilon|\mathbf{A})^{n_\mu} \overline{Z(\mu + i\varepsilon|\mathbf{A})}^{m_\mu} \right]$$

$$Z(\mu|\mathbf{A}) = \int_{\mathbb{R}^N} d\phi e^{-\frac{1}{2}i\phi \cdot [\mathbf{A} - \mu \mathbf{1}]\phi}$$

- ▶
 1. replica method
 2. path integral
 3. steepest descent for $N \rightarrow \infty$
 4. replica symmetric soln
 5. analytical continuation to *imaginary* (n_μ, m_μ)
 6. limits $\varepsilon \downarrow 0, \Delta \downarrow 0$

Replica symmetric theory

$$\mathcal{W}[\{x\}] = \frac{1}{\mathcal{C}^2} \sum_{k>0} \rho(k) \frac{k}{\langle k \rangle} \frac{\mathcal{A}[\{x\}] \mathcal{F}_{k-1}[\{x\}]}{\int \{dx'\} \mathcal{A}[\{x'\}] \mathcal{F}_k[\{x\}]}$$

$$\mathcal{C}^2 = \sum_{k>0} \rho(k) \frac{k}{\langle k \rangle} \frac{\int \{dx\} \mathcal{A}[\{x\}] \mathcal{F}_{k-1}[\{x\}]}{\int \{dx\} \mathcal{A}[\{x\}] \mathcal{F}_k[\{x\}]}$$

with

$$\mathcal{F}_k[\{x\}] = \left[\prod_{\ell \leq k} \int \{dx_\ell\} \mathcal{W}[\{x_\ell\}] \right] \delta_{\mathbb{F}} [x - F[x_1, \dots, x_k]]$$

$$\mathcal{A}[\{x\}] = e^{-\frac{1}{2} \int d\mu \hat{\rho}(\mu) \frac{d}{d\mu} \text{sgn}[x(\mu)]}$$

spectrum:

$$\begin{aligned} \varrho(\mu) = & -\frac{1}{2} \frac{d}{d\mu} \left\{ \sum_k \rho(k) \frac{\int \{dx\} \mathcal{A}[\{x\}] \mathcal{F}_k[\{x\}] \text{sgn}[x(\mu)]}{\int \{dx\} \mathcal{A}[\{x\}] \mathcal{F}_k[\{x\}]} \right. \\ & + \langle k \rangle \mathcal{C}^2 \int \{dx dx'\} \mathcal{W}[\{x\}] \mathcal{W}[\{x'\}] \mathcal{B}[\{x\}, \{x'\}] \\ & \left. \times \theta[x(\mu)x'(\mu)] \theta[1 - x(\mu)x'(\mu)] \text{sgn}[x(\mu) + x'(\mu)] \right\} \end{aligned}$$

$$\mathcal{B}[\{x\}, \{x'\}] = e^{\int d\mu \hat{\rho}(\mu) \frac{d}{d\mu} \left\{ \theta[x(\mu)x'(\mu)] \theta[1 - x(\mu)x'(\mu)] \text{sgn}[x(\mu) + x'(\mu)] \right\}}$$

Limit of locally tree-like graphs

$$\hat{\rho}(\mu) \rightarrow 0 \text{ for all } \mu: \quad p(\mathbf{A}) \propto \prod_i \delta_{k_i, \sum_j A_{ij}}$$

- ▶ entropy per node:

$$S = \frac{1}{2} \langle k \rangle \left[\log \left(\frac{N}{\langle k \rangle} \right) + 1 \right] + \sum_k p(k) \log \tilde{p}(k) + \epsilon_N$$



- ▶ spectrum for regular graphs, $k > 1$:
McKay's '81 formula

$$\varrho(\mu) = \theta \left[2\sqrt{k-1} - |\mu| \right] \frac{k\sqrt{4(k-1) - \mu^2}}{2\pi(k^2 - \mu^2)}$$

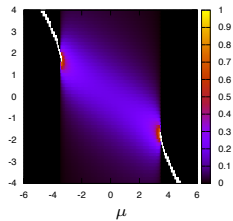
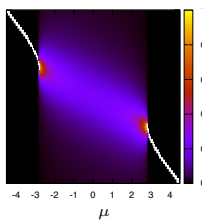
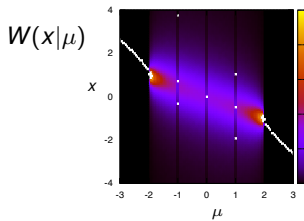
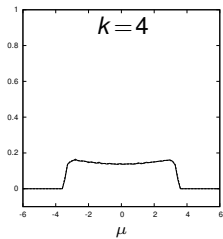
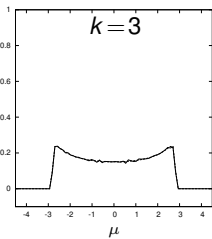
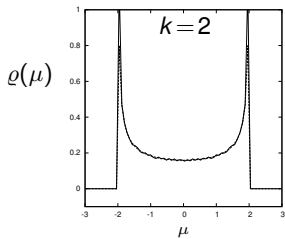


- ▶ spectrum for arbitrary $p(k)$:
Dorogovtsev et al '03 formula

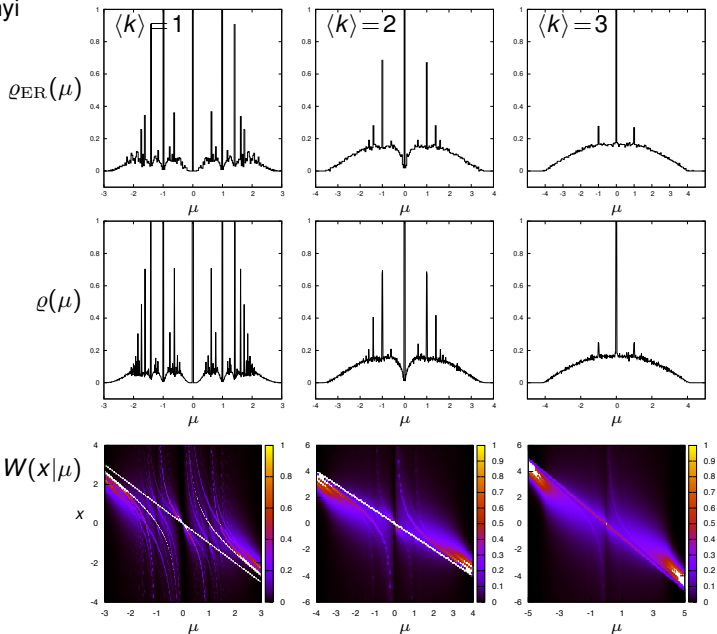
$$G(z|\mu) = 1 - \sqrt{z} \int_0^\infty \frac{dy}{\sqrt{y}} e^{iy\mu} \Phi(G(y|\mu)) J_1(2\sqrt{yz})$$



