Weakly reinforced Pólya urns on countable networks

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Abstract

We study the long-time asymptotics of a network of weakly reinforced Pólya urns. In this system, which extends the WARM introduced by R. van der Hofstad et. al. (2016) to countable networks, the nodes fire at times given by a Poisson point process. When a node fires, one of the incident edges is selected with a probability proportional to its weight raised to a power \( \alpha < 1 \), and then this weight is increased by 1.

We show that for \( \alpha < 1/2 \) on a network of bounded degrees, every edge is reinforced a positive proportion of time, and that the limiting proportion can be interpreted as an equilibrium in a countable network. Moreover, in the special case of regular graphs, this homogenization remains valid beyond the threshold \( \alpha = 1/2 \).

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1 Introduction

Pólya’s urn process is the paradigm model for a random process incorporating reinforcement effects. However, when thinking in the direction of applications, a single urn does not represent complex interactions accurately. For instance, in the field of social sciences, the formation of friendship networks could be related to reinforcement effects in social interactions [7]. In economy, in a network of companies competing on a variety of products, the global reputation could influence the market shares of the products differently [1, 2]. Finally, in the domain of neuroscience, it is plausible that synapses that were successful in the past should be selected again in the future and reinforced with a higher probability [5].

In the present paper, we focus on the network formation model proposed in [5] and study its long-time behavior in the regime of weak reinforcement. In that model, starting from a base network, nodes are picked sequentially at random. Once a node is selected, we choose one of the incident edges with a probability proportional to its weight to some power \( \alpha > 0 \) and increase that weight by 1. The analysis of [5] concerns the asymptotic proportion of times that edges are reinforced in the regime of strong reinforcement, where \( \alpha > 1 \). In this regime, the limiting proportion is random and coincides with some

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stable equilibrium in an associated dynamical system, which is typically concentrated on a small subset of the edges if \( \alpha \) is large.

In contrast, we consider the regime of weak reinforcement, where \( \alpha < 1 \) and find that in many settings the reinforcement proportions converge to a uniquely determined limit equilibrium, which is supported on all edges of the base network. Figure 1 illustrates the network at an early and at a late time point.

![Figure 1: Weakly reinforced Pólya model after time 2 (left) and time 100 (right). Dashed edges have not been reinforced until the considered time. The widths of solid edges are proportional to time-normalized weights.](image)

Hence, together with the analysis in [5, 6], our main result is a first step towards a network-based analog of Rubin’s dichotomy for classical Pólya urns: while for strong reinforcement some edges are only reinforced finitely often, in the weakly reinforced regime all edges are reinforced a positive proportion of time. Although this description outlines the broader picture, more research is needed to carve out the precise conditions for this dichotomy. Indeed, [3] describes an example of a meticulously designed network together with firing rates where even for \( \alpha > 1 \) there is percolation of edges that are reinforced a positive proportion of time.

Similarly to the setting of [4], one major difficulty in the analysis stems from the choice of the base graph. In contrast to [1, 5], we do not restrict our attention to finite base graphs, but work on a countable network with uniformly bounded degrees. In particular, the highly developed machinery of stochastic approximation algorithms invoked in [1, 5] is not available as the state space of the associated dynamical system would be infinite-dimensional. In [4], the very strong reinforcement leads to an effective decomposition of the countable network into finite islands separated by edges that are never reinforced. This trick recovers the finite-dimensional setting. However, the strategy cannot work in the regime of weak reinforcement, where we expect all edges to be reinforced a positive proportion of time.

Hence, in the present paper, we follow an entirely different plan to prove our main result. Namely, convergence of the normalized edge weights to an equilibrium for (1) all graphs of uniformly bounded degree if \( \alpha < 1/2 \), (2) regular graphs if \( \alpha < 1/2 + \varepsilon \) for some \( \varepsilon > 0 \), and (3) \( \mathbb{Z} \) if \( \alpha < 1 \). This plan consists of three critical steps. First, we invoke a compactness argument to obtain an equilibrium on the entire countable network. Second, we establish that all edges are reinforced a positive proportion of time. This step rests on a percolation argument, where we decompose the network again into...
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finite islands separated by edges that are now reinforced a positive proportion of time. Finally, to obtain convergence to the equilibrium, we rely on a homogenizing bootstrap argument. It formalizes the intuition that for weak reinforcement, deviations from the equilibrium decrease over time.

