Local algorithms on random graphs and graph limits

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### Dense graph limit

(by Lovász, Szegedy, Borgs, Chayes, Sós, Vesztergombi et al.)

Sampling from a graph: we choose a constant number of uniform random vertices and observe the spanned subgraph.

Accordingly, we say that a sequence of graphs  $(G_n)$  is convergent if for all  $k \in \mathbb{Z}$ , the probability distribution of the subgraphs spanned by k uniform random vertices  $s_k(G_n)$  is convergent for  $n \to \infty$ .

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Graphon is a symmetric measurable function  $W : [0,1] \times [0,1] \rightarrow [0,1]$ . Graphs  $G \in \{0,1\}^{n \times n}$  are represented by  $W(x,y) = G(\lfloor nx \rfloor, \lfloor ny \rfloor)$ .

The topology on the set of graphons are defined by the topology induced by  $s_k(W)$ , or by the metric

$$\inf_{\substack{P_1,P_2: [0,1] \to [0,1] \text{ bijection,} \\ P_1^{-1},P_2^{-1} \text{measure-preserving}}} \sup_{\substack{X,Y \subset [0,1] \\ \text{measurable}}} \int_{X \times Y} |W_1(x,y) - W_2(x,y)| \, \mathrm{d}(x,y).$$

Theorem. The two definitions are equivalent.

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- ▶ Let  $N_G(F)$  = number of labelled copies of G' in G, and t(F, G) the homomorphism densities. Theorem (Chung, Graham, Wilson):

 $\forall p \in [0,1], \ \forall F \in \mathcal{G}, \ \forall \varepsilon > 0, \ \exists \delta > 0:$ 

if  $N_G(P_1) \ge pn^2$  but  $N_G(C_4) \le (1 + \delta)pn^4$ , then  $N_G(F) \in (1 \pm \varepsilon)p^{V(F)}$ 

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1

If  $t(C_4, W) = t(P_1, W)^4$ , then W is constant.

## Sparse graph limits

### (by Benjamini, Schramm, Elek, Hatami, Lovász, Szegedy, Gamarnik, Sudan et al.)

Sampling from a graph with degrees bounded by d: we choose a constant number of uniform random vertices v and observe  $B_r(v)$ , the constant-radius neighborhood of v.

Accordingly, we say that a sequence of graphs  $(G_n)$  with degree bound d is convergent if for all  $r \in \mathbb{Z}$ , the distribution of  $B_r(G_n)$  is convergent for  $n \to \infty$ .

#### Graph limits

 Probability distributions on rooted connected (finite or) infinite trees. Necessary: unimodularity (some consistency condition).
 Aldous-Lyons: is it sufficient? (Elek: probably not.)
 Theorem (Cs). The question whether there is a unimodular random rooted graph supported on a set of neighborhoods is undecidable. The same applies for limits of finite graphs. Sampling from a graph with degrees bounded by d: we choose a constant number of uniform random vertices v and observe  $B_r(v)$ , the constant-radius neighborhood of v.

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### Graph limits

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- Graphing or measurable graph is a finite union of measure-preserving partial bijections on a measurable vertex set. It is a richer structure: Theorem. Every convergent sequence of graphs local-globally tends to a graphing.

### Local-global limit and local algorithms

For a finite set of colors  $C = \{1, 2, ..., c\}$ , consider all possible vertex colorings of a graph G. Consider the colored *r*-neighborhood distributions, let us call now them structures. These define the local-global topology on graphs. For graphings, we want measurable colorings.

Lemma. For each unimodular random graph, there exists a graphing (Bernoulli-graphing) with the weakest possible structure: only those structures which can be made by local algorithms.

Questions:

- Is there a graph sequence local-globally converging to each graphing?
- ► Is the Bernoulli-graphing the local-global limit of a graph sequence?
- Is the sequence of random graphs asymptotically the least structured?

Theorem. (Gamarnik, Sudan) For large d, random graphs have larger independence ratio as what can be constructed by any local algorithm.

# Local (graph) algorithms

Local algorithm is a function  $I: [0,1]^{V(B_r(v))} \to C$ . We assign a random variable q to each vertex (or edge, etc.) and the output at v is  $f(q(\mathcal{B}_r(v)), [g, B_r(G)])$ .

- ► g is a global randomization
- $B_r(G)$  is the exact statistics of r-neighborhoods

Theorem. (Cs) Access to  $B_r(G)$  does not help.

Theorem. (Bollobás) A random 3-regular graph has an independence ratio < 0.46. (Later improved to 0.455.) Rephrased theorem. Let  $H(\cdot)$  denote the entropy of the output at a random vertex, and H(-) is the entropy on two neighboring vertex. Then

$$H(-) \geq \frac{4}{3}H(\cdot).$$

Theorem. (Cs) We can construct an independence ratio 0.445 by a local algorithm.

### Entropy bounds for local algorithms

Reveal the random seeds one by one. The Shapley–Shannon information i(x, v) of a seed s(x) to the output c(v) is the expected mutual information between them when the seed is revealed. We know that for neighboring vertices v, w,

$$i(x, v) \ge 0$$
  

$$H(c(v)) = \sum_{x} i(x, v)$$
  

$$H(c(v), c(w)) \ge \sum_{x} \max(i(x, v), i(x, w)).$$

All negative results about local algorithms are followed by such inequalities (+ graph automorphisms).

If we exchange the entropy function to other functions, we can get correlation-inequalities and other bounds.