Regular treelike graphs:



Erdős-Rényi
graphs:

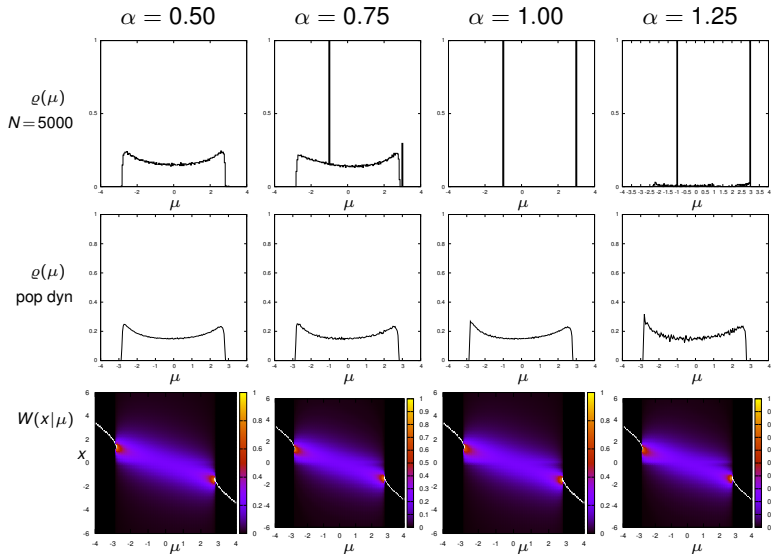


Spectra of loopy graphs

regular graphs, $k = 3$

$$\hat{\rho}(\mu) = \alpha \text{Tr}(\mathbf{A}^3)$$

2016 ...
what is going on?



Possible causes of the deviation

- ▶ Numerical precision?

extend MCMC simulation times in graph generation,
relation between μ -resolution in $x(\mu)$ and pop dynamics convergence

- ▶ Analytical technicalities?

issues related to complex logarithm,
other saddle-point types,
order or limits $\epsilon \downarrow 0$, $\Delta \downarrow 0$, $n_\mu \rightarrow i(\Delta/\pi) \frac{d}{d\mu} \hat{\rho}(\mu)$, ...

- ▶ Some fundamental problem in the theory?

Numerical generation of tailored loopy graphs

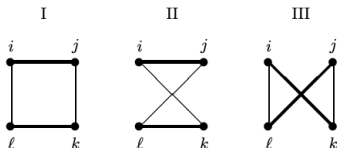
- desired measure:

$$p(\mathbf{A}) = \frac{e^{-H(\mathbf{A})}}{Z} \prod_{i=1}^N \delta_{k_i, \sum_j A_{ij}}$$

- initialize \mathbf{A} to *any* graph with correct degrees

- MCMC process:

elementary moves:
edge swaps
(preserve degrees)



move acceptance prob:

$$w(\mathbf{A} \rightarrow F\mathbf{A}) = \frac{1}{1 + \exp[\Delta H(\mathbf{A}) + \log(1 + \Delta n(\mathbf{A})/n(\mathbf{A}))]}$$

- entropic term $\Delta n(\mathbf{A})/n(\mathbf{A})$ reflects *change* in nr of available moves

$$n(\mathbf{A}) = \underbrace{\frac{1}{4}N^2\langle k \rangle^2 + \frac{1}{4}N\langle k \rangle - \frac{1}{2}N\langle k^2 \rangle}_{\text{invariant}} + \underbrace{\frac{1}{4}\text{Tr}(\mathbf{A}^4) + \frac{1}{2}\text{Tr}(\mathbf{A}^3) - \frac{1}{2}\sum_{ij} k_i A_{ij} k_j}_{\text{state dependent}}$$

Analytical technicalities

are we maybe too cavalier with complex logarithms?

- ▶ basic building block:
Edwards-Jones formula

$$\varrho(\mu) = -\lim_{\epsilon \downarrow 0} \frac{2}{\pi N} \operatorname{Im} \frac{\partial}{\partial \mu} \log Z_N(\mu + i\epsilon | \mathbf{A}), \quad Z_N(\mu | \mathbf{A}) = \int_{\mathbb{R}^N} d\mathbf{x} e^{-\frac{1}{2} \mathbf{i} \mathbf{x} \cdot (\mathbf{A} - \mu \mathbf{I}) \mathbf{x}}$$

- ▶ naive derivation uses $\log \prod_i z_i = \sum_i \log z_i$
(true only for $z_i \in \mathbb{R}$)

any conventional defn of
complex log (wherever the cut):

$$\begin{aligned} \lim_{N \rightarrow \infty} \frac{1}{N} \operatorname{Im} \log Z_N &= \lim_{N \rightarrow \infty} \frac{1}{N} \operatorname{Im} \left(\log |Z_N| + i \operatorname{Arg}(Z_N) \right) \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} \operatorname{Arg}(Z_N) = 0 \dots \end{aligned}$$

- ▶ E-J formula has been very successful,
subtleties become problematic only for imaginary replicas?

Possible causes of the deviation

- ▶ Numerical precision?

extend MCMC simulation times in graph generation,
relation between μ -resolution in $x(\mu)$ and pop dynamics convergence

- ▶ Analytical technicalities?

issues related to complex logarithm,
other saddle-point types,
order or limits $\epsilon \downarrow 0$, $\Delta \downarrow 0$, $n_\mu \rightarrow i(\Delta/\pi) \frac{d}{d\mu} \hat{g}(\mu)$, ...