The rest of the manuscript is organized as follows. Section 2 introduces the model and states the main contribution of the paper, a homogenization result in the sub-linear regime for three different combinations of graphs and exponents. In Section 3, we consider graphs of bounded degree and \( \alpha < 1/2 \). Finally, Section 4 deals with regular graphs of degree \( \Delta \) and covers \( \alpha < 1/2 + \epsilon \) for \( \Delta > 2 \) and \( \alpha < 1 \) for the graph \( \mathbb{Z} \).

## 2 Model definition and main result

Let \( G = (V, E) \) be a countable graph with uniformly bounded degrees. That is, the vertex set \( V \) is countable and

\[
\Delta := \sup_{v \in V} \deg(v) < \infty.
\]

If \( \Delta = \deg(v) \) for every \( v \in V \) then the graph is regular with degree \( \Delta \), which we write as \( \Delta \)-regular.

We investigate a system of random variables

\[
N_t := \{N_t(e)\}_{e \in E}
\]

of interacting Pólya-type urns on the edge set \( E \) at continuous time \( t \geq 0 \) on a probability space \((\Omega, \mathcal{F})\). The dynamics of \( N_t \) are a continuous-time analog of the process considered in [5]. Loosely speaking, every vertex has a Poisson clock and whenever that clock rings at a node \( v \in V \), the dynamics choose and increment the weights on one of the incident edges \( E_v = \{e \in E : v \in e\} \) by 1. The choice of the edge happens using the power-weighted Pólya-scheme

\[
\text{pol}_{v,e}(N_t) := \frac{N_t(e)^\alpha}{\sum_{e' \in E_v} N_t(e')^\alpha}, \quad \alpha \geq 0.
\]  

Henceforth, we tacitly assume that \( \alpha < 1 \). More precisely, the weights \( N_t(e) \) are initialized at 1 and the dynamics of \( N_t \) are governed by an iid family of Poisson processes \( \{P_v\}_{v \in V} \) on \([0, \infty) \times [0, 1]\) with intensity 1 counting clock-ring events at vertex \( v \in V \) in a time window \([0, \infty)\) with marks in the range \([0, 1]\). If the process \( P_t \) contains an atom of the form \((t, u) \in [0, \infty) \times [0, 1]\), then increment the mass of an edge \( e_i \in E_v = \{e_1, \ldots, e_{\deg(v)}\} \) by 1 if \( u \in U_{v,e_i} \), where \( \{U_{v,e_i}\}_{i \leq \deg(v)} \) is a partition of \([0, 1]\) given by

\[
U_{v,e_i} = \left( \sum_{j \leq i-1} \text{pol}_{v,e_j}(N_t), \sum_{j \leq i} \text{pol}_{v,e_j}(N_t) \right).
\]

For the existence of the process \( \{N_t\}_{t \geq 0} \) on bounded-degree graphs, see [4].

Finally, a non-negative vector \( \mu \in \mathbb{R}^E_+ \) defines an equilibrium on \( G \) if

\[
\mu(e) := \frac{\sum_{v \in V} \text{pol}_{v,e}(N_t) \mu(e)}{\sum_{e' \in E_v} \mu(e')^\alpha}, \quad \mu(e) > 0
\]  

holds for all \( e \in E \) with \( \mu(e) > 0 \). This is a straightforward extension of the notion of finite equilibria from [5] to countable networks. Now, let

\[
X_t := \frac{N_t}{t}
\]

denote the time-normalized system of weights at time \( t > 0 \). We say that \( X_t \) exhibits homogenization if the equilibrium measure \( \mu \) exists, is unique and \( X_t \rightarrow \mu \) almost surely.
Conjecture 2.1 (Homogenization). Let $G$ be a countable graph with uniformly bounded degrees. Then, $X_t$ exhibits homogenization.

In the present work, we verify this conjecture for three combinations of $G$ and $\alpha$.

Theorem 2.2 (Homogenization). $X_t$ exhibits homogenization in the following cases.

1. The degree of $G$ is uniformly bounded and $\alpha < 1/2$.
2. $G$ is $\Delta$-regular and $\alpha < 1/2 + \varepsilon$ for some $\varepsilon > 0$.
3. $G = \mathbb{Z}$ and $\alpha < 1$.

3 Proof of Theorem 2.2 (1)

In this section, we establish part (1) of Theorem 2.2. That is, we show homogenization for $\alpha < 1/2$ and graphs of bounded degree. First, in Section 3.1, we prove existence of a non-vanishing equilibrium. Then, in Section 3.2, we show that $X_t$ converges to such an equilibrium, which is in fact uniquely determined.

