

1 Introduction

We have been discussing the Gibbs measure on graphs $\nu(G) = e^{\beta \cdot \Delta(G)} \mu(G)$, where $\mu = \mathcal{G}(n, 1/2)$ and $\Delta(G)$ is the number of triangles in G (or $3 \times$ that quantity). So we'd like to understand triangle counts in general graphs to see where this measure lives.

More generally, we'd like to understand the count of any fixed subgraph. In these notes, we'll introduce a class of models where the subgraph counts are easy to understand, and then prove that *every* graph looks like a graph from one of these models, in the sense that the subgraph counts are similar. The main reference for these notes is Lovász's book "Large Networks and Graph Limits" [2], mostly chapters 7-10.

1.1 Subgraph densities

To set some notation, for any fixed finite graph F , let $V(F)$ and $E(F)$ denote the vertex set and edge set respectively. If G is another graph, we define the *subgraph density of F in G* as

$$t(F, G) = \mathbb{P}[|V(F)| \text{ random vertices in } G \text{ form a copy of } F].$$

Here, " $|V(F)|$ random vertices" means, for each $v \in F$, sampling $X_v \in V(G)$ uniformly, independently for different v . These vertices "form a copy of F " if, for each $uv \in E(F)$, $X_u X_v \in E(G)$. Note that this means the subgraph F *does not need to be induced* by the vertices.

A word of warning: for some graphs F , it is not easy to see how $t(F, G)$ relates to the actual subgraph *count* of F in G . For instance, $t(\square, -) = \frac{1}{8}$, but a single edge has no squares as subgraphs. However, when $|V(G)| \gg |V(F)|^2$ so that there is little chance of sampled vertices being the same, we have

$$t(F, G) \approx \frac{\text{number of labeled copies of } F \text{ in } G}{|V(G)|^{|V(F)|}} = |V(F)|! \frac{\text{number of unlabeled copies of } F \text{ in } G}{|V(G)|^{|V(F)|}}.$$

1.2 Stochastic block model

A stochastic block model S is a generalization of the $\mathcal{G}(n, p)$, which, for some $k \geq 1$, is given via a partition $Q = (q_1, \dots, q_k)$ of unity (a probability measure on $\{1, \dots, k\}$), and a $k \times k$ matrix $P = (p_{ij})$.

We can sample a graph on n vertices from S as follows; we'll call it $G(n, S)$, and the distribution will be called $\mathcal{G}(n, S)$. First, assign each $v \in V(G(n, S)) = \{1, \dots, n\}$ to a class x_v in $\{1, \dots, k\}$ independently according to the partition Q . Then, for each $u, v \in V(G(n, S))$, add an edge uv to $E(G(n, S))$ with probability $P_{x_u x_v}$, independently for each pair of vertices.

It is easy to estimate $t(F, G(n, S))$. By an application of Fubini's theorem, we have

$$\lim_{n \rightarrow \infty} \mathbb{E}[t(F, G(n, S))] = \mathbb{P}[G(|V(F)|, S) \text{ contains a copy of } F].$$

The right-hand side is equal to

$$\sum_{(x_v) \in \{1, \dots, k\}^{V(F)}} \prod_{v \in V(F)} q_{x_v} \prod_{uv \in E(F)} P_{x_u x_v}.$$

Overloading our notation, we will denote the above quantity by $t(F, S)$. In fact, $t(F, G(n, S)) \rightarrow t(F, S)$ almost surely. This follows from concentration results for Lipschitz functions, but we won't go into the details since we don't need this fact.

1.3 The (weak) regularity lemma

The regularity lemma states that every graph G looks like it could have been sampled from some stochastic block model S , in the sense that $t(F, G) \approx t(F, S)$ for all small enough finite graphs. Of course, this would be trivial if we allowed the number k of partitions in the SBM to depend on the graph; just make a different part for each vertex of G and set the edge probabilities to be 1 if the corresponding edge is present in G , and 0 otherwise; this gives an SBM S with $t(F, S) = t(F, G)$ for all F (if this last part is not clear now, it will become clear later in the notes).

The strength of the regularity lemma is the fact that the error in the approximation $t(F, G) \approx t(F, S)$ only depends on the number of parts k in the SBM, and not on the graph G . Here's the statement.