- ▶ Some fundamental problem in the theory?

*find exactly solvable loopy graph models
that don't require imaginary replicas*

Definition of the toy model

- ▶ ensemble of degree-2 regular random graphs

$$p(\mathbf{A}) = \frac{1}{Z_N(\boldsymbol{\alpha})} \times \overbrace{e^{\sum_{\ell=3}^K \ell \alpha_{\ell} n_{\ell}(\mathbf{A})}}^{\text{control nrs of cycles}} \times \overbrace{\prod_{i=1}^N \delta_{2, \sum_j A_{ij}}}^{\text{graphs: sets of rings}}$$

$n_{\ell}(\mathbf{A})$: nr of length- ℓ cycles, K finite

- ▶ generating function

$$\phi_N(\boldsymbol{\alpha}) = \frac{1}{N} \log[Z_N(\boldsymbol{\alpha})/N!], \quad Z_N(\boldsymbol{\alpha}) = \sum_{\mathbf{A}} e^{\sum_{\ell=3}^K \ell \alpha_{\ell} n_{\ell}(\mathbf{A})} \prod_{i=1}^N \delta_{2, \sum_j A_{ij}}$$

fraction of nodes
in length- ℓ cycles

$$m_{\ell} = \frac{\ell}{N} \langle n_{\ell}(\mathbf{A}) \rangle = \frac{\partial}{\partial \alpha_{\ell}} \phi_N(\boldsymbol{\alpha})$$

- ▶ solution:

combinatorics of dividing N nodes into closed rings of different lengths

Solution

- ▶ result of combinatorics

$$\phi_N(\alpha) = \frac{1}{N} \log \int_{-\pi}^{\pi} \frac{d\omega}{2\pi} e^{Nf_N(\omega, \alpha)}$$
$$f_N(\omega, \alpha) = i\omega + \sum_{\ell=3}^K \frac{e^{(\alpha_\ell - i\omega)\ell}}{2\ell N} + \sum_{\ell=K+1}^N \frac{e^{-i\omega\ell}}{2\ell N}$$

- ▶ Large N

$$\phi(\alpha) = \lim_{N \rightarrow \infty} \text{extr}_\omega f_N(\omega, \alpha)$$

- ▶ finite $\{\alpha_\ell\}$: $m_\ell = 0$ for all ℓ ,
 $\{\alpha_\ell\}$ must scale differently with N
to induce finite cycle densities!
- ▶ $\alpha_\ell = \tilde{\alpha}_\ell + \ell^{-1} \log N$

$$\tilde{f}_N(\omega, \alpha) = i\omega + \sum_{\ell=3}^K \frac{e^{(\tilde{\alpha}_\ell - i\omega)\ell}}{2\ell} + \sum_{\ell=K+1}^N \frac{e^{-i\omega\ell}}{2\ell N}$$

Phenomenology of the toy model

fraction of nodes
in large rings

$$m_\infty = 1 - \sum_{\ell=3}^K m_\ell$$

graph spectrum:

$$\varrho(\mu) = \sum_{\ell=3}^K m_\ell \varrho_\ell(\mu) + m_\infty \frac{\theta(2-|\mu|)}{\pi\sqrt{4-\mu}}$$

$$\varrho_\ell(\mu) = \frac{1}{\ell} \sum_{r=0}^{\ell-1} \delta[\mu - 2 \cos(2\pi r/\ell)]$$

► Two phases

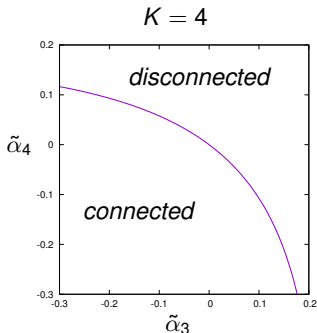
connected phase: $m_\infty > 0$

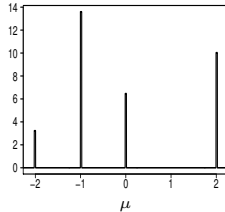
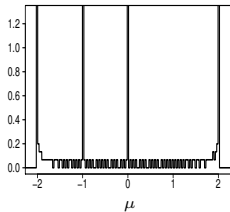
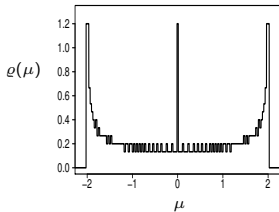
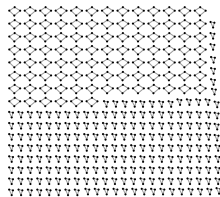
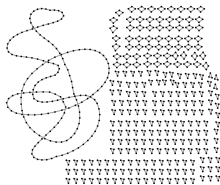
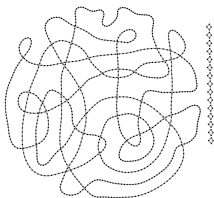
presence of extensively large rings,
found for small $\{\alpha_\ell\}$

disconnected phase: $m_\infty = 0$

only short rings,
found for large $\{\alpha_\ell\}$

percolation transition: $\sum_{\ell=3}^K e^{\ell\tilde{\alpha}_\ell} = 2$





disconnected phase :
$$m_\ell = \frac{1}{2} x^\ell e^{\ell \tilde{\alpha}_\ell}, \quad 1 = \frac{1}{2} \sum_{\ell=3}^K x^\ell e^{\ell \tilde{\alpha}_\ell}$$

connected phase :
$$m_\ell = \frac{1}{2} e^{\ell \tilde{\alpha}_\ell}, \quad m_\infty = 1 - \frac{1}{2} \sum_{\ell=3}^K e^{\ell \tilde{\alpha}_\ell}$$

