Group Walk Random Graphs

Agelos Georgakopoulos

WARWICK

Vancouver, 10.9.14

1269 papers on MathSciNet with "random graph" in their title

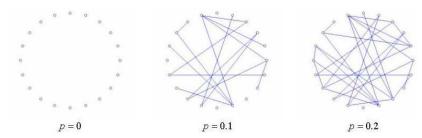
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... most of which on the Erdős-Renyi model G(n, p):

- n vertices
- each pair joined with an edge, independently, with same probability p = p(n).



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- 10. [Palásti, I. On the connectedness of random graphs. Studies in Math. Stat.: Theory & Applications. 1968]
- => gives a short summary of some previously published results concerning the connectedness of random graphs.

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100. [Bollobás, B. Long paths in sparse random graphs. Combinatorica. 1982]

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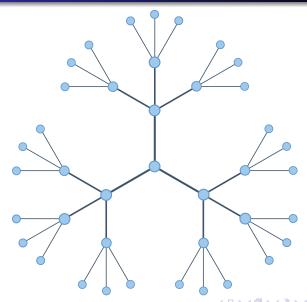
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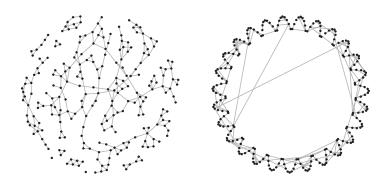
1000. [Doku-Amponsah, K.; Mörters, P. Large deviation principles for empirical measures of colored random graphs. Ann. Appl. Probab. 2010]

=> derives large deviation principles for the empirical neighbourhood measure of colored random graphs, defined as the number of vertices of a given colour with a given number of adjacent vertices of each colour. . . .

Random Graphs from trees

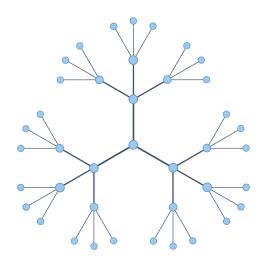


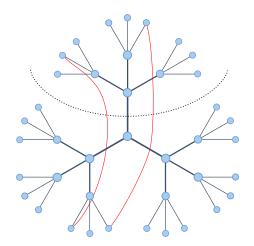
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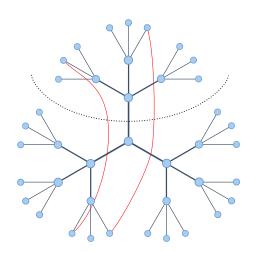


Simulation on the binary tree by A. Janse van Rensburg.



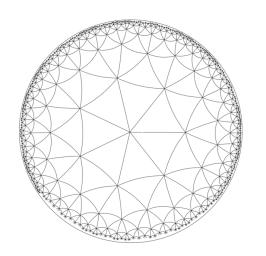






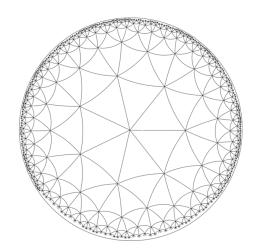
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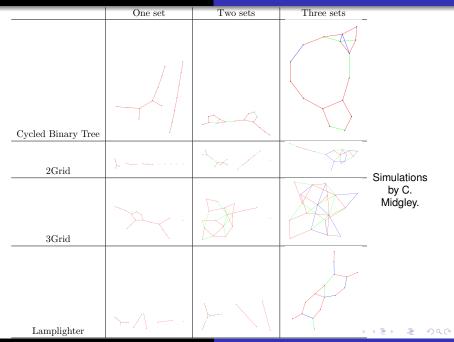


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Problems

Problem 1: The (expected) number of connected components (or isolated vertices) is asymptotically proportional to $|B_n|$.

Problem 2: The expected diameter of the largest component is asymptotically $c \log |B_n|$.

Backed by simulations by C. Midgley.

What's the point?

Metaproblem 1: Which properties of the random graphs are determined by the group of the host graph *H* and do not depend on the choice of a generating set?

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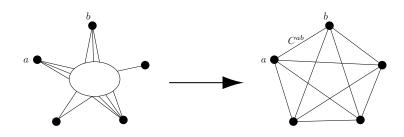
Metaproblem 2: Which group-theoretic properties of the host group are reflected in graph-theoretic properties of the random graphs?

Energy and Douglas' formula

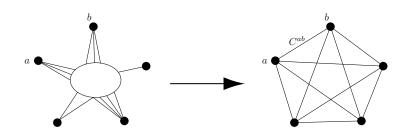
The classical Douglas formula

$$E(h) = \int_0^{2\pi} \int_0^{2\pi} (\hat{h}(\eta) - \hat{h}(\zeta))^2 \Theta(\zeta, \eta) d\eta d\zeta$$

calculates the (Dirichlet) energy of a harmonic function h on $\mathbb D$ from its boundary values $\hat h$ on the circle $\partial \mathbb D$.