3.1 Existence of equilibrium

Henceforth, we call an equilibrium $\mu \in \mathbb{R}_E^+$ non-vanishing if $\mu(e) > 0$ for every $e \in E$. Before discussing existence of non-vanishing equilibria on general graphs, we present the $\Delta$-regular case as a particularly easy example.

Example 3.1 ($\Delta$-regular graphs). Let $G$ be a $\Delta$-regular graph. Then, $\mu \equiv 2/\Delta$ defines a non-vanishing equilibrium. Indeed, checking Equation (2.2) leads to

$$\sum_{v \in E} \frac{\mu^\alpha}{\sum_{e' \in E_v} \mu^\alpha} = \sum_{e \in E} \frac{1}{\sum_{e' \in E_v} 1} = \frac{2}{\Delta} = \mu.$$  

The general case of bounded-degree graphs does not admit such an easy analysis as the equilibrium could in principle assume an infinite number of different values. Nevertheless, we deduce existence from a compactness argument.

Proposition 3.2 (Existence of non-vanishing equilibria). Every graph with degrees uniformly bounded by $\Delta$ exhibits at least one non-vanishing equilibrium $\mu$ with $\mu(e) \geq 2/(\Delta^{1/(1-\alpha)})$ for all $e \in E$.

Proof. By Equation (2.2), the equilibria correspond to fixed points of the continuous operator

$$T : \mathbb{R}_E^+ \to \mathbb{R}_E^+ \quad \mu(\cdot) \mapsto \sum_{e \in E_v} \frac{\mu^\alpha}{\sum_{e' \in E_v} \mu^\alpha}.$$  

As a product of locally convex and Hausdorff spaces, $\mathbb{R}_E^+$ is again locally convex and Hausdorff. Now, define the closed set $C = [2\Delta^{-1/(1-\alpha)}, 2]^E$, which as a product of convex and compact sets, is again convex and compact. We claim that $T(C) \subseteq C$. Once this claim is established, Schauder’s fixed point theorem yields a non-vanishing equilibrium in $C$.

To prove $T(C) \subseteq C$, note that for every $\mu \in C$ and $e = \{v_1, v_2\} \in E$ we have

$$T(\mu)(e) = \frac{\mu(e)^\alpha}{\sum_{e' \in E_{v_1}} \mu(e')^\alpha} + \frac{\mu(e)^\alpha}{\sum_{e' \in E_{v_2}} \mu(e')^\alpha} \leq 2,$$  

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and

\[ T(\mu)(e) = \sum_{i \leq 2} \frac{\mu(e)^{\alpha}}{\sum_{e' \in E_i} \mu(e')^{\alpha}} \geq \sum_{i \leq 2} \frac{\mu(e)^{\alpha}}{2^i \deg(v_i)} \geq \frac{2\mu(e)^{\alpha}}{2^i \Delta} \geq \frac{2}{\Delta^{1/(1-\alpha)}}, \]

as claimed. \qed

3.2 Proof of Theorem 2.2 (1)

The proof of part (1) of Theorem 2.2 is based on three pivotal ingredients.

First, as each edge is incident to two vertices firing at rate 1, its weight grows at rate at most 2. In other words, for every \( e \in E \), almost surely,

\[ X_+(e) := \lim sup_{t \to \infty} X_t(e) \leq 2. \]

Lemma 3.3 (Rate upper bound). Let \( e \in E \) be arbitrary. Then, \( P(X_+(e) \leq 2) = 1. \)

Second, we derive a positive lower bound for the growth rate

\[ X_-(e) := \lim inf_{t \to \infty} X_t(e). \]

This task is slightly more subtle than the upper bound, since arbitrarily small rates can occur with positive probability. To approach this challenge, we fix throughout this section a non-vanishing equilibrium \( \mu \) as in Theorem 3.2 and write

\[ E_e := \{ e' \in E : e' \cap e \neq \emptyset \} \]

for the family of edges adjacent to a given edge \( e \in E \). Then, an edge \( e \in E \) is \( \delta \)-stable if

\[ \min_{e' \in E_e} \frac{X_-(e')}{\mu(e')} \geq \delta. \]

Lemma 3.4 (Rate lower bound). There exists \( \delta_0 > 0 \) such that with probability 1, all connected components of \( \delta_0 \)-unstable edges are finite.

Finally, we rely on a bootstrapping procedure to push the weights iteratively closer to 1. To that end, define

\[ C(e) := [X_-(e), X_+(e)] \]

as the smallest interval containing the accumulation points of \( X_t(e) \).