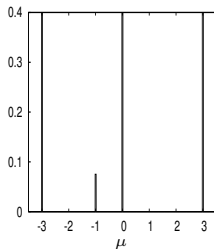
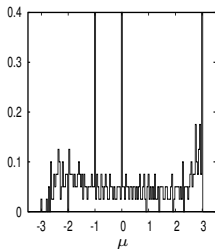
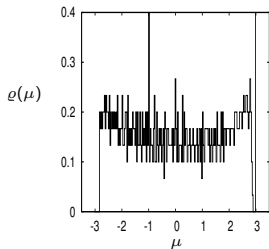
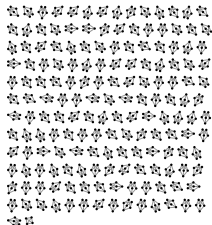
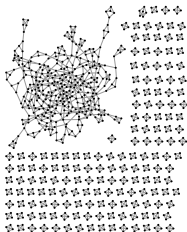
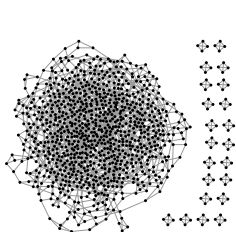
Graphs with degree $k \geq 3$

no longer solvable directly,
but similar phenomenology ...

$$p(\mathbf{A}) = \frac{1}{Z_N(\alpha)} e^{\alpha_3 n_3(\mathbf{A}) + \alpha_4 n_4(\mathbf{A})} \prod_{i=1}^N \delta_{3, \sum_j A_{ij}}$$

$$\alpha_\ell = \tilde{\alpha}_\ell + \ell^{-1} \log N$$

$k = 3$:



Lessons from the toy model

- ▶ to achieve $\mathcal{O}(1)$ short loop densities and modified spectra, we need (for regular random graphs):

$$\hat{\varrho}(\mu) \rightarrow \log N \varphi(\mu) + \hat{\varrho}(\mu)$$

but this gives integrals in the replica theory that may no longer be amenable to steepest descent ...

- ▶ options

1. keep $\hat{\varrho}(\mu) = \mathcal{O}(1)$:

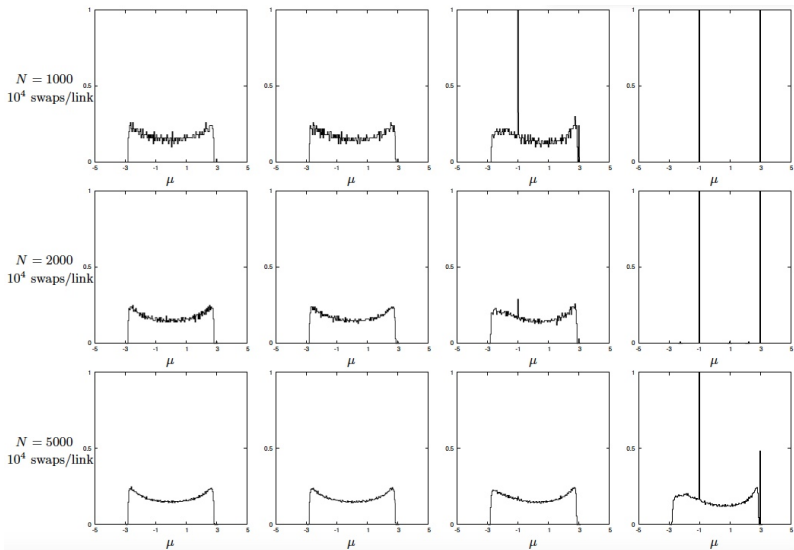
study *finite size corrections* to loop numbers and spectra, extend theory to include also sub-leading order in N

2. choose $\hat{\varrho}(\mu) = \mathcal{O}(\log N)$:

requires extending steepest descent arguments to expressions of the form

$$f = \lim_{M \rightarrow \infty} \frac{1}{M} \log \int_{\mathbb{R}^M} dx e^{MF(x)}$$

simulations indeed confirm that with $\hat{\rho}(\mu) = \alpha\mu^3$, the deviations from McKay's spectrum, unexplained by the replica theory, were *finite size effects* ...



steepest descent type integrals in high dimensions

- ▶ define

$$I(d, N) = \frac{1}{N} \log \int_{\mathbb{R}^d} d\mathbf{x} e^{-N\Phi(\mathbf{x})}, \quad d, N \rightarrow \infty, \quad d/N = \alpha \text{ finite}$$

- ▶ standard steepest descent:
 $d/N \rightarrow 0$, sufficiently fast

- ▶ For $\alpha > 0$:

$$I(d, N) = \min_{\mathbf{x}} \Phi(\mathbf{x}) + \frac{1}{2} \alpha \log\left(\frac{d}{2\pi\alpha}\right) + \frac{\alpha}{2d} \sum_{i=1}^d \log a_i + o(1)$$

(as if approx by Gaussian integral)
provided:

$$\lim_{d \rightarrow \infty} \frac{1}{d} \sum_{i=1}^d \frac{1}{a_i} = 0 \quad \text{and} \quad \lim_{d \rightarrow \infty} \frac{1}{d^2} \sum_{i=1}^d \frac{1}{a_i^2} = 0$$

$\{a_i\}$: eigenvals of curvature matrix of Φ at minimum

Imaginary replica analysis in two leading orders

ensemble:

$$p(\mathbf{A}) = \frac{e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})}}{Z[\hat{\rho}]} \prod_{i=1}^N \delta_{q, \sum_j A_{ij}} \quad Z[\hat{\rho}] = \sum_{\mathbf{A}} e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})} \prod_{i=1}^N \delta_{q, \sum_j A_{ij}}$$

