

Problem sheet 4

Problem 4.1. Show that the function $h: \mathcal{G} \rightarrow [0, 1]$ defined as $h(\mathcal{G}, o) = |\mathcal{G}|^{-1} \mathbb{1}_{\{|\mathcal{G}| < \infty\}}$ is continuous in the weak local topology.

Proof. We have to show that the inverse image $h^{-1}(\mathcal{B})$ of each open subset $\mathcal{B} \subset [0, 1]$ is open in \mathcal{G} . Since the topology of open sets in $[0, 1]$ is generated by open intervals, it suffices to consider $\mathcal{B} = (a, b) \subset [0, 1]$. Let

$$n(a) = \max\{n \in \mathbb{N}: \frac{1}{n} > a\} \quad \text{and} \quad n(b) = \min\{n \in \mathbb{N}: \frac{1}{n} < b\}.$$

By definition, each finite rooted graph (\mathcal{G}, o) with $|\mathcal{V}| \in [n(b), n(a)]$ yields $h(\mathcal{G}, o) \in (a, b)$. Next, write \mathcal{G}_n for the subset of all rooted graph on exactly n vertices. Then, \mathcal{G}_n is open as

$$\mathcal{G}_n = \bigcup_{\mathcal{G} \in \mathcal{G}_n} \{\mathcal{G}\},$$

which is a finite union of open sets (recall that singletons of finite graphs are open in the local topology). Therefore,

$$h^{-1}(a, b) = \bigcup_{n=n(b)}^{n(a)} \mathcal{G}_n$$

is a union of open sets and thus open itself. This finishes the proof. \square

Problem 4.2. Consider the Stochastic block model on the vertices $\mathcal{V}_n = \mathcal{V}_n^{(1)} \cup \mathcal{V}_n^{(2)}$, where $\mathcal{V}_n = [n]$ and $\mathcal{V}_n^{(1)}$ contains of all even labels and $\mathcal{V}_n^{(2)}$ of all uneven labels. We connect each vertex of the same type independently with probability a/n and each pair of different types with probability b/n . Identify the Poisson BGW tree \mathcal{T} that describes the weak local limit and identify, for which a, b the tree has positive probability of survival.

Proof. First of all, recall that the stochastic block model is an IRG with type space $\{0, 1\}$ with limiting type distribution $\mu(1) = \mu(2) = 1/2$ (in the current situation). Moreover, the connection kernel is given as $\kappa_n(x_i, x_j) = \kappa(x_i, x_j) = a + (b - a)\mathbb{1}_{\{a \neq b\}}$. Hence, the limiting BGW tree \mathcal{T} has Poisson offspring with mean

$$\lambda_{i,j} = \kappa(i, j)\mu(j) = \frac{1}{2} \left(a\mathbb{1}_{\{i=j\}} + b\mathbb{1}_{\{i \neq j\}} \right)$$

by Theorem 3.1. This yields the expectation matrix

$$M = \frac{1}{2} \begin{pmatrix} a & b \\ b & a \end{pmatrix}.$$

In order to determine the supercritical regime, we have to identify the largest eigenvalue ρ . To this end, we calculate

$$0 = \det(M - \lambda \mathbb{1}) = \left(\frac{a}{2} - \lambda\right)^2 - \frac{b^2}{4} = \lambda^2 - a\lambda + \frac{a^2 - b^2}{4}.$$

This equation has the solutions

$$\lambda_{1,2} = \frac{a}{2} \pm \sqrt{\frac{a^2}{4} - \frac{a^2 - b^2}{4}} = \frac{a \pm b}{2}.$$

Thus, $\rho = (a + b)/2$ and the BGW tree is supercritical if $\rho > 1 \Leftrightarrow a + b > 2$. \square

Problem 4.3 (Theorem 4.6). Prove Theorem 4.6 of the lecture notes: Let (\mathcal{G}_n) be the inhomogeneous random graph associated with the kernels (κ_n) that converge graphically to κ . Then the empirical degree-degree distribution converges to the probability measure ν_2 given by

$$\nu_2(k, l) = \frac{1}{c} \int \frac{\lambda(x)^{k-1}}{(k-1)!} e^{-\lambda(x)} \left(\int \frac{\lambda(y)^{l-1}}{(l-1)!} e^{-\lambda(y)} \kappa(x, y) \mu(dy) \right) \mu(dx),$$

where $c = \int \int \kappa(x, y) \mu(dy) \mu(dx)$.

