Bipodal structure in oversaturated random graphs

Richard Kenyon∗ Charles Radin† Kui Ren ‡ Lorenzo Sadun§

September 17, 2015

Abstract

We study the asymptotics of large simple graphs constrained by the limiting density of edges and the limiting subgraph density of an arbitrary fixed graph $H$. We prove that, for all but finitely many values of the edge density, if the density of $H$ is constrained to be slightly higher than that for the corresponding Erdős-Rényi graph, the typical large graph is bipodal with parameters varying analytically with the densities. Asymptotically, the parameters depend only on the degree sequence of $H$.

Key words. graph limits, entropy, bipodal structure, phases, universality

1 Introduction

We study the asymptotics of large, simple, labeled graphs constrained to have subgraph densities $\epsilon$ of edges, and $\tau$ of some fixed subgraph $H$ with $\ell \geq 2$ edges. To study the asymptotics we use the graphon formalism of Lovász et al [8, 9, 2, 1, 10] and the large deviations theorem of Chatterjee and Varadhan [5], from which one can reduce the analysis to the study of the graphons which maximize the entropy subject to the density constraints [13, 14, 12, 6]. See definitions in Section 2.

The phase space is the subset of $[0, 1]^2$ consisting of accumulation points of all pairs of densities $\bar{\tau} = (\epsilon, \tau)$ achievable by finite graphs. (See Figure 1 for the case where $H$ is a triangle.) Within the phase space is the ‘Erdős-Rényi curve’ (ER curve) $\{(\epsilon, \tau) \mid \tau = \epsilon^\ell\}$, attained when edges are chosen independently. In this paper we study the typical behavior of large graphs for $\tau$ just above the ER curve. We will show that the qualitative behavior of such graphs is the same for all choices of $H$ and for all but finitely many choices of $\epsilon$ depending on $H$.

To be precise, we show that for fixed $H$, for $\epsilon$ outside a finite set, and for $\tau$ close enough to $\epsilon^\ell$, there is a unique entropy-maximizing graphon (up to measure-preserving transformations of the unit interval); furthermore it is bipodal and depends analytically on $(\epsilon, \tau)$, implying

∗Department of Mathematics, Brown University, Providence, RI 02912; rkenyon at math.brown.edu
†Department of Mathematics, University of Texas, Austin, TX 78712; radin@math.utexas.edu
‡Department of Mathematics and ICES, University of Texas, Austin, TX 78712; ren@math.utexas.edu
§Department of Mathematics, University of Texas, Austin, TX 78712; sadun@math.utexas.edu
that the entropy is an analytic function of $(\epsilon, \tau)$. In particular we prove the existence of one or more well-defined thermodynamic phases just above the ER curve. This is the first proof, as far as we know, of the existence of a phase in any constrained-density graphon model, where by phase we mean a (maximal) open set in the phase space where the entropy varies analytically with the constraint parameters. Conjecturally, phases form an open dense subset of the phase space.

A bipodal graphon is a function $g : [0, 1]^2 \to [0, 1]$ of the form:

$$g(x, y) = \begin{cases} 
    p_{11} & x, y < c, \\
    p_{12} & x < c < y, \\
    p_{12} & y < c < x, \\
    p_{22} & x, y > c.
\end{cases} \quad (1)$$

Here $c, p_{11}, p_{12}$ and $p_{22}$ are constants taking values between 0 and 1. We prove that as $\tau \searrow \epsilon$, the parameters $c \to 0$, $p_{22} \to \epsilon$, and $p_{11}$ and $p_{12}$ approach the solutions of a problem in single-variable calculus. The inputs to that calculus problem depend only on the degrees of the vertices of $H$.

We say that a finite graph $H$ is $k$-starlike if all the vertices of $H$ have degree $k$ or 1, where $k > 1$ is a fixed integer. $k$-starlike graphs include $k$-stars (where one vertex has degree $k$ and $k$ vertices have degree 1), and the complete graph on $k + 1$ vertices. For fixed $k$, all $k$-starlike graphs behave essentially the same for our asymptotics. We prove our results first for $k$-stars, and then apply perturbation theory to show that the differences between different $k$-starlike graphs are irrelevant, and then prove the general case.