- find $\phi[\hat{\rho}]$ in leading two orders in N

$$\phi[\hat{\rho}] = \frac{1}{N} \log Z[\hat{\rho}], \quad \varrho(\mu) = \frac{\delta \phi[\hat{\rho}]}{\delta \hat{\rho}(\mu)}$$

$$e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})} = \lim_{\Delta \rightarrow 0} \lim_{n_\mu \rightarrow i \frac{\Delta}{\pi} \hat{\rho}'(\mu)} \lim_{m_\mu \rightarrow -n_\mu} \prod_{\mu} Z(\mu_\epsilon|\mathbf{A})^{n_\mu} \overline{Z(\mu_\epsilon|\mathbf{A})}^{m_\mu}$$

$$Z(\mu_\epsilon|\mathbf{A}) = \int \prod_{i=1}^N d\phi^i \exp \left[-\frac{i}{2} \sum_{ij} \phi^i (A_{ij} - \mu_\epsilon \delta_{ij}) \phi^j \right], \quad \mu_\epsilon \equiv \mu + i\epsilon$$

replicas:

$$\prod_{\mu} Z(\mu_\epsilon|\mathbf{A})^{n_\mu} \overline{Z(\mu_\epsilon|\mathbf{A})}^{m_\mu}$$

$$= \int d\Phi d\Psi e^{-\frac{i}{2} \sum_{\mu, \alpha_\mu} \sum_{ij} \phi_{\mu, \alpha_\mu}^i \phi_{\mu, \alpha_\mu}^j (A_{ij} - \mu_\epsilon \delta_{ij}) + \frac{i}{2} \sum_{\mu, \beta_\mu} \sum_{ij} \psi_{\mu, \beta_\mu}^i \psi_{\mu, \beta_\mu}^j (A_{ij} - \overline{\mu_\epsilon} \delta_{ij})}$$

connect notation to Metz et al (2014) and Lucibello et al (2014)
 (finite size calculations for *treelike* graphs)

► $\varphi \in \mathbb{R}^d$, where $d = \sum_{\mu} n_{\mu} + \sum_{\mu} m_{\mu}$:

$$\varphi = \begin{pmatrix} \{\phi_{\mu, \alpha_{\mu}}\} \\ \{\psi_{\mu, \beta_{\mu}}\} \end{pmatrix}, \quad \varphi \cdot \varphi' = \sum_{\mu, \alpha_{\mu}} \phi_{\mu, \alpha_{\mu}} \phi'_{\mu, \alpha_{\mu}} - \sum_{\mu, \beta_{\mu}} \psi_{\mu, \beta_{\mu}} \psi'_{\mu, \beta_{\mu}}$$

$d \times d$ matrix μ :

$$\mu = \begin{pmatrix} \mu_1 \varepsilon \mathbf{I}_{n_{\mu_1}} & \mathbf{0} & \dots & \mathbf{0} \\ \mathbf{0} & \ddots & \mathbf{0} & \vdots \\ \mathbf{0} & \mu_M \varepsilon \mathbf{I}_{n_{\mu_M}} & \mathbf{0} & \vdots \\ \vdots & \mathbf{0} & \overline{\mu_1} \varepsilon \mathbf{I}_{m_{\mu_1}} & \mathbf{0} \\ \mathbf{0} & \dots & \mathbf{0} & \ddots & \mathbf{0} \\ \mathbf{0} & \dots & \mathbf{0} & \mathbf{0} & \overline{\mu_M} \varepsilon \mathbf{I}_{m_{\mu_M}} \end{pmatrix}$$

and

$$\nu(\varphi) = e^{\frac{1}{2} i \varphi \cdot \mu \varphi}, \quad \lim = \lim_{\Delta \rightarrow 0} \lim_{n_{\mu} \rightarrow i \frac{\Delta}{\pi}} \lim_{\hat{\rho}'(\mu)} \lim_{m_{\mu} \rightarrow -n_{\mu}}$$

Result over sum over \mathbf{A}
and expansion:

$$\phi[\hat{\rho}] = \lim \frac{1}{N} \log \int \mathcal{D}P \mathcal{D}\hat{P} e^{N S[P, \hat{P}]} + \mathcal{O}\left(\frac{1}{N^2}\right),$$

$$\begin{aligned} S[P, \hat{P}] &= -\frac{q}{2}\left(1 + \frac{q-2}{2N}\right) + i \int d\varphi d\omega P(\varphi, \omega) \hat{P}(\varphi, \omega) \\ &+ \frac{q}{2}\left(1 + \frac{q}{N}\right) \int d\varphi d\varphi' d\omega d\omega' P(\varphi, \omega) P(\varphi', \omega') e^{-i\varphi \cdot \varphi' - i\omega - i\omega'} \\ &- \frac{q^2}{4N} \int d\varphi d\varphi' d\omega d\omega' P(\varphi, \omega) P(\varphi', \omega') e^{-2i\varphi \cdot \varphi' - 2i\omega - 2i\omega'} \\ &- \frac{q}{2N} \int d\varphi d\omega P(\varphi, \omega) e^{-i\varphi \cdot \varphi - 2i\omega} + \log \int \frac{d\omega d\varphi}{2\pi} \nu(\varphi) e^{i\omega q - i\hat{P}(\varphi, \omega)} \end{aligned}$$

*same form as Metz et al (2014),
but different defs of φ and $\nu(\varphi)$*

- ▶ retrace calculations of Metz et al ...

$$\phi[\hat{\rho}] = \lim \left\{ \log Z_q + \sum_{\ell=3}^{\infty} \frac{(q-1)^\ell}{2N\ell} \text{Tr}(\mathbf{M}^\ell) \right\}$$

with

$$Z_q = \int d\varphi \nu(\varphi) \left[\int d\varphi' U_1(\varphi, \varphi') W_0(\varphi') \right]^q, \quad U_1(\varphi, \varphi') = e^{-i\varphi \cdot \varphi'}$$

$$M(\varphi, \varphi') = \frac{\nu(\varphi)}{Z_q} \left[\int d\varphi' U_1(\varphi, \varphi') W_0(\varphi') \right]^{q-2} U_1(\varphi, \varphi'),$$

order par eqn:

$$W_0(\varphi) = \frac{\nu(\varphi)}{Z_q} \left[\int d\varphi' U_1(\varphi, \varphi') W_0(\varphi') \right]^{q-1}$$

