

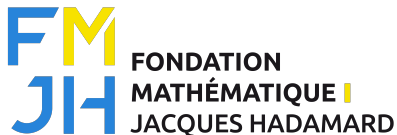


A Look at Limits in Random Graphs

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- ① Convergence
- ② Local Topology
- ③ Pointed Graphs
- ④ Interlude
- ⑤ Rooted Graphs
- ⑥ Main Idea (and yes, SGT)

Benefactors



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- 1 Convergence
- 2 Local Topology
- 3 Pointed Graphs
- 4 Interlude
- 5 Rooted Graphs
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The Prairie

Definition

If $x_n = \{x_1, x_2, x_3, x_4, \dots\}$ is a sequence, then

$$\lim_{n \rightarrow \infty} x_n = x$$

if for every $\epsilon > 0$ there exists some $N > 0$ such that

$$|x_n - x| < \epsilon \text{ when } n \geq N \text{ for every natural number } n$$

i.e. under certain conditions your sequence will get arbitrarily *close* to some limiting object

The Prairie

Definition (?)

If $G_n = \{\text{---}, \triangle, \text{---}, \text{---}\}$ is a sequence, then

$$\lim_{n \rightarrow \infty} G_n = \text{---}$$

if for every $\epsilon > 0$ there exists some $N > 0$ such that

$$|G_n - \text{---}| < \epsilon \text{ when } n \geq N \text{ for every natural number } n$$

i.e. What do we want arbitrarily *close* to mean in the space of graphs? What conditions would we then require for a limit to exist?

The Treeline

Theorem

If $f_n = \{f_1, f_2, f_3, f_4, \dots\}$ is a sequence of continuous functions such that f_n converges uniformly to f , then f is continuous.

i.e. under certain conditions your sequence will get arbitrarily *close* to some limiting object *and* you know something about the object

The Treeline

Theorem (?)

If $f_n(G_n) = \{f_1(\text{edge}), f_1(\text{triangle}), f_2(\text{path of length 2}), f_3(\text{star}), \dots\}$ is a sequence of _____ functions such that $f_n(G_n)$ converges _____ to $f(\text{graph})$, then $f(\text{graph})$ is _____.

i.e. After we have a notion of arbitrary *closeness*, what properties can we infer about limiting objects? What restrictions do we need for the convergence if any?

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Pointed vs. Rooted

Definition

A *pointed* graph is a pair $g^\bullet = (g, \rho)$ where $\rho \in V(g)$ is a distinguished node. Let $\mathcal{G}^\bullet$ denote the set of (locally finite, connected, countable) pointed graphs up to isomorphism.

Definition

A *rooted* graph is a pair $\vec{g} = (g, \vec{e})$ where $\vec{e} \in E(g)$ is a distinguished edge. Let $\vec{\mathcal{G}}$ denote the set of (locally finite, connected, countable) rooted graphs up to isomorphism.

Defining sequences that converge considering the *global* geometry of the graphs is relatively straight forward and boring. We care about *local* geometry of graphs i.e. subgraphs around a root node ρ , $B_r(G, \rho)$. Allowing us to define distances:

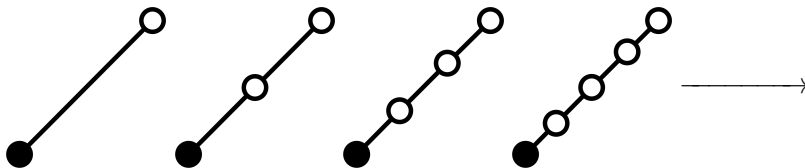
- Are the building blocks of radius r in two graphs different?

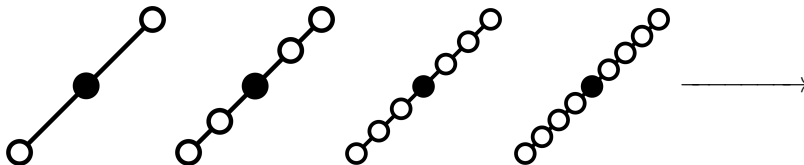
$$\mathbb{1}_{B_r(G, \rho) \not\cong B_r(G', \rho')}$$