Lemma 3.5 (Bootstrap on bounded degree graphs). Let \( \alpha < 1/2 \) and \( \gamma > 1 \). If \( e \in E \) is such that \( C(e') \subset [\gamma^{-1} \mu(e'), \gamma \mu(e')] \) holds for all \( e' \in E_e \), then \( C(e) \subset [\gamma^{-2\alpha} \mu(e), \gamma^{2\alpha} \mu(e)] \).

Before establishing Lemmas 3.3–3.5, we elucidate how they can be combined to yield the proof of part (1) of Theorem 2.2.

Proof of Theorem 2.2(1). Let \( \delta_0 > 0 \) be as in Theorem 3.4. Assume without loss of generality that \( \delta_0 \leq \Delta^{-1/(1-\alpha)} \) so that \( 2 \leq \mu(e) \Delta^{1/(1-\alpha)} \leq \mu(e)/\delta_0 \) for any \( e \in E \) where the first inequality follows from \( \mu(e) \geq 2/\Delta^{1/(1-\alpha)} \) given by Theorem 3.2. We first claim that \( X_-(e) \geq \delta_0 \mu(e) \) almost surely for every \( e \in E \). Indeed, assume the contrary and let \( S \) be a connected component of \( \delta_0 \)-unstable edges. By Theorem 3.4, \( S \) is finite and we choose \( e^* \in S \) (randomly) such that \( \gamma := \frac{\mu(e^*)}{X_+(e^*)} \) is maximal. Then, \( \gamma^{-1} \geq \delta_0 \) almost surely since otherwise Theorem 3.3 and the bootstrap property of Theorem 3.5, combined with \( 2 \leq \mu(e)/\delta_0 \leq \gamma \mu(e) \), would give that \( C(e^*) \subset [\gamma^{-2\alpha} \mu(e^*), 2] \), thereby contradicting that

\[ \frac{X_-(e^*)}{\mu(e^*)} = \gamma^{-1} < \gamma^{-2\alpha}. \]
Hence, the infimum
\[ g_* := \inf\{ \rho > 1 : C(e) \subseteq [\rho^{-1} \mu(e), \rho \mu(e)] \text{ for every } e \in E \} \]
is at most \( \delta_0^{-1} \). Now, \( g_* = 1 \) almost surely since otherwise the bootstrap property yields
\[ C(e) \subseteq [g_*^{-2\alpha} \mu(e), g_*^{2\alpha} \mu(e)] \text{ for every } e \in E, \]
thereby contradicting the choice of \( g_* \). \( \square \)

It remains to establish the auxiliary results. We start by proving Lemma 3.3.

**Proof of Lemma 3.3.** One extremal case is that each clock-ring event for \( P_e \) increments the value for an edge \( e = \{v, w\} \in E \). The process
\[ Y_t := P_e([0, t] \times [0, 1]) + P_w([0, t] \times [0, 1]) \]
counts the normalized occurrences for this upper bound and has an expected value of
\[ \mathbb{E}[Y_t] = \frac{\mathbb{E}[P_e([0, t] \times [0, 1])] + \mathbb{E}[P_w([0, t] \times [0, 1])]}{t} = \frac{2t}{t} = 2. \]
Then, the strong law of large numbers for homogeneous Poisson point processes, gives that almost surely \( \lim_{t \to \infty} Y_t = 2 \). Now, \( X_t(e) \leq Y_t + \frac{1}{t} \) for all \( t \geq 0 \), implies that almost surely, \( \limsup_{t \to \infty} X_t(e) \leq \lim_{t \to \infty} Y_t = 2 \). \( \square \)

Next, we verify the bootstrap property.