To state our results more precisely, we need some notation. Let

$$S_0(w) = -\frac{1}{2}[w \log w + (1 - w) \log(1 - w)], \quad (2)$$

and define the graphon entropy (or entropy for short) of a graphon $g$ to be

$$s(g) = \int_0^1 \int_0^1 S_0(g(x, y)) dx dy. \quad (3)$$
Let
\[ \psi_k(\epsilon, \bar{\epsilon}) = \frac{2[S_0(\bar{\epsilon}) - S_0(\epsilon) - S'_0(\epsilon)(\bar{\epsilon} - \epsilon)]}{\bar{\epsilon}^k - \epsilon^k - k\epsilon^{k-1}(\bar{\epsilon} - \epsilon)}. \] (4)

This function has a removable singularity at \( \bar{\epsilon} = \epsilon \), which we will fill by defining
\[ \psi_k(\epsilon, \epsilon) = \frac{2S'_0(\epsilon)}{k(k-1)\epsilon^{k-2}}. \] (5)

For fixed \( \epsilon \), let \( \zeta_k(\epsilon) \) be the value of \( \bar{\epsilon} \) that maximizes \( \psi_k(\epsilon, \bar{\epsilon}) \). (We will prove that this maximizer is unique and depends continuously on \( \epsilon \).)

**Theorem 1.1.** Let \( H \) be a \( k \)-starlike graph with \( \ell \geq 2 \) edges. Let \( \epsilon \in (0, 1) \) be any point other than \( (k-1)/k \). Then there is a number \( \tau_0 > \epsilon^\ell \) (depending on \( \epsilon \)) such that for all \( \tau \in (\epsilon^\ell, \tau_0) \), the entropy-maximizing graphon at \( (\epsilon, \tau) \) is unique (up to measure-preserving transformations of \([0, 1]\)) and bipodal. The parameters \( (c, p_{11}, p_{12}, p_{22}) \) are analytic functions of \( \epsilon \) and \( \tau \) on the region \( \epsilon \neq (k-1)/k, \tau \in (\epsilon^\ell, \tau_0(\epsilon)) \). Furthermore, as \( \tau \searrow \epsilon^\ell \) we have that \( p_{22} \to \epsilon, p_{12} \to \zeta_k(\epsilon), p_{11} \) satisfies \( S'_0(p_{11}) = 2S'_0(p_{12}) - S'_0(p_{22}) \), and \( c = O(\tau - \epsilon^\ell) \).

Theorem 1.1 proves that there is part of a phase just above the ER curve for \( \epsilon < (k-1)/k \) and also for \( \epsilon > (k-1)/k \); numerical evidence suggests these are in fact parts of a single phase; the only ‘singular’ behavior is the manner in which the graphon approaches the constant graphon associated with the ER curve. We will see in Theorem 1.2 that this behavior is only slightly more complicated for general \( H \) than it is for \( k \)-starlike \( H \).

When \( H \) has vertices with different degrees \( > 1 \), the problem resembles that of a formal positive linear combination of \( k \)-stars. As in the \( k \)-starlike case, we first solve the problem for the linear combination of \( k \)-stars and then use perturbation theory to extend the results to arbitrary \( H \).

**Theorem 1.2.** Let \( H \) be an arbitrary graph with \( \ell \) edges with at least one vertex of degree \( 2 \) or greater. Then there exists a finite set \( B_H \subset (0, 1) \) such that if \( \epsilon \notin B_H \), then there is a number \( \tau_0 > \epsilon^\ell \) (depending on \( \epsilon \)) such that for all \( \tau \in (\epsilon^\ell, \tau_0) \), the entropy-maximizing graphon at \( (\epsilon, \tau) \) is unique (up to measure-preserving transformations of \([0, 1]\)) and bipodal. The parameters \( (c, p_{11}, p_{12}, p_{22}) \) are analytic functions of \( \epsilon \) and \( \tau \) on the region \( \epsilon \notin B_H, \tau \in (\epsilon^\ell, \tau_0(\epsilon)) \). Furthermore, as \( \tau \searrow \epsilon^\ell \) we have that \( p_{22} \to \epsilon, p_{12} \to \zeta_k(\epsilon), p_{11} \) satisfies \( S'_0(p_{11}) = 2S'_0(p_{12}) - S'_0(p_{22}) \), and \( c = O(\tau - \epsilon^\ell) \).