*same form as Metz et al (2014),
but different defs of φ and $\nu(\varphi)$*

► RS ansatz,

now differences with Metz et al
become apparent

$$W_0(\varphi) = c \int d\mathbf{X} W(\mathbf{X}) [e^{-\frac{1}{2}i\varphi \cdot \mathbf{X}\varphi} / Z(\mathbf{X})]$$

$$\int d\mathbf{X} W(\mathbf{X}) = 1, \quad Z(\mathbf{X}) = \prod_{\mu} \left(\frac{2\pi}{iX(\mu)} \right)^{\frac{n_{\mu}}{2}} \left(\frac{2\pi}{iX(\mu)} \right)^{\frac{m_{\mu}}{2}}$$

$$\mathbf{X} = \begin{pmatrix} x(\mu_1) \mathbf{I}_{n_{\mu_1}} & \mathbf{0} & \dots & \mathbf{0} \\ \mathbf{0} & \ddots & \mathbf{0} & \vdots \\ \mathbf{0} & x(\mu_M) \mathbf{I}_{n_{\mu_M}} & \mathbf{0} & \vdots \\ \vdots & \mathbf{0} & \overline{x(\mu_1)} \mathbf{I}_{m_{\mu_1}} & \mathbf{0} \\ \mathbf{0} & \dots & \mathbf{0} & \ddots & \mathbf{0} \\ \mathbf{0} & \dots & \mathbf{0} & \mathbf{0} & \overline{x(\mu_M)} \mathbf{I}_{m_{\mu_M}} \end{pmatrix}$$

- ▶ RS order par eqn:

$$W(\mathbf{X}) = \frac{Z(\mathbf{X})}{\mathcal{Z}_{q-1}} \int \left(\prod_{k=1}^{q-1} d\mathbf{X}_k W(\mathbf{X}_k) \right) \delta\left(\mathbf{X} + \boldsymbol{\mu} + \sum_{k=1}^{q-1} \mathbf{X}_k^{-1}\right)$$

$$\mathcal{Z}_q = \int d\mathbf{X} \left(\prod_{k=1}^q d\mathbf{X}_k W(\mathbf{X}_k) \right) Z(\mathbf{X}) \delta\left(\mathbf{X} + \boldsymbol{\mu} + \sum_{k=1}^q \mathbf{X}_k^{-1}\right)$$

- ▶ special soln:

$$W(\mathbf{X}) = \delta(\mathbf{X} - \mathbf{X}^*), \quad \mathbf{x}^*(\mu) = -\frac{1}{2}\mu_\epsilon - \frac{1}{2}i\sqrt{4(q-1) - \mu_\epsilon^2}$$

$$\phi[\hat{\rho}] = \lim \left\{ \frac{1}{2}q \log Z(\mathbf{X}^*) - \frac{1}{2}(q-2) \log Z(\mathbf{X}^* - (\mathbf{X}^*)^{-1}) \right. \\ \left. + \sum_{\ell=3}^{\infty} \frac{(q-1)^\ell}{2N\ell} \frac{1}{Z^\ell(\mathbf{X}^*)} \prod_{\mu} \left[Z(\mu_\epsilon | \mathbf{A}_{\ell,\mu}^*)^{n_\mu} \overline{Z(\mu_\epsilon | \mathbf{A}_{\ell,\mu}^*)}^{m_\mu} \right] \right\}$$

$\mathbf{A}_{\ell,\mu}^*$: $\ell \times \ell$ adjacency matrix of length- ℓ loop,
with complex field on diagonal:

$$(\mathbf{A}_{\ell,\mu}^*)_{kk'} = \delta_{k,k'+1} + \delta_{k,k'-1} + \frac{2-q}{\mathbf{x}^*(\mu)} \delta_{kk'} \quad (\text{with } k \text{ mod } \ell)$$

► imaginary replica limits

$$\phi[\hat{\varrho}] = \int d\mu \hat{\varrho}(\mu) \varrho_0(\mu) + \frac{1}{N} \sum_{\ell=3}^{\infty} \frac{(q-1)^\ell}{2^\ell} e^{\ell \int d\mu \hat{\varrho}(\mu) G_\ell(\mu)} + o\left(\frac{1}{N}\right)$$

finite size corrections

$$\varrho(\mu) = \varrho_0(\mu) + \frac{1}{2N} \sum_{\ell=3}^{\infty} (q-1)^\ell e^{\ell \int d\mu \hat{\varrho}(\mu) G_\ell(\mu)} G_\ell(\mu) + o\left(\frac{1}{N}\right)$$

with

$$\varrho_0(\mu) = \frac{q}{2\pi} \frac{\sqrt{4(q-1) - \mu^2}}{q^2 - \mu^2} \theta[2\sqrt{q-1} - |\mu|] \quad (\text{McKay's formula})$$

$$G_\ell(\mu) = \frac{1}{\ell\pi} \text{Im} \left\{ \sum_{k=1}^{\ell} \frac{1 + i \frac{q-2}{x^*(\mu)} [4(q-1) - \mu^2]^{-\frac{1}{2}}}{2 \cos(2\pi k/\ell) - \frac{q-2}{x^*(\mu)} - \mu} \right\} - \frac{1}{\pi} \frac{\theta(2\sqrt{q-1} - |\mu|)}{\sqrt{4(q-1) - \mu^2}}$$

► interpretation

$(q-1)^\ell / 2^\ell$: expected nr of length- ℓ loops in random regular graph

$\ell G_\ell(\mu)$: correction to McKay formula due to single length- ℓ loop

→ finite size corrections from well separated loops

- ▶ remaining integrals that involve $\hat{\varrho}(\mu)$:

$$\varrho(\mu) = \varrho_0(\mu) + \frac{1}{2N} \sum_{\ell=3}^{\infty} (q-1)^{\ell} e^{\ell \mathcal{J}_{\ell}[\hat{\varrho}]} G_{\ell}(\mu) + o\left(\frac{1}{N}\right)$$

$$\mathcal{J}_{\ell}[\hat{\varrho}] = \sum_{\rho=1}^{\infty} \frac{1}{(q-1)^{\rho\ell/2}} \frac{2}{\pi} \int_{-1}^1 \frac{dt}{\sqrt{1-t^2}} \hat{\varrho}(2t\sqrt{q-1}) T_{\rho\ell}(t).$$

$T_n(t)$: Chebyshev polynomials, $t \in [-1, 1]$:

$$T_n(t) = \operatorname{Re}(t + i\sqrt{1-t^2})^n : \quad T_0(t) = 1, \quad T_1(t) = t, \quad T_2(t) = 2t^2 - 1, \dots$$

$$\frac{2}{\pi} \int_{-1}^1 \frac{dt}{\sqrt{1-t^2}} T_n(t) T_m(t) = \delta_{nm} (1 + \delta_{m0})$$