Proof. We apply Proposition 4.4 and note that uniform integrability in the assumptions was only used to ensure sparsity. Hence, it is not required here by Lemma 1.5 and we have

$$\nu_2(k, l) = \frac{k}{\mathbb{E}\mathcal{D}(o)} \mathbb{P}(\mathcal{D}(o) = k, \mathcal{D}(V) = l).$$

As $c = \mathbb{E}(\mathcal{D}(o))$, it hence remains to calculate $k\mathbb{P}(\mathcal{D}(o) = k, \mathcal{D}(V) = l)$ for the limiting BGW tree. Recall that the number of offsprings of a type x vertex in the limiting BGW tree is Poisson distributed with parameter $\lambda(x) = \int \kappa(x, y) \mu(dy)$. Thus, denoting (o, x) for the root of type x ,

$$\begin{aligned} \mathbb{P}(\mathcal{D}(o) = k, \mathcal{D}(V) = l) &= \int_{\mathcal{S}} \mathbb{P}(\mathcal{D}(o, x) = k, \mathcal{D}(V) = l) \mu(dx) \\ &= \int_{\mathcal{S}} \mathbb{P}(\mathcal{D}(o, x) = k) \mathbb{P}(\mathcal{D}(V) = l \mid \mathcal{D}(o, x) = k) \mu(dx) \\ &= \int_{\mathcal{S}} \frac{\lambda(x)^k}{k!} e^{-\lambda(x)} \mathbb{P}(\mathcal{D}(V) = l \mid \mathcal{D}(o, x) = k) \mu(dx) \end{aligned}$$

To calculate the remaining probability, recall that, given $\mathcal{D}(o, x) = k$, the k children of the root are independently distributed according to the normalised intensity measure $\kappa(x, y) \mu(dy) / \lambda(x)$, which follows from the Poisson point process structure of the neighbourhood. As V is chosen uniformly at random among these k vertices, it follows the same distribution. Hence,

$$\mathbb{P}(\mathcal{D}(V) = l \mid \mathcal{D}(o, x) = k) = \int_{\mathcal{S}} \mathbb{P}(\mathcal{D}(v, y) = l) \frac{\kappa(x, y) \mu(dy)}{\lambda(x)},$$

where (v, y) is a root's child of type y . In order to have degree l , the vertex v that is already connected to the root must have $l - 1$ children itself. As each vertex produces its offspring independently, we have

$$\mathbb{P}(\mathcal{D}(v, y) = l) = \frac{\lambda(y)^{l-1}}{(l-1)!} e^{-\lambda(y)}.$$

Putting everything together, we finally infer,

$$\begin{aligned} k\mathbb{P}(\mathcal{D}(o) = k, \mathcal{D}(V) = l) &= \int_{\mathcal{S}} \frac{\lambda(x)^k}{(k-1)!} e^{-\lambda(x)} \mathbb{P}(\mathcal{D}(V) = l \mid \mathcal{D}(o, x) = k) \mu(dx) \\ &= \int \frac{\lambda(x)^k}{(k-1)!} e^{-\lambda(x)} \left(\int \frac{\lambda(y)^{l-1}}{(l-1)!} e^{-\lambda(y)} \frac{\kappa(x, y) \mu(dy)}{\lambda(x)} \right) \mu(dx), \end{aligned}$$

as claimed. \square

Problem 4.4 (Global clustering). In the lecture, we studied the *local clustering coefficient*, describing the probability that two neighbours of a uniformly chosen vertex are neighbours themselves and thus form a triangle. In contrast the *global clustering coefficient* is defined as

$$c_n^{\text{glob}} = \frac{\sum_{v \in \mathcal{G}_n} \Delta(\mathcal{G}_n, v)}{\frac{1}{2} \sum_{v \in \mathcal{G}_n} \mathcal{D}(v)(\mathcal{D}(v) - 1)},$$

i.e. the overall proportion of wedges that complete a triangle. Assume $(\mathcal{G}_n)_n$ converges weakly locally to $(\mathcal{G}, o) \in \mathcal{G}$ and that the sequence $(\mathcal{D}(O_n)^2)_n$ is uniformly integrable. Show that

$$c_n^{\text{glob}} \longrightarrow \frac{\mathbb{E}[\Delta(\mathcal{G}, o)]}{\frac{1}{2} \mathbb{E}[\mathcal{D}(o)(\mathcal{D}(o) - 1)]},$$

in probability, as $n \rightarrow \infty$.