The key differences between the Theorems 1.1 and 1.2 are:

- For \( k \)-starlike graphs, the set \( B_H \) of bad values of \( \epsilon \) consists of a single point, and this point is explicitly known: \( \epsilon = (k-1)/k \).
- For \( k \)-starlike graphs, the behavior of \( \zeta_k \) is explicit. It is a continuous and strictly decreasing function of \( \epsilon \), and gives an involution of \((0, 1)\). (That is, \( \zeta_k(\zeta_k(\epsilon)) = \epsilon \).) For \( k = 2 \) it is given by \( \zeta_2(\epsilon) = 1 - \epsilon \). In the general case, the limiting value of \( p_{12} \), and its dependence on \( \epsilon \), appear to be much more complicated. We do not know whether this limiting value is always continuous across the bad set \( B_H \).
The organization of this paper is as follows. In Section 2 we review the formalism of graphons and establish basic notation. In Section 3 we establish a number of technical results for $k$-star models. Using these results, in Section 4 we prove Theorem 1.1 for the case that $H$ is a $k$-star. In Section 5 we show that just above the ER curve a model with an arbitrary $k$-starlike $H$ can be approximated by a $k$-star model. By bounding the error terms, we prove Theorem 1.1 in full generality. In Section 6 we consider formal positive linear combinations of $k$-stars, and prove a theorem much like Theorem 1.2 for those models. Finally, in Section 7 we show that the model for an arbitrary $H$ can be approximated by a formal linear combination of $k$-stars, thus completing the proof of Theorem 1.2.

2 Notation and background

We consider a simple graph $G$ (undirected, with no multiple edges or loops) with a vertex set $V(G)$ of labeled vertices. For a subgraph $H$ of $G$, let $T_H(G)$ be the number of maps from $V(H)$ into $V(G)$ which preserve edges. The density $\tau_H(G)$ of $H$ in $G$ is then defined to be

$$\tau_H(G) := \frac{|T_H(G)|}{n^{|V(H)|}},$$

(6)

where $n = |V(G)|$. An important special case is where $H$ is a ‘$k$-star’, a graph with $k$ edges, all with a common vertex, for which we use the notation $\tau_k(G)$. In particular $\tau_1(G)$, which we also denote $\epsilon(G)$, is the edge density of $G$.

For $\alpha > 0$ and $\bar{\tau} = (\epsilon, \tau_H)$ define $Z_{\bar{\tau}}^{n,\alpha}$ to be the number of graphs $G$ on $n$ vertices with densities satisfying

$$\epsilon(G) \in (\epsilon - \alpha, \epsilon + \alpha), \quad \tau_H(G) \in (\tau_H - \alpha, \tau_H + \alpha).$$

(7)

Define the (constrained) entropy $s_{\bar{\tau}}$ to be the exponential rate of growth of $Z_{\bar{\tau}}^{n,\alpha}$ as a function of $n$:

$$s_{\bar{\tau}} = \lim_{\alpha \searrow 0} \lim_{n \to \infty} \frac{\ln(Z_{\bar{\tau}}^{n,\alpha})}{n^2}.$$  

(8)

The double limit defining the entropy $s_{\bar{\tau}}$ is known to exist [13]. To analyze it we make use of a variational characterization of $s_{\bar{\tau}}$, and for this we need further notation to analyze limits of graphs as $n \to \infty$. (This work was recently developed in [8, 9, 2, 1, 10]; see also the recent book [11].) The (symmetric) adjacency matrices of graphs on $n$ vertices are replaced, in this formalism, by symmetric, measurable functions $g : [0,1]^2 \to [0,1]$; the former are recovered by using a partition of $[0,1]$ into $n$ consecutive subintervals. The functions $g$ are called graphons.