Tests of the theory

- ▶ *flat measures:*

$$\hat{\varrho}(\mu) = 0$$

$$\varrho(\mu) = \varrho_0(\mu) + \frac{1}{2N} \sum_{\ell=3}^{\infty} (q-1)^{\ell} G_{\ell}(\mu) + o(N^{-1})$$

reproduces formula of Metz et al (2014) ✓

- ▶ *toy model:*

$$q = 2, \quad \hat{\varrho}(\mu) = \alpha \mu^3$$

$$\varrho(\mu) = \varrho_0(\mu)$$

$$+ \frac{1}{2N} \sum_{\ell=3}^{\infty} e^{6\alpha\delta_{\ell 3}} \left\{ \frac{1}{\ell} \sum_{k=1}^{\ell} \delta \left[\mu - 2 \cos \left(\frac{2\pi k}{\ell} \right) \right] - \frac{1}{\pi} \frac{\theta(2-|\mu|)}{\sqrt{4-\mu^2}} \right\} + o\left(\frac{1}{N}\right) \quad \checkmark$$

triangle density:

$$m(\alpha) \equiv \int d\mu \varrho(\mu) \mu^3 = N^{-1} e^{6\alpha} \quad \checkmark$$

Degree- q triangularly constrained regular graphs

$$\hat{\varrho}(\mu) = \alpha \mu^3$$

- ▶ finite size corrections to spectrum:

$$\varrho(\mu) = \varrho_0(\mu) + \frac{1}{N} \varrho_1(\mu) + \frac{1}{N} \tilde{\varrho}_1(\mu) + o\left(\frac{1}{N}\right),$$

$\varrho_0(\mu)$: McKay formula

$\varrho_1(\mu)$: corrections for non-deformed measure

$\tilde{\varrho}_1(\mu)$: impact of promoting triangles

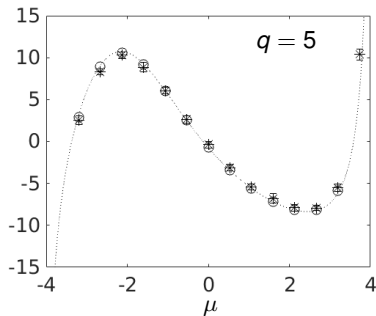
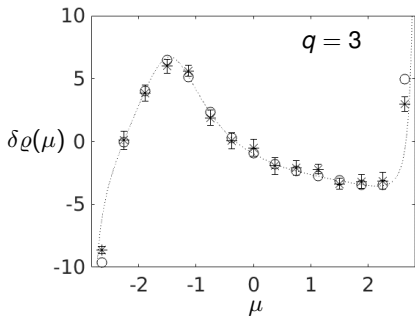
$$\tilde{\varrho}_1(\mu) = \frac{(q-1)^3}{2\pi} (e^{6\alpha} - 1) \frac{\theta[2\sqrt{q-1} - |\mu|]}{\sqrt{4(q-1) - \mu^2}} \left\{ \frac{q-2}{3} \left[\frac{2q + \mu}{q^2 - 3(q-1) + \mu q + \mu^2} + \frac{1}{q-\mu} \right] - 1 \right\}$$

- ▶ triangle density:

$$m(\alpha) = \frac{1}{N} (q-1)^3 e^{6\alpha} + o\left(\frac{1}{N}\right).$$

- ▶ accurate for $\alpha < \alpha_1(N)$,
theory implicitly assumes triangles are well separated
 $\alpha > \alpha_1(N)$: alternative saddle point?

► predictions versus simulations



$$\delta\rho(\mu) = N[\varrho(\mu) - \varrho_0(\mu)]$$

left: $q = 3$, $N = 1000$, $\alpha = 0.416$

right: $q = 5$, $N = 2000$, $\alpha = 0.431$

both examples:

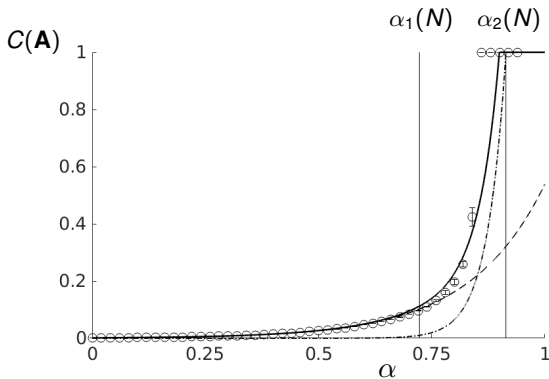
clustering coeff $\langle C(\mathbf{A}) \rangle \approx 0.02$

Beyond the region of validity

$\alpha > \alpha_1(N)$?

add contributions
from triangles
in cliques

here: $q = 3$



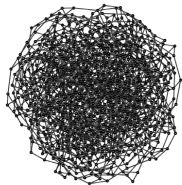
dashed: (i) present replica theory, isolated triangles

dotted-dashed: (ii) triangles in cliques

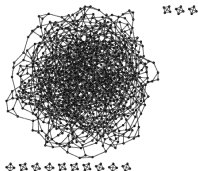
solid: (i)+(ii)