Theorem 1 (weak regularity lemma I). *For any $k \geq 4$ and any graph G , there is a stochastic block model S with k parts such that*

$$|t(F, G) - t(F, S)| < \frac{2|E(F)|}{\sqrt{\log k}}$$

for every finite graph F .

A brief word on why this is called the “weak” regularity lemma is in order. The original regularity lemma, due to Szemerédi in 1975 [3], states that for every $\varepsilon > 0$, there is some number $S(\varepsilon) \in \mathbb{N}$ such that every graph $G = (V, E)$ has an *equitable partition* $V = V_1 \sqcup \dots \sqcup V_k$ (meaning the pieces differ in size by at most 1) with $k \leq S(\varepsilon)$, and such that the bipartite graph between V_i and V_j is ε -homogeneous, for all but εk^2 pairs of indices i, j . Here ε -homogeneity means that every subgraph has close to “the right number of edges”, and ε measures the error tolerance.

This is stronger than Theorem 1 for a few reasons: first, in Theorem 1 there is no guarantee of an equitable partition. Additionally, ε -homogeneity is a slightly stronger condition and it implies that the subgraph densities are close to what they should be. However, the original regularity lemma is much more difficult to prove. Moreover the number $S(\varepsilon)$ is humongous: it is a power tower $2^{2^{\dots}}$ of height about $1/\varepsilon^2$.

The weaker regularity lemma, Theorem 1, has a much more reasonable partition size of about $\exp(2/\varepsilon^2)$ (of course, the ε means something different here). And, more importantly for us, it is easier to prove. Additionally, the formulation deals more directly with subgraph densities, the quantity of interest to us, although we will need to make a detour through some more analytic formulations in order to prove it. Theorem 1 is due to Frieze and Kannan in 1999 [1].

2 Graphons

It is much easier to understand and prove the weak regularity lemma if we use the language of graphons, which are typically thought of as a form of *graph limit*, although this idea will not be too relevant for us.

2.1 Definitions

Simply put, a graphon is a symmetric measurable function $W : [0, 1]^2 \rightarrow [0, 1]$. Here “symmetric” means that $W(x, y) = W(y, x)$. Every graph can be turned into a graphon in a canonical (though not bijective) way, and from every graphon one can sample a finite graph, in a way which generalizes the SBM.

To turn a graph G into a graphon W_G , first split up the interval $[0, 1]$ into pieces Q_v for $v \in V(G)$, with equal measure, and then define

$$W_G = \sum_{uv \in E(G)} \mathbf{1}_{Q_u \times Q_v}.$$

In words, this is a matrix plot of the adjacency matrix of G . Note that the map $G \mapsto W_G$ is not bijective, since subdividing each Q_v into two, for instance, gives $W_{G^\boxtimes} = W_G$. Additionally, since we consider graphs to be isomorphic when the vertices are relabeled, we consider graphons to be isomorphic when the points of $[0, 1]$ are relabeled by a *measure-preserving transformation* (which does not need to be invertible (!)). This is why the ambiguity in the definition of Q_v above is okay, but in general we won't worry too much about this notion of isomorphism in these notes.

To sample a graph $G(n, W)$ from a graphon W , sample n points $x_1, \dots, x_n \in [0, 1]$ uniformly and independently, and add each edge uv to $E[G(n, W)]$ with probability $W(x_u, x_v)$ independently of one another. This is completely analogous to the stochastic block model, and indeed a stochastic block model with k parts can be thought of as a special kind of graphon which is constant on k^2 sets $Q_i \times Q_j$, for some partition $Q = (Q_1, \dots, Q_k)$ of $[0, 1]$. From here on out, we call this type of graphon a *stepfunction with k steps*.

We can generalize the subgraph density of a SBM to a graphon as follows:

$$t(F, W) = \mathbb{P}[G(|V(F)|, W) \text{ contains } F] = \int_{[0,1]^{V(F)}} \prod_{uv \in E(F)} W(x_u, x_v) \prod_{v \in V(F)} dx_v.$$

As before, we have $t(F, G(n, W)) \rightarrow t(F, W)$ almost surely. Also note that for any finite graph G , we have $t(F, G) = t(F, W_G)$. This hints that we can generalize Theorem 1 to graphons.

Theorem 2 (weak regularity lemma II). *For any $k \geq 4$ and any graphon W , there is a stepfunction S with k steps such that*

$$|t(F, W) - t(F, S)| < \frac{2|E(F)|}{\sqrt{\log k}}$$

for every finite graph F .