**Proof of Lemma 3.5.** Let \( \rho > 1 \) and \( e = \{v_1, v_2\} \in E \) be such that \( C(e') \subseteq [\rho^{-1} \mu(e'), \rho \mu(e')] \) for all \( e' \in E_e \). Moreover, for \( \epsilon > 0 \) set \( g_{\epsilon} = (1 + \epsilon) \rho \). Then, there exists a random time \( T < \infty \) such that \( X_t(e') \in [g_{\epsilon}^{-1} \mu(e'), g_{\epsilon} \mu(e')] \) for all \( t \geq T \) and \( e' \in E_e \). In particular, for every \( i \in \{1, 2\} \),
\[ \text{pol}_{v_i,e}(X_t) = \frac{N_i(e')^{\alpha}}{\sum_{e' \in E_{v_i}} N_i(e')^{\alpha}} \geq \frac{(g_{\epsilon}^{-1} \mu(e'))^{\alpha}}{\sum_{e' \in E_{v_i}} (g_{\epsilon} \mu(e'))^{\alpha}} = g_{\epsilon}^{-2\alpha} \frac{\mu(e)^{\alpha}}{\sum_{e' \in E_{v_i}} \mu(e')^{\alpha}}. \]
Therefore, using that \( \mu \) is an equilibrium, \( \{N_i(e)\}_{i \geq T} \) is dominated from below by a Poisson process with intensity
\[ g_{\epsilon}^{-2\alpha} \frac{\mu(e)^{\alpha}}{\sum_{e' \in E_{v_i}} \mu(e')^{\alpha}} + g_{\epsilon}^{-2\alpha} \frac{\mu(e)^{\alpha}}{\sum_{e' \in E_{v_i}} \mu(e')^{\alpha}} = g_{\epsilon}^{-2\alpha} \mu(e). \]
Hence, almost surely, \( X_-(e) \geq g_{\epsilon}^{-2\alpha} \mu(e) \). Similar arguments yield that \( X_+(e) \leq g_{\epsilon}^{2\alpha} \mu(e) \). \( \square \)

The proof of Lemma 3.4 is the most challenging part of the auxiliary results. The main ingredient is Theorem 3.6 below. It states that \( X_-(e) \) is bounded away from 0 with a high probability, even when conditioning on \( F_{c \epsilon} \), where for \( c \in E \) we let
\[ F_{c \epsilon} := \sigma(\{P_{c e} \}_{v \in \cup_{c' \in E_{c \epsilon}, c' \neq c}}) \]
denote the \( \sigma \)-algebra generated by all Poisson processes at nodes that are at distance at least 2 away from the edge \( e \).

**Lemma 3.6 (Compact containment of \( C(e) \)).** For every \( \epsilon > 0 \) there exists \( \delta > 0 \) such that almost surely
\[ \inf_{e \in E} \mathbb{P}(X_-(e) \geq \delta | F_{c \epsilon}) \geq 1 - \epsilon. \]
In particular, \( \mathbb{P}(X_-(e) > 0 | F_{c \epsilon}) = 1 \) for every \( e \in E \).
Hence, we conclude from dependent percolation theory in the form of [8, Theorem 0.0] that the edges violating the lower bound are restricted to finite, well-separated islands. To make the presentation self-contained, we give an elementary direct proof.

**Proof of Lemma 3.4.** Let \( v \in V \) be arbitrary. We resort to a first-moment argument and show that for sufficiently small \( \delta \) the expected number of length-\( n \) self-avoiding paths of \( \delta \)-unstable edges tends to 0 as \( n \to \infty \). Since the number of length-\( n \) self-avoiding paths in \( G \) starting from \( v \) is at most \( \Delta^n \), it suffices to show that the probability for any fixed self-avoiding path to consist of \( \delta \)-unstable edges only is of the order at most \((2\Delta)^{-n}\).

Note that the number of edges that are at graph distance at most 3 of a given edge is bounded above by \(2\Delta^3\). Hence, any self-avoiding path of length \( n \) contains at least \( n/(2\Delta^3) \) edges that are of pairwise distance at least 4, so that by Lemma 3.6, the probability that they are all \( \delta \)-unstable is at most \( \varepsilon(\delta)^n/(2\Delta^3) \). In particular, choosing \( \delta > 0 \) such that \( \varepsilon(\delta)^{1/(2\Delta^3)} \leq 1/(2\Delta) \) concludes the proof. \( \square \)

It remains to prove Lemma 3.6. To that end, we invoke a conditioning argument in the spirit of [1, Lemma 5.2] to provide a bootstrap result propagating large edge weights at a current time point to a considerable duration into the future. For \( k, \ell \geq 1 \) put

\[
ak_k,\ell := \Delta^{-\frac{1}{\alpha}} 2^k(\ell - 1) - \Delta \sum_{i \leq \ell} \alpha^i.
\]

**Lemma 3.7** (Bootstrapped lower bound). There exists a constant \( c = c(\alpha, \Delta) > 0 \) such that for all \( k \) large enough, all \( \ell \geq 1 \) and all \( e \in E \), almost surely,

\[
P(N_{2^{k+\ell+1}}(e) \leq a_k,\ell+1 \text{ and } N_{2^{k+\ell}}(e) \geq a_k,\ell \mid \mathcal{F}_e) \leq e^{-2\varepsilon(1+\alpha)}.
\]

**Proof.** Let \( \varepsilon > 0 \) be arbitrary. Then, the Poisson concentration inequality [9, Lemma 1.2] implies that the event

\[
A_e := \{ N_{2^{k+\ell+1}}(e') \leq 2^{1+\varepsilon+k(\ell+1)} \text{ for all } e' \in E_e \}
\]

has a probability tending to 1 with an error decaying exponentially in \( 2^{\ell+1} \).