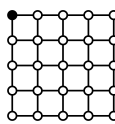
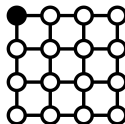
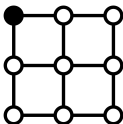
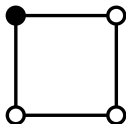
- If so, they contribute 2^{-r} to the distance.
- Look at all blocks of radius r :

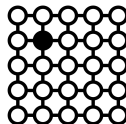
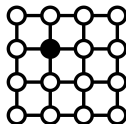
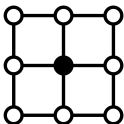
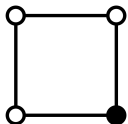
$$d_{loc}(G, G') = \sum_{r \geq 0} \min(1, \mathbb{1}_{B_r(G, \rho) \not\cong B_r(G', \rho')}) 2^{-r}$$

So, a sequence of graphs G_n converges if the local structure around the root stabilizes. What it stabilizes to determines the limiting object.









Equivalent Definition of Convergence

- So, we have a way of defining convergence locally.
- Using the local topology, we can take a fixed graph sequence and find its limit as n grows.
- But, we'd like to do the same with random graph sequences too!

Definition

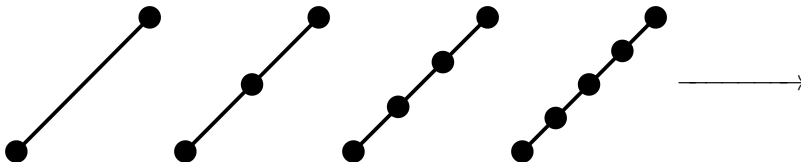
Let g be some finite graph. The *uniformly pointed version* of g is the random graph, $U^\bullet(g)$, whose law is $\frac{1}{|V|} \sum_{x \in V(g)} \delta_{(g,x)}$

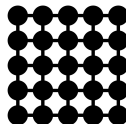
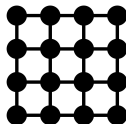
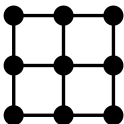
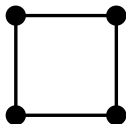
Equivalently,

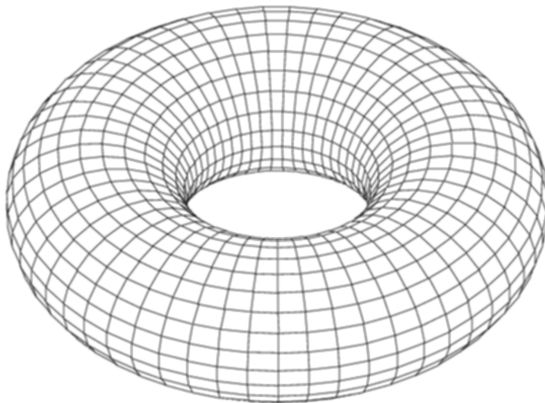
$$\mathbb{E}[f(U^\bullet(g))] = \frac{1}{|V|} \sum_{x \in V(g)} f(g, x)$$

(for any measurable $f : \mathcal{G}^\bullet \rightarrow \mathbb{R}$)

That is, $U^\bullet(g)$ is the random graph obtained by considering neighborhoods of uniformly chosen nodes. In other words, expressing what the “average” local geometry behaves like







Benjamini-Schramm Convergence

Definition

A sequence of unpointed random finite graphs G_n converges in the Benjamini-Schramm sense towards $G_\infty^\bullet = (G_\infty, \rho_\infty)$ if under d_{loc} , $U^\bullet(G_n)$ converges in distribution to $G_\infty^\bullet$.

Equivalently if for any bounded continuous $\phi : \mathcal{G}^\bullet \rightarrow \mathbb{R}_+$,

$$\mathbb{E} \left[\frac{1}{|V|} \sum_{x \in V(G_n)} \phi(G_n, x) \right] \rightarrow \mathbb{E}[\phi(G_\infty^\bullet)]$$

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Unimodularity

Definition

A random pointed graph (G, ρ) is *unimodular* if

$$\mathbb{E} \left[\sum_{x \in V(G)} f(G, \rho, x) \right] = \mathbb{E} \left[\sum_{x \in V(G)} f(G, x, \rho) \right]$$

(for any $f : \mathcal{G}^{\bullet\bullet} \rightarrow \mathbb{R}_+$)

That is, the average mass a distinguished node receives equals the average mass it sends.