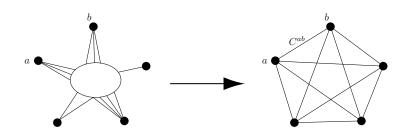
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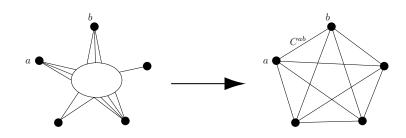


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How can we generalise this to an arbitrary domain? To an infinite graph?



The Poisson integral representation formula

The classical Poisson formula

$$h(z) = \int_0^1 \hat{h}(\theta) P(z, \theta) d\theta$$

where $P(z,\theta) := \frac{1-|z|^2}{|e^{2\pi i\theta}-z|^2}$, recovers every continuous harmonic function h on $\mathbb D$ from its boundary values $\hat h$ on the circle $\partial \mathbb D$.

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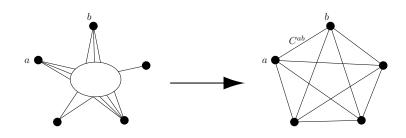
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- this $\hat{h} \in L^{\infty}(\mathcal{P}_G)$ is unique up to modification on a null-set;
- conversely, for every $\hat{h} \in L^{\infty}(\mathcal{P}_G)$ the function $z \mapsto \int_{\mathcal{P}_G} \hat{h}(\eta) d\nu_z(\eta)$ is bounded and harmonic.

i.e. there is Poisson-like formula establishing an isometry between the Banach spaces $H^{\infty}(G)$ and $L^{\infty}(\mathcal{P}_G)$.

Selected work on the Poisson boundary

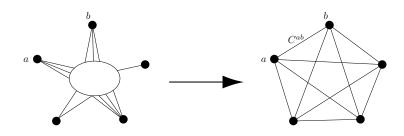
- Introduced by Furstenberg to study semi-simple
 Lie groups [Annals of Math. '63]
- Kaimanovich & Vershik give a general criterion using the entropy of random walk [Annals of Probability '83]
- Kaimanovich identifies the Poisson boundary of hyperbolic groups, and gives general criteria [Annals of Math. '00]



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[Doob '62] generalises this to Green spaces (or Riemannian manifolds) using their *Martin boundary*.



The energy of harmonic functions

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For every locally finite network G, there is a measure C on $\mathcal{P}^2(G)$ such that for every harmonic function u the energy E(u) equals

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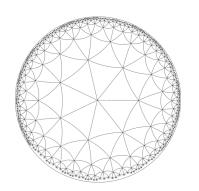


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$$C(X,Y) := \lim_n \mathbb{E}(\sharp \text{ edges } xy \text{ in } \mathcal{G}_n(H)$$
 with x 'close to' X , and y 'close to' Y)

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where the **Naim Kernel** Θ is defined as

$$\Theta(\zeta, \eta) := \frac{1}{G(o, o)} \lim_{z_n \to \zeta, y_n \to \eta} \frac{F(z_n, y_n)}{F(z_n, o) F(o, y_n)}$$

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Remark:

$$\frac{1}{\Theta(z, y)} = G(o, o) \Pr_{z}(o < y \mid y),$$

where $Pr_z(o < y|y)$ is the conditional probability to visit o before y subject to visiting y.

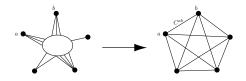
Convergence of the Naim Kernel

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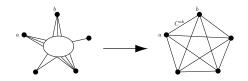
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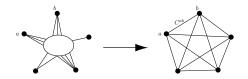
Problem: Let $(z_i)_{i\in\mathbb{N}}$ and $(w_i)_{i\in\mathbb{N}}$ be independent simple random walks from o. Then $\lim_{n,m\to\infty} \Theta(z_n,w_m)$ exists almost surely.



$$E(h) = \sum_{a,b \in B} (h(a) - h(b))^2 C_{ab},$$



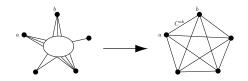
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Proposition

For every measurable $X, Y \subseteq \mathcal{P}(G)$

$$C_n(X, Y) = \mathbb{E}(\Theta^n(x_n, y_n) \mathbf{1}_{XY}).$$

Therefore, $C(X, Y) = \lim_n \mathbb{E}(\Theta^n(x_n, y_n) \mathbf{1}_{XY}).$



Random Interlacements *I* [Sznitman]:

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Claim:
$$C(X, Y) = v(1_{XY}W^*)$$
.



The effective conductance measure C, The Naim kernel Θ , Random Interlacements I, and Group Walk Random Graphs $\mathcal{G}_n(H)$

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Can we use them to study groups?