Proof. (a): Consider the function $h(G, o) = \Delta(G, o)$, which is continuous but unbounded. Note that

$$\frac{1}{n} \sum_v h(\mathcal{G}_n, v) = h(\mathcal{G}_n, O_n) \leq \frac{1}{2} \mathcal{D}(O_n)(\mathcal{D}(O_n) - 1) \leq \mathcal{D}(O_n)^2.$$

Therefore, we can choose $\varepsilon, \delta > 0$ and $K > 1$, such that for all $k > K$

$$\begin{aligned} \mathbb{P}\left(\frac{1}{n} \sum_v \Delta(\mathcal{G}_n, v) \mathbb{1}_{\{\mathcal{D}(v) > k\}} > \delta\right) &\leq \frac{1}{\delta} \mathbb{E}[\Delta(\mathcal{G}_n, O_n) \mathbb{1}_{\{\mathcal{D}(O_n) > k\}}] \\ &\leq \frac{1}{\delta} \mathbb{E}[\mathcal{D}(O_n)^2 \mathbb{1}_{\{\mathcal{D}(O_n)^2 > k\}}] \\ &< \frac{\varepsilon}{2}, \end{aligned}$$

by the uniform integrability of $\mathcal{D}(O_n)^2$. Next, for $k > K$, we have by local convergence

$$\mathbb{P}\left(\left|\Delta(\mathcal{G}_n, O_n) \mathbb{1}_{\{\mathcal{D}(O_n) \leq k\}} - \Delta(\mathcal{G}, o) \mathbb{1}_{\{\mathcal{D}(o) \leq k\}}\right| > \delta\right) \leq \frac{\varepsilon}{2}.$$

Finally, we have $\mathbb{E}[\mathcal{D}(o)^2] < \infty$ by uniform integrability. Therefore,

$$\mathbb{E}[\Delta(\mathcal{G}, o)] - \mathbb{E}[\Delta(\mathcal{G}, o) \mathbb{1}_{\{\mathcal{D}(o) \leq k\}}] < \delta,$$

for sufficiently large $k > K$. Combined, this yields

$$\begin{aligned} & \mathbb{P}\left(\left|\frac{1}{n} \sum_v \Delta(\mathcal{G}_n, v) - \mathbb{E}[\Delta(\mathcal{G}, o)]\right| > 3\delta\right) \\ & \leq \mathbb{P}\left(\left|\frac{1}{n} \sum_v \Delta(\mathcal{G}_n, v) - \mathbb{E}[\Delta(\mathcal{G}, o)] \mathbb{1}_{\{\mathcal{D}(o) \leq k\}}\right| > 2\delta\right) \\ & \leq \mathbb{P}\left(\left|\frac{1}{n} \sum_v \Delta(\mathcal{G}_n, v) \mathbb{1}_{\{\mathcal{D}(v) \leq k\}} - \mathbb{E}[\Delta(\mathcal{G}, o)] \mathbb{1}_{\{\mathcal{D}(o) \leq k\}}\right| > \delta\right) \\ & \quad + \mathbb{P}\left(\left|\frac{1}{n} \sum_v \Delta(\mathcal{G}_n, v) \mathbb{1}_{\{\mathcal{D}(v) > k\}} > \delta\right)\right) \\ & \leq \varepsilon. \end{aligned}$$

Put differently,

$$\frac{1}{n} \sum_{v \in \mathcal{G}_n} \Delta(\mathcal{G}_n, v) \longrightarrow \mathbb{E}[\Delta(\mathcal{G}, o)], \quad \text{in probability.}$$

With exactly the same arguments, we also have

$$\frac{1}{n} \sum_{v \in \mathcal{G}_n} \mathcal{D}(v)(\mathcal{D}(v) - 1) \longrightarrow \mathbb{E}[\mathcal{D}(o)(\mathcal{D}(o) - 1)], \quad \text{in probability.}$$

Combined, we infer

$$2c^{\text{glob}} = \frac{\sum_v \Delta(\mathcal{G}_n, v)}{n} \cdot \frac{n}{\sum_v \mathcal{D}(v)(\mathcal{D}(v) - 1)} \longrightarrow \frac{\mathbb{E}[\Delta(\mathcal{G}, o)]}{\mathbb{E}[\mathcal{D}(o)(\mathcal{D}(o) - 1)]},$$

in probability. \square

Problem 4.5 (Proposition 4.11). Finish the proof of Proposition 4.11: Let (\mathcal{G}_n) be a sequence of random graphs converging weakly locally to a random rooted graph (\mathcal{G}, o) . Then, in probability,

$$\frac{\mathcal{K}_n}{n} \longrightarrow \mathbb{E}\left[\frac{1}{|\mathcal{G}|}\right] \geq 0.$$

