## On the expected size of a patriarchal mafia with Poisson distributed collaborations

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Rome, 23/2/17
Partly joint work with J. Haslegrave
Disclaimer: The topic of this talk has nothing to do with the location

## How Mafia's grow

A network evolves in (continuous or discrete) time with the following rules:

- When a (Poisson) clock ticks, nodes split into two;
- When a node $x$ splits into two nodes $x^{\prime}, x^{\prime \prime}$, each of its existing edges gets inherited by $x^{\prime}$ or $x^{\prime \prime}$ independently with probability $1 / 2$;
- Moreover, a Poisson(k)-distributed number of new edges are added between $x^{\prime}$ and $x^{\prime \prime}$.


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If it is finite, is its expected size finite or infinite?
If finite, how does it depend on $k$ ?

## Random Graphs

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Real-world networks?

- Preferential attachment networks
- Geometric random graphs


## Geometric Random Graphs Literature

[Remco Van Der Hofstad. Random graphs and complex networks. Lecture Notes, 2013.]
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## Random Graphs from trees



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## Proposition

For every two measurable subsets $X, Y$ of the Poisson (or Martin) boundary $\partial G$,
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## converges.

We use the limit to define a measure on $\partial G \times \partial G$ via

$$
C(X, Y):=\lim \mathbb{E}(\# \text { edges } \ldots)
$$

## Energy and Douglas' formula

The classical Douglas formula [Douglas '31]

$$
E(h)=\int_{0}^{2 \pi} \int_{0}^{2 \pi}(\hat{h}(\eta)-\hat{h}(\zeta))^{2} \Theta(\zeta, \eta) d \eta d \zeta
$$

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

## Effective conductance

We call $C$ the effective conductance measure, because

## Theorem (G \& V. Kaimanovich '12-'17+)

For every locally finite network $G$, and every harmonic function $h$, we have

$$
E(h)=\int_{\partial G \times \partial G}(\widehat{h}(\eta)-\widehat{h}(\zeta))^{2} d C(\eta, \zeta) .
$$

History: Douglas '31, Naim '57, Doob '62, Silverstein '74
Finite version: $E(h)=\sum_{a, b \in B}(h(a)-h(b))^{2} C_{a b}$


## Random Interlacements and C

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Theorem (G \& Kaimanovich '17+)
For every transient, locally finite graph $G$,

$$
C(X, Y)=v\left(1_{X Y} W^{*}\right)
$$

## Long range percolation

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In long range percolation on $\mathbb{Z}$, with edge rates $\lambda /|x-y|^{s}$, percolation occurs for large enough $\lambda$ if $s \leq 2$.

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$$
\text { How large is } R_{\infty}^{\lambda}(T) ?
$$

## The expected size of the TWRG

Let $C_{o}^{\lambda}$ denote the component of a uniformly random vertex of $R_{n}^{\lambda}(T)\left(\operatorname{or} R_{\infty}^{\lambda}(T)\right)$.

Theorem (G \& Haslegrave, state of the art 2/17)

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e^{\lambda} \leq \mathbb{E}\left(\left|C_{o}^{\lambda}\right|\right) \leq e^{e^{b \lambda}}
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Conjecture:

$$
\mathbb{E}\left(\left|C_{o}^{\lambda}\right|\right) \sim \lambda^{\lambda}
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(backed by simulations)

## Outlook

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## Thank you!

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