For a graphon $g$ define the degree function $d(x)$ to be $d(x) = \int_0^1 g(x, y)dy$. The $k$-star density of $g$, $\tau_k(g)$, then takes the simple form

$$\tau_k(g) = \int_0^1 d(x)^k dx.$$  

(9)

For any fixed graph $H$, the $H$-density $\tau_H$ of $g$ can be similarly expressed as an integral of a product of factors $g(x_i, x_j)$.

The following is Theorem 4.1 in [14]:

4
Theorem 2.1 (The Variational Principle). For any feasible set $\bar{\tau}$ of values of the densities $\bar{\tau}(g) := (\epsilon, \tau_H)$ we have $s_\tau = \max[s(g)]$, where the entropy is maximized over all graphons $g$ with $\bar{\tau}(g) = \bar{\tau}$.

(Instead of using $s(g)$, some authors use the rate function $I(g) := -s(g)$, and then minimize $I$.) The existence of a maximizing graphon $g = g_\tau$ for any constraint $\bar{\tau}(g) = \bar{\tau}$ was proven in [13], again adapting a proof in [5]. If the densities are that of edges and $k$-star subgraphs we refer to this maximization problem as a star model, though we emphasize that the result applies much more generally [13, 14].

We consider two graphs equivalent if they are obtained from one another by relabeling the vertices. For graphons, the analogous operation is applying a measure-preserving map $\psi$ of $[0, 1]$ into itself, replacing $g(x, y)$ with $g(\psi(x), \psi(y))$, see [11]. The equivalence classes of graphons under relabeling are called reduced graphons, and graphons are equivalent if and only if they have the same subgraph densities for all possible finite subgraphs [11]. In the remaining sections of the paper, whenever we claim that a graphon has a property (e.g. monotonicity in $x$ and $y$, or uniqueness as an entropy maximizer), the caveat “up to relabeling” is implied.

The graphons which maximize the constrained entropy can tell us what ‘most’ or ‘typical’ large constrained graphs are like: if $g_\tau$ is the only reduced graphon maximizing $S(g)$ with $\bar{\tau}(g) = \bar{\tau}$, then as the number $n$ of vertices diverges and $\alpha_n \to 0$, exponentially most graphs with densities $\bar{\tau}(G) \in (\tau_i - \alpha_n, \tau_i + \alpha_n)$ will have reduced graphon close to $g_\tau$ [13]. This is based on large deviations from [5]. We emphasize that this interpretation requires that the maximizer be unique; this has been difficult to prove in most cases of interest and is an important focus of this work.

A graphon $g$ is called $M$-podal if there is decomposition of $[0, 1]$ into $M$ intervals (‘vertex clusters’) $C_j$, $j = 1, 2, \ldots, M$, and $M(M + 1)/2$ constants $p_{ij}$ such that $g(x, y) = p_{ij}$ if $(x, y) \in C_i \times C_j$ (and $p_{ji} = p_{ij}$). We denote the length of $C_j$ by $c_j$.

3 Technical properties of star models

For each star model, all entropy-maximizing graphons are multipodal with a fixed upper bound on the number of clusters, also called the podality [6]. For any fixed podality $M$, an $M$-podal graphon is described by $N = M(M + 3)/2$ parameters, namely the values $p_{ij}$ ($1 \leq i \leq j \leq M$) and the widths $c_i$ ($1 \leq i \leq M$) of the clusters. When it does not cause confusion, we will use $g$ to denote the vector

$$(c_1, \ldots, c_M, p_{11}, \ldots, p_{1M}, p_{22}, \ldots, p_{2M}, \ldots, p_{M-1M-1}, p_{M-1M}, p_{MM}), \quad (10)$$

which contains all these parameters. The problem of optimizing the graphon then reduces to a finite-dimensional calculus problem. To be precise, let us recall that for an $M$-podal graphon, we have

$$\epsilon(g) = \sum_{1 \leq i, j \leq M} c_i c_j p_{ij}, \quad \tau_k(g) = \sum_{1 \leq i \leq M} c_i d_i^k, \quad s(g) = \sum_{1