Stability

Theorem

Let G_n be a sequence of graphs that converges in the Benjamini-Schramm sense. If $U^\bullet(G_n)$ is unimodular and converges to $G_\infty^\bullet$, then $G_\infty^\bullet$ is also unimodular.

Why is unimodularity nice?

We can pick $f(G, \rho, x)$!

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Defining the SRW

Simple random walk:

- Take a graph g .
- Pick a vertex $x \in V(g)$ to start at.
- Looking at the neighbors of x in g , pick one uniformly at random.
- Move to that vertex.
- Repeat previous two steps.

Stationarity and Reversibility

Definition

Let $\vec{G} = (\vec{G}, \vec{e})$ be a random rooted graph. Let $\{\vec{E}_i\}_{i \geq 0}$ denote the sequence of visited edges during a simple random walk on $\vec{G}$.

- If the law of $(\vec{G}, \vec{E}_k)$ is the same as $(\vec{G}, \vec{e})$ for every $k \geq 0$, $\vec{G}$ is *stationary*.
- If the law of $(\vec{G}, \vec{E}_0)$ is the same as $(\vec{G}, \vec{E}_0)$, $\vec{G}$ is *reversible*.

Stability

Theorem

Let G_n be a sequence of stationary and reversible random graphs which converges to $\vec{G}_\infty$ in the Benjamini-Schramm sense. Then $\vec{G}_\infty$ is stationary and reversible.

Correspondence!

Theorem

Let $G^\bullet = (G, \rho)$ be a pointed unimodular random graph. Let the rooted graph $\vec{G} = (G, \rho_e)$ denote the graph obtained by choosing an edge on ρ uniformly at random. Biasing by the degree of ρ_e returns a stationary and reversible random graph.

Theorem

Let $\vec{G} = (G, \vec{e})$ be a rooted stationary and reversible random graph. Let $G^\bullet = (G, \vec{e}_*)$ denote the pointed graph obtained by distinguishing the origin node of $\vec{e}$, $\vec{e}_*$. Biasing by the inverse degree of $\vec{e}_*$ returns a unimodular random graph.

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The Framework

A major portion of this body of work can be summarized as follows:

- Find a Benjamini-Schramm limit for a sequence of graphs and find its law.
- Pick your favorite graph parameter.
- Determine whether we can use the law of the limiting object to find the limit of the graph parameter on our sequence i.e. for a graph parameter $\phi(G_n)$, $\mathbb{E}[U^\bullet(G_n)] \rightarrow \mathbb{E}[G_\infty^\bullet] \implies$ the limit of $\phi(G_n)$ exists and can be found using the law of $G_\infty^\bullet$

Participating Properties

Among the graph parameters ϕ that we know this is true are:

- number of spanning trees
- size of maximum matching
- rank of adjacency matrix
- maximal subgraph density
- spectral statistics!

Spectral Convergence

Theorem

If G_n converges to $G_\infty^\bullet$ in the Benjamini-Schramm sense where μ_{G_n} and μ_{G_∞} denote the corresponding laws, then

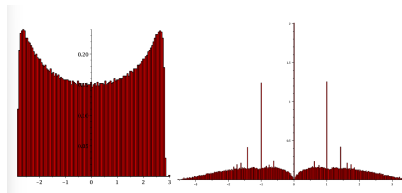
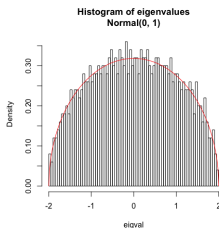
$$\sup_{\lambda \in \mathbb{R}} |\mu_{G_n}((-\infty, \lambda]) - \mu_{G_\infty}((-\infty, \lambda])| \rightarrow 0$$

as $n \rightarrow \infty$

In other words, we can know that the spectral distribution concentrates and (oftentimes) what it concentrates to!

Utility

This is particularly helpful since theorems like Wegner's semi-circle law only apply to dense graphs. With the help of Benjamini-Schramm, we can know more about larger regimes of graphs including (most importantly) sparse ones.



End

- Thanks!
- garrett.kepler@wsu.edu

References

- “Spectra of sparse random graphs” by Justin Salez link: [here](#).
- “Eigenvalues of random matrices...” by Kenneth Tay link: [here](#).
- “A random walk among random graphs” Nicolas Curien link: [here](#).