Further, under the event \( \{ N_{2^{k+\ell}}(e) \geq a_k,\ell \} \cap A_e \), Equation (2.1) has a lower bound for times \( t \in T_e := [2^{k+\ell}, 2^{k(\ell+1)}] \) given by

\[
\text{pol}_{v,e}(N_t) = \frac{N_t(e)^\alpha}{\sum_{e' \in E_e} N_t(e')^\alpha} \geq \frac{a_k,\ell^\alpha}{a_k,\ell^\alpha + \Delta2^{1+\varepsilon+k(\ell+1)}1^\alpha}.
\]

Since \( \Delta a_k,\ell+1 = a_k,\ell^\alpha 2^{k(\ell+1)1^-1} \), putting \( \alpha_e = \alpha(1 + \varepsilon) \), we deduce that

\[
\text{pol}_{v,e}(N_t) \geq \frac{1}{1 + 2\alpha_e 2^{k(\ell+1)}1^\alpha a_k,\ell+1 2^{k(\ell+1)(1-\alpha)}} = \frac{1}{1 + 2\alpha_e 2^{k(\ell+1)(1-\alpha)}a_k,\ell+1}.
\]

Hence,

\[
\text{pol}_{v,e}(N_t) \geq \frac{2^{-\alpha_e 2^{k(\ell+1)2-\ell-1}}a_k,\ell+1}{1 + 2^{-\alpha_e}} =: G_k,\ell.
\]

In particular, for \( t \in T_e \), the Poisson processes \( P_v \) and \( P_w \) having points in \([2^{k+\ell}, t] \times U_e \) where \( U_e \subseteq [0,1] \) and \( |U_e| \geq G_k,\ell \) implies weight increases of the edge \( e \) (by one or more) in the time interval \([2^{k+\ell}, t]\). Therefore, using Equation (3.1), we find a constant \( c \) such that

\[
P(N_{2^{k+\ell+1}}(e) \leq a_k,\ell+1 \text{ and } N_{2^{k+\ell}}(e) \geq a_k,\ell) \leq P(N_{2^{k+\ell+1}}(e) = N_{2^{k+\ell}}(e) \leq a_k,\ell+1 \text{ and } N_{2^{k+\ell}}(e) \geq a_k,\ell).
\]
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\[ \leq \mathbb{P}(P_{i}(T_{e} \times U_{e}) + P_{0}(T_{e} \times U_{e}) \leq a_{k,t+1} \text{ and } N_{2k^{*}}(e) \geq a_{k,t}) + O(e^{-\alpha k^{2}t}) \]
\[ \leq \mathbb{P}(\text{Poi}(2|T_{e}|G_{k,t}) \leq a_{k,t+1}) + O(e^{-\alpha k^{2}t}). \]

Since

\[ 2|T_{e}|G_{k,t} = a_{k,t+1}2^{1-\alpha + k(1-\alpha)^{a}} \frac{1 - 2^{-k}}{1 + 2^{-k\alpha}} \]

there exists \( \tilde{\alpha} > 1 \) such that \( 2|T_{e}|G_{k,t} > \tilde{\alpha} \cdot a_{k,t+1} \) holds for all \( k \) large enough. Thus, by [9, Lemma 1.2],

\[ \mathbb{P}(N_{2k^{*+1}}(e) \leq a_{k,t+1} \text{ and } N_{2k^{*}}(e) \geq a_{k,t}) \leq \exp((1 - \tilde{\alpha} + \log(\tilde{\alpha}))a_{k,t+1}). \]

Now, we conclude the proof by noting that \( a_{k,t+1} \geq \Delta^{-\frac{1}{2}} 2^{k(1-\alpha)} \) and that the coefficient of \( a_{k,t+1} \) is negative.

Theorem 3.7 propagates the lower bounds on the edge weights through time. As \( t \to \infty \), these lower bounds allow to exclude vanishing edge weights.

Proof of Lemma 3.6. Defining \( F = \cap_{\ell \geq 1}\{N_{2k^{*}}(e) \geq a_{k,t}\} \) and \( \delta_{0} = \liminf_{t \to \infty} a_{k,t}2^{-kt} > 0 \), we note that \( X_{-}(e) \geq \delta_{0} \) holds under the event \( F \). Hence, it remains to establish a lower bound on \( \mathbb{P}(F|\mathcal{F}_{e}) \).