Proof. Define the function $h: \mathcal{G} \rightarrow [0, 1]$ via $h(G, o) = |\mathcal{C}(o)|^{-1} \mathbb{1}_{\{|\mathcal{C}(o)| < \infty\}}$. Recall that, in the space \mathcal{G} of rooted connected locally finite graphs, we identify the graph G with the connected component $\mathcal{C}(o)$ of its root. Hence, $h(G, o) = |\mathcal{G}|^{-1} \mathbb{1}_{\{|\mathcal{G}| < \infty\}}$ and it is continuous and bounded by Problem 4.1. Moreover, in a finite graph all components are finite. Note further

$$\mathcal{K}_n = \sum_{v \in \mathcal{G}_n} |\mathcal{C}(v)|^{-1},$$

as each component appears in the sum precisely as often as its size. Therefore, local convergence yields

$$\frac{\mathcal{K}_n}{n} = \frac{1}{n} \sum_{v \in \mathcal{G}_n} h(\mathcal{G}_n, v) \longrightarrow \mathbb{E}[h(\mathcal{G}, o)] = \mathbb{E}[|\mathcal{G}|^{-1} \mathbb{1}_{\{|\mathcal{G}| < \infty\}}]$$

However, as $1/\infty = 0$, we also have

$$\mathbb{E}[|\mathcal{G}|^{-1}] = \mathbb{E}[|\mathcal{G}|^{-1} \mathbb{1}_{\{|\mathcal{G}| < \infty\}}] + \mathbb{E}[0 \cdot \mathbb{1}_{\{|\mathcal{G}| = \infty\}}] = \mathbb{E}[|\mathcal{G}|^{-1} \mathbb{1}_{\{|\mathcal{G}| < \infty\}}],$$

in probability, which concludes the proof. \square

Problem 4.6. Let $(\mathcal{G}_n)_n$ be a sequence of random graphs converging weakly locally to a rooted graph (\mathcal{G}, o) , and $\mathcal{V}_n^{(2)}$ be the set of all vertices of degree at least two. Show that

$$\frac{n}{|\mathcal{V}_n^{(2)}|} \longrightarrow \frac{1}{\mathbb{P}(\mathcal{D}(o) \geq 2)} \quad (1)$$

in probability. Using (1), show that $\frac{1}{n} \sum_{v \in \mathcal{V}_n^{(2)}} \mathbf{c}_n^{\text{loc}}(v) \longrightarrow \mathbb{E} \left[\frac{\Delta(\mathcal{G}, o)}{\frac{1}{2} \mathcal{D}(o)(\mathcal{D}(o) - 1)} \mathbb{1}_{\{\mathcal{D}(o) \geq 2\}} \right]$ in probability implies

$$\mathbf{c}_n^{\text{av}} \longrightarrow \mathbb{E} \left[\frac{\Delta(\mathcal{G}, o)}{\frac{1}{2} \mathcal{D}(o)(\mathcal{D}(o) - 1)} \mid \mathcal{D}(o) \geq 2 \right],$$

in probability.

Proof. Clearly,

$$\frac{|\mathcal{V}_n^{(2)}|}{n} = \frac{1}{n} \sum_{v \in \mathcal{G}_n} \mathbb{1}_{\{\mathcal{D}(v) \geq 2\}} \longrightarrow \mathbb{P}(\mathcal{D}(o) \geq 2),$$

in probability by local convergence. This proves (1). Hence, in probability,

$$\begin{aligned} \mathbf{c}_n^{\text{av}} &= \frac{n}{|\mathcal{V}_n^{(2)}|} \cdot \frac{\sum_{v \in \mathcal{V}_n^{(2)}} \mathbf{c}_n^{\text{loc}}(v)}{n} \longrightarrow \frac{1}{\mathbb{P}(\mathcal{D}(o) \geq 2)} \mathbb{E} \left[\frac{\Delta(\mathcal{G}, o)}{\frac{1}{2} \mathcal{D}(o)(\mathcal{D}(o) - 1)} \mathbb{1}_{\{\mathcal{D}(o) \geq 2\}} \right] \\ &= \mathbb{E} \left[\frac{\Delta(\mathcal{G}, o)}{\frac{1}{2} \mathcal{D}(o)(\mathcal{D}(o) - 1)} \mid \mathcal{D}(o) \geq 2 \right]. \end{aligned}$$

□