$\alpha_1(N)$ and $\alpha_2(N)$ grow logarithmically with N

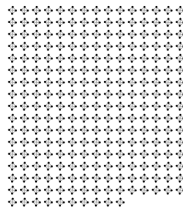
Qualitative picture for $q \geq 3$



connected
 $\alpha < \alpha_1(N)$



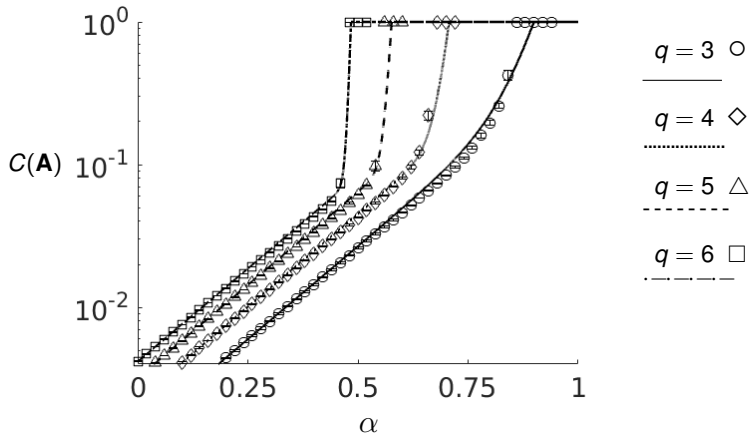
partially connected
 $\alpha_1(N) < \alpha < \alpha_2(N)$



disconnected
 $\alpha > \alpha_2(N)$

- ▶ *small* α :
isolated triangles embedded in giant component
- ▶ *intermediate* α :
triangles start touching, increasing nr of cliques of $q+1$ nodes
- ▶ *large* α :
graphs shattered fully into cliques of $q+1$ nodes
(graphs with max nr of triangles, ground state)

comparison to simulations with $N = 1000$
degree $q = 3 \dots 6$



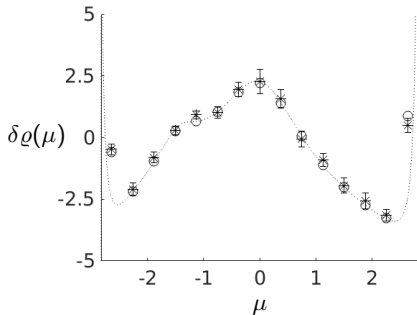
Other ensembles with $q = 3$

$$\rho(\mathbf{A}) = \frac{e^{\alpha \text{Tr}(\mathbf{A}^3) + \beta \text{Tr}(\mathbf{A}^4)}}{Z(\alpha, \beta)} \prod_{i=1}^N \delta_{3, \sum_j A_{ij}}, \quad Z(\alpha, \beta) = \sum_{\mathbf{A} \in G} e^{\alpha \text{Tr}(\mathbf{A}^3) + \beta \text{Tr}(\mathbf{A}^4)} \prod_{i=1}^N \delta_{3, \sum_j A_{ij}}$$

prediction:

$$\varrho(\mu) = \varrho_0(\mu) + \frac{1}{N} \varrho_1(\mu) + \frac{(q-1)^3}{2N} (e^{6\alpha} - 1) G_3(\mu) + \frac{(q-1)^4}{2N} (e^{8\beta} - 1) G_4(\mu) + o(N^{-1})$$

$N = 2000$,
 $\alpha = \beta = 0.2$

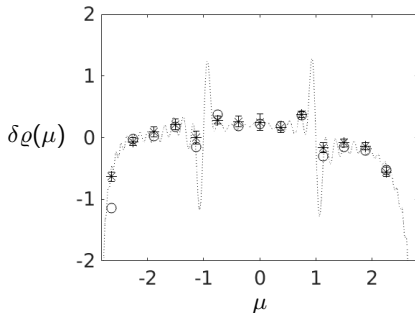


$$\rho(\mathbf{A}) = \frac{e^{\alpha \int_{-1}^1 d\mu \varrho(\mu|\mathbf{A})}}{Z(\alpha)} \prod_{i=1}^N \delta_{3, \sum_j A_{ij}}, \quad Z(\alpha, \beta) = \sum_{\mathbf{A} \in \mathcal{G}} e^{\alpha \int_{-1}^1 d\mu \varrho(\mu|\mathbf{A})} \prod_{i=1}^N \delta_{3, \sum_j A_{ij}}$$

prediction:

$$\varrho(\mu) = \rho_0(\mu) + \frac{1}{N} \rho_1(\mu) + \sum_{\ell \geq 3} \frac{(q-1)^\ell}{2N} \left[e^{\alpha \ell \int_{-1}^1 d\mu G_\ell(\mu)} - 1 \right] G_\ell(\mu),$$

$N = 1000,$
 $\alpha = 0.5$



simulations non-trivial ...

Summary

- ▶ analytical approach to (processes on) loopy random graphs, max entropy ensembles with constrained *degrees and spectrum*
- ▶ imaginary replica formula to handle spectral constraint:

$$e^{N \int d\mu \hat{\rho}(\mu) \varrho(\mu|\mathbf{A})} = \lim_{\varepsilon, \Delta \downarrow 0} \prod_{\mu} \left[Z(\mu + i\varepsilon|\mathbf{A})^{in(\mu)} \overline{Z(\mu + i\varepsilon|\mathbf{A})}^{-in(\mu)} \right]$$
$$Z(\mu|\mathbf{A}) = \int d\phi e^{-\frac{1}{2}i\phi \cdot [\mathbf{A} - \mu \mathbf{I}] \phi}, \quad n(\mu) = \frac{\Delta}{\pi} \frac{d}{d\mu} \hat{\rho}(\mu)$$

- ▶ RS eqns: reproduced tree-like formulae for $\hat{\rho}(\mu) \rightarrow 0$, look natural (message passing + nontrivial acceptance probs), but disagreed with simulations of loopy ones ...
- ▶ simple solvable toy model: $p(k) = \delta_{k2}$, $\hat{\rho}(\mu) = \sum_{\ell \geq 3} \alpha_{\ell} \mu^{\ell}$
finite nonzero loop *densities* require $\alpha_{\ell} = O(\log N)$
- ▶ imaginary replica theory correctly describes *finite size corrections* to spectra of *weakly loopy* graphs
- ▶ finite nonzero loop *densities*: $\hat{\rho}(\mu) = O(\log N)$, need to revisit steepest descent evaluation of path integrals ...

Papers, talks, seminars

<https://nms.kcl.ac.uk/ton.coolen>

Coolen, J. Phys. Conference Series 699 (2016)

Coolen, Annibale, Roberts, *Generating random networks and graphs* (2017)

Aguirre-Lopez, Barucca, Fekom, Coolen, J. Phys. A 51 (2018)

Aguirre-Lopez and Coolen, preprint arXiv:1907.06703 (2019)

January 2020



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