Since \( a_{k+1} \leq 1 \), the bound \( N_{2k}(e) \geq a_{k,1} \) holds for any \( k \geq 1 \). Hence, if the event \( F \) does not occur, then there exists \( t_{0} \geq 1 \) such that \( N_{2k^{*+1}}(e) \leq a_{k,t_{0}+1} \) and \( N_{2k^{*}} \geq a_{k,t_{0}} \).

In particular, by Lemma 3.7, almost surely,

\[ 1 - \mathbb{P}(F|\mathcal{F}_{e}) \leq \sum_{\ell \geq 1} \mathbb{P}(N_{2k^{*+1}}(e) \leq a_{k,t+1} \text{ and } N_{2k^{*}} \geq a_{k,t} | \mathcal{F}_{e}) \leq \sum_{\ell \geq 1} e^{-\alpha k^{2}(1-\alpha)}, \]

which becomes smaller than \( \epsilon \) for sufficiently large \( k > 0 \). \( \square \)

### 4 Regular graphs

Throughout this section, we assume \( G \) to be \( \Delta \)-regular. The key to the proof of parts (2) and (3) of Theorem 2.2 is the following bootstrap property.

**Lemma 4.1** (Bootstrap on regular graphs). Let \( a < 2/\Delta, b > 2/\Delta \) and \( e \in E \) be such that \( \mathcal{C}(e) \subseteq [a, b] \) holds almost surely for all \( e' \in E_{e} \). Furthermore, define the function

\[ f(r, s) := \frac{2r^{\alpha}}{r^{\alpha} + (\Delta - 1)s^{\alpha}}. \]

Then, \( \mathcal{C}(e) \subseteq [f(a, b), f(b, a)] \) holds almost surely. In particular, for \( G = \mathbb{Z} \) there exist \( a' \in [a, 1) \) and \( b' \in (1, b] \) such that \( a' / b' \geq (a/b)^{\alpha} \) and \( \mathcal{C}(e) \subseteq [a', b'] \) holds almost surely.

Before proving Lemma 4.1 we explain how it implies part (3) of Theorem 2.2.

**Proof of part (3) of Theorem 2.2.** First, \( X_{-}(e) \) is strictly positive by Theorem 3.4 using a similar argument to the proof of part (1) of Theorem 2.2. Hence, Theorem 4.1 gives that \( X_{-}(e) \geq X_{-}(e)^{2^{-\alpha}}, i.e., X_{-}(e) \geq 2^{-\alpha/(1-\alpha)} \). In other words, almost surely \( \mathcal{C}(e) \subseteq [a_{1}, b_{1}] \) where \( a_{1} := 2^{-\alpha/(1-\alpha)} \) and \( b_{1} := 2 \). Applying Lemma 4.1 iteratively yields sequences \( \{a_{i}\}_{i \geq 1} \) and \( \{b_{i}\}_{i \geq 1} \) such that

1. \( \{a_{i}\}_{i \geq 1} \) is increasing and bounded above by 1,
2. \( \{b_{i}\}_{i \geq 1} \) is decreasing and bounded below by 1,
3. \( b_{i+1}/a_{i+1} \leq (b_{i}/a_{i})^{\alpha} < b_{i}/a_{i} \), and
4. \( \mathcal{C}(e) \subseteq \bigcap_{i \geq 1} [a_{i}, b_{i}] \) holds almost surely for all \( e \in E \).
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Since the first three items imply that $a_i$ and $b_i$ converge to 1, we arrive at

$$P(C(e) = 1) = P(\bigcap_{i \geq 0} \{C(e) \subseteq [a_i, b_i]\}) = \lim_{t \to \infty} P(C(e) \subseteq [a_i, b_i]) = 1,$$

as asserted. \qed

Next, we prove Lemma 4.1.

**Proof of Lemma 4.1.** Let

$$G_{a,b} := \cap_{e' \in E_{wa}} \{C(e') \subseteq [a, b]\}$$

denote the event that $C(e') \subseteq [a, b]$ holds for all $e' \in E_{wa}$. Under this event, $X_t^{v,w}$ gains mass at a rate of at least

$$a'' = f(at, bt) = f(a, b),$$

for $t$ large enough. More precisely, under $G_{a,b}$ one can find a sequence $\{\varepsilon_t\}_{t \geq 0}$ with $\varepsilon_t \searrow 0$ such that almost surely

$$X_t^{v,w} \geq P_v([0, t] \times U_{v,t}) + P_w([0, t] \times U_{w,t}) + 1$$

for all $t > 0$ where $|U_{v,t}| = |U_{w,t}| = a''/2 - \varepsilon_t$. Analogous arguments to the one in the proof of Theorem 3.3 give that almost surely

$$\liminf_{t \to \infty} X_t^{v,w} \geq a',$$

where $a' := a \lor a''$. Similar arguments give the upper bound $b' := b \land b''$ where $b'' := f(b, a)$. In the special case $\Delta = 2$ (i.e. $G = Z$) we find $a'/b' \geq (a/b)^n$ as desired. \qed