2.2 Cut metric

Now we have put the two objects W and S in the statement of the weak regularity lemma on equal footing (they are both graphons). But to actually prove this, we will need to introduce a metric which induces the same topology on graphons as the one where $W_n \rightarrow W$ if $t(F, W_n) \rightarrow t(F, W)$ for all F .

First, for any symmetric function $W : [0, 1]^2 \rightarrow \mathbb{R}$, we define

$$\|W\|_{\square} = \sup_{S, T \subseteq [0, 1]} \left| \int_{S \times T} W(x, y) dx dy \right|.$$

This is called the *cut norm*, since it measures the maximal “edge weight” between two sets of “vertices.”

One way to measure the distance between two graphons W and U would be to calculate $\|W - U\|_{\square}$. However, this does not account for the fact that graphons should be considered isomorphic under relabeling $[0, 1]$ by a measure-preserving transformation. The fix is simple enough: define the *cut metric*

$$\delta_{\square}(W, U) = \inf_{\substack{\varphi: [0, 1] \rightarrow [0, 1] \\ \text{measure-preserving}}} \|W - U^{\varphi}\|_{\square},$$

where U^{φ} is the graphon U under relabeling by φ , i.e. $U^{\varphi}(x, y) = U(\varphi(x), \varphi(y))$. The “min-max” feature in the cut metric can make it a bit tough to reason about. Luckily, in these notes, as we will see, we don’t really need to worry about δ_{\square} ; we will only need to work with the cut norm $\|\cdot\|_{\square}$.

2.2.1 Example: Erdős-Rényi graph converges to “one-half” graphon

To help make sense of this, let’s see (at least heuristically) why $\delta_{\square}(W_{G(n, 1/2)}, \frac{1}{2}) \rightarrow 0$, which means that in the topology of the cut metric, the Erdős-Rényi graph converges to the graphon which is the constant $\frac{1}{2}$ on the unit square. It suffices to show that $\|W_{G(n, 1/2)} - \frac{1}{2}\|_{\square} \rightarrow 0$ for any realization of $W_{G(n, 1/2)}$; let’s just fix an ordering on the vertices and say that vertex i corresponds to the interval $Q_i = [\frac{i-1}{n}, \frac{i}{n}]$.

For any $S, T \subseteq [0, 1]$ of positive measure, $W_{G(n, 1/2)} - \frac{1}{2}$ consists of a bunch of tiny squares with value $\pm \frac{1}{2}$, with the sign being decided independently for each square (ignoring the symmetry across the diagonal, which can be easily accounted for). So, by some uniform law of large numbers,

$$\int_{S \times T} \left(W_{G(n, 1/2)}(x, y) - \frac{1}{2} \right) dx dy = o(1)$$

uniformly among all $S, T \subseteq [0, 1]$.

This explains why it is very important that the absolute value bars are on the *outside* of the integral. If they were on the inside we would just recover the L^1 norm, and $\|W_{G(n,1/2)} - \frac{1}{2}\|_1 = \frac{1}{2}$ for every n . Of course, by the Triangle inequality and Jensen's inequality, we have

$$\|W\|_{\square} \leq \|W\|_1 \leq \|W\|_2 \leq \|W\|_{\infty}.$$

These comparisons are not so important for us in these notes, but we will make use of the L^2 norm of a graphon in order to prove Theorem 2.

2.2.2 Functional equivalence, and maximum is obtained

Finally, we present a technical result about the cut norm which we will need for later proofs.

Lemma 3. *For any symmetric function $W : [0, 1]^2 \rightarrow \mathbb{R}$,*

$$\|W\|_{\square} = \sup_{S, T \subseteq [0, 1]} \left| \int_{S \times T} W(x, y) dx dy \right| = \sup_{f, g : [0, 1] \rightarrow [0, 1]} \left| \int_{[0, 1]^2} f(x)g(y)W(x, y) dx dy \right|,$$

and both suprema are attained (i.e. they are maxima).

Proof sketch. The supremum over functions is attained because the set of functions $[0, 1] \rightarrow [0, 1]$ is compact, and the given integral is a continuous function of f and g , both statements holding for the weak* topology. Additionally, the maximizing functions can be chosen to be $\{0, 1\}$ -valued, e.g. by replacing the maximizing f by $\mathbf{1}_{\{f>0\}}$ and similarly for g . ■

2.3 Counting Lemma

As alluded to earlier, $\delta_{\square}(W_n, W) \rightarrow 0$ if and only if $t(F, W_n) \rightarrow t(F, W)$ for every finite graph F . The counting lemma is one half of this equivalence

Lemma 4 (counting lemma). *Let F be a simple graph and let U, W be graphons. Then*

$$|t(F, W) - t(F, U)| \leq |E(F)| \cdot \|W - U\|_{\square}.$$

Note that we can get away with $\|W - U\|_{\square}$ in the right-hand side, instead of $\delta_{\square}(W, U)$, since the left-hand side does not depend on the “coupling” between W and U , i.e. it remains fixed if we permute only W (and not U) by some measurable map. So we can just take the minimum over measurable maps permuting W to obtain

$$|t(F, W) - t(F, U)| \leq |E(F)| \cdot \delta_{\square}(W, U).$$

Proof of counting lemma. By the definition of $t(F, W)$, we have

$$|t(F, W) - t(F, U)| = \left| \int_{[0, 1]^{V(F)}} \left(\prod_{uv \in E(F)} W(x_u, x_v) - \prod_{uv \in E(F)} U(x_u, x_v) \right) \prod_{v \in V(F)} dx_v \right|.$$

Fix some ordering on $E(F)$, and let $u_i v_i$ denote the i th edge. Then the difference in the interior of the integral can be written as a telescoping sum:

$$\begin{aligned} \prod_i W(x_{u_i}, x_{v_i}) - \prod_i U(x_{u_i}, x_{v_i}) &= \sum_{j=1}^{|E(F)|} \left(\prod_{i \leq j} W(x_{u_i}, x_{v_i}) \prod_{i > j} U(x_{u_i}, x_{v_i}) - \prod_{i < j} W(x_{u_i}, x_{v_i}) \prod_{i \geq j} U(x_{u_i}, x_{v_i}) \right) \\ &= \sum_{j=1}^{|E(F)|} \prod_{i < j} W(x_{u_i}, x_{v_i}) \prod_{i > j} U(x_{u_i}, x_{v_i}) (W(x_{u_j}, x_{v_j}) - U(x_{u_j}, x_{v_j})). \end{aligned}$$

So, using the triangle inequality and Fubini's theorem, we can write

$$|t(F, W) - t(F, U)| \leq \sum_{j=1}^{|E(F)|} \left| \int_{[0,1]^2} f_j(x_{u_j}) g_j(x_{v_j}) (W(x_{u_j}, x_{v_j}) - U(x_{u_j}, x_{v_j})) dx_{u_j} dx_{v_j} \right|, \quad (1)$$

where

$$f_j(x_{u_j}) = \int_{[0,1]^{V(F) \setminus \{u_j, v_j\}}} \prod_{\substack{v \sim u_j \\ v \neq v_j}} W(x_{u_j}, x_v) \prod_{v \in V(F) \setminus \{u_j, v_j\}} dx_v$$

and

$$g_j(x_{v_j}) = \int_{[0,1]^{V(F) \setminus \{u_j, v_j\}}} \prod_{\substack{u \neq u_j \\ u \sim v_j}} W(x_u, x_{v_j}) \prod_{v \in V(F) \setminus \{u_j, v_j\}} dx_v.$$

Since these are averages of products of functions which take values between 0 and 1, we have $0 \leq f_j, g_j \leq 1$ for each j . So, by Lemma 3, each term of (1) is at most $\|W - U\|_{\square}$. This finishes the proof. \blacksquare

3 Proof of the weak regularity lemma

By Lemma 4 (the counting lemma), Theorem 2 is implied by the following reformulation, which we will prove in this final section.

Theorem 5 (weak regularity lemma III). *For any $k \geq 4$ and any graphon W , there is a stepfunction S with k steps such that*

$$\|W - S\|_{\square} < \frac{2}{\sqrt{\log k}}.$$

3.1 Reduction to polynomial-size description

The bound given in Theorem 5 implies that we can obtain error δ with $k = \exp(2/\delta^2)$ steps. This is exponentially large, but luckily there is a description of the stepfunction which has polynomial size in terms of the error.

Theorem 6 (weak regularity lemma IV). *For any $m \geq 1$ and any graphon W , there are m pairs of subsets $S_i, T_i \subseteq [0, 1]$ and m real numbers $a_i \in [0, 1]$ such that*

$$\left\| W - \sum_{i=1}^m a_i \mathbf{1}_{S_i \times T_i} \right\|_{\square} < \frac{1}{\sqrt{m}}.$$

By symmetrizing the sum by averaging with $\sum_{i=1}^m a_i \mathbf{1}_{T_i \times S_i}$ (which does not change the cut norm), we get a stepfunction with at most 2^{2m} pieces. We can thus obtain Theorem 5 from Theorem 6 when $k = 2^{2m}$. Since we need m to be an integer, for k not a power of 4, we can take $m = \lfloor \log_4(k) \rfloor$ in general, and obtain the bound

$$\frac{1}{\sqrt{m}} = \frac{1}{\sqrt{\lfloor \log_4(k) \rfloor}} \leq \frac{2}{\sqrt{\log(k)}},$$

the last bound holding whenever $k \geq 4$. So it just remains to prove Theorem 6.

3.2 Finishing the proof

We can prove Theorem 6 by successively removing the best approximation $a \mathbf{1}_{S \times T}$ from W . The following lemma tells us how good these approximations can be, in terms of the L^2 norm.

Lemma 7 (L^2 increments). *For every symmetric function $W : [0, 1]^2 \rightarrow \mathbb{R}$, there are two sets $S, T \subseteq [0, 1]$ and a real number $a \in [0, \|W\|_{\infty}]$ such that*

$$\|W - a \mathbf{1}_{S \times T}\|_2^2 \leq \|W\|_2^2 - \|W\|_{\square}^2.$$

Proof. Let $S, T \subseteq [0, 1]$ such that

$$\|W\|_{\square} = \left| \int_{S \times T} W(x, y) dx dy \right|,$$

as guaranteed by Lemma 3. Let $a = \frac{1}{\lambda(S \times T)} \int_{S \times T} W(x, y) dx dy$, where λ is the Lebesgue measure. Then

$$\begin{aligned} \|W - a\mathbf{1}_{S \times T}\|_2^2 &= \|W\|_2^2 - \int_{S \times T} aW(x, y) dx dy + a^2\lambda(S \times T) \\ &= \|W\|_2^2 - \frac{1}{\lambda(S \times T)} \left(\int_{S \times T} W(x, y) dx dy \right)^2 \\ &\leq \|W\|_2^2 - \|W\|_{\square}^2, \end{aligned}$$

since $\lambda(S \times T) \leq 1$, and the integral in the second line is $\pm \|W\|_{\square}$. ■

With Lemma 7 in hand, we can finish the proof of the weak regularity lemma.

Proof of Theorem 6. Apply Lemma 7 repeatedly to get pairs of sets S_i, T_i , and real numbers a_i , such that

$$\begin{aligned} \|W - a_1\mathbf{1}_{S_1 \times T_1}\|_2^2 &\leq \|W\|_2^2 - \|W\|_{\square}^2, \\ \|W - a_1\mathbf{1}_{S_1 \times T_1} - a_2\mathbf{1}_{S_2 \times T_2}\|_2^2 &\leq \|W - a_1\mathbf{1}_{S_1 \times T_1}\|_2^2 - \|W - a_1\mathbf{1}_{S_1 \times T_1}\|_{\square}^2 \\ &\leq \|W\|_2^2 - \|W\|_{\square}^2 - \|W - a_1\mathbf{1}_{S_1 \times T_1}\|_{\square}^2 \\ &\vdots \\ \left\| W - \sum_{i=1}^k a_i \mathbf{1}_{S_i \times T_i} \right\|_2^2 &\leq \|W\|_2^2 - \sum_{j=0}^{k-1} \left\| W - \sum_{i=1}^j a_i \mathbf{1}_{S_i \times T_i} \right\|_{\square}^2. \end{aligned}$$

Now the left-hand side is ≥ 0 , and $\|W\|_2^2 \leq 1$, so there is at least one $j \in \{0, \dots, k-1\}$ for which

$$\left\| W - \sum_{i=1}^j a_i \mathbf{1}_{S_i \times T_i} \right\|_{\square}^2 \leq \frac{1}{k}.$$

This finishes the proof, since we can just set $a_{j+1}, \dots, a_k = 0$. ■

References

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