Now, let $G$ be $\Delta$-regular for some $\Delta \geq 2$. In this case we do not have $a'/b' \geq (a/b)^n$ to prove Theorem 2.2, but we still have the sequences $\{a_i\}_{i \geq 1}$ and $\{b_i\}_{i \geq 1}$ with $b_{i+1}/a_{i+1} \leq b_i/a_i$. The idea is to show that for $\alpha$ sufficiently close to $1/2$ the inequality is strict.

**Lemma 4.2.** For $\alpha = 1/2 + \varepsilon$ with $\varepsilon$ small enough and $0 < a < 2/\Delta < b < 2$ we have $f(a, b) > a$ or $f(b, a) < b$.

**Proof.** Assume both statements are wrong, i.e. $f(a, b) \leq a$ and $f(b, a) \geq b$, then this gives

$$\rho \leq \rho^{2\alpha} \frac{1 + \frac{\rho^{-\alpha}}{\Delta - 1}}{1 + \frac{\rho^{1/2-\varepsilon}}{\Delta - 1}}$$

(4.1)

for $\rho := b/a$. Insert $\alpha = 1/2 + \varepsilon$ and take log on both sides to get

$$0 \leq 2\varepsilon \log \rho + \log \left(1 + \frac{\rho^{-1/2-\varepsilon}}{\Delta - 1}\right) - \log \left(1 + \frac{\rho^{1/2+\varepsilon}}{\Delta - 1}\right).$$

Note that equality holds for $\rho = 1$ so if the right-hand side is decreasing in $\rho$ then this contradicts the above inequality for $\rho > 1$. To that end, note that the derivative

$$\frac{1}{\rho} \left(2\varepsilon - (1/2 + \varepsilon) \left(\frac{1}{1 + (\Delta - 1)\rho^{-1/2-\varepsilon}} + \frac{1}{1 + (\Delta - 1)\rho^{1/2+\varepsilon}}\right)\right)$$

of the right-hand side is bounded above for all $\rho \geq 1$ by

$$\frac{1}{\rho} \left(2\varepsilon - \frac{(1/2 + \varepsilon)}{\Delta}\right)$$

which is negative for $\varepsilon$ small enough and hence we get a contradiction to Equation (4.1). \qed
Not improving both bounds at each application of the bootstrap property means that the proof for a uniform lower bound on \( C(e) \) in part (1) of Theorem 2.2 does not work since we do not immediately get a contradiction. The following Lemma helps remedy this.

**Lemma 4.3.** For \( a < 2\Delta^{-1/(1-\alpha)} \) we have \( f(a, 2) > a \).

**Proof.** Straightforward calculation as
\[
 f(a, 2) = \frac{2a^\alpha}{a^\alpha + (\Delta - 1)2^\alpha} > \frac{2a^\alpha}{2a^\alpha} = a \cdot \frac{2^{1-\alpha}}{a^{1-\alpha} \Delta} > a. 
\]

**Proof of part (2) of Theorem 2.2.** Take the \( \epsilon \) from Theorem 4.2 and consider \( \alpha = 1/2 + \epsilon \). \( X_\cdot(e) \) is bounded above by Theorem 3.3 and strictly positive by Theorem 3.4 and Theorem 4.3 using analogous arguments as in the proof of part (1) of this theorem. So we find \( a \leq 2/\Delta \) and \( b \geq 2/\Delta \) such that \( C(e) \subseteq [a, b] \) for all \( e \in E \). We choose \( a \) maximal and \( b \) minimal with that property. To derive a contradiction, assume that \( a < b \). Then, by Theorem 4.1 and Theorem 4.2 we can tighten the bounds as
\[
 C(e) \subseteq [a \lor f(a, b), b \land f(b, a)] \subsetneq [a, b]. 
\]
This contradicts the maximality/minimality of \( a \) and \( b \), and since \( f(a, b) \leq 2/\Delta \leq f(b, a) \) therefore gives \( a = b = 2/\Delta \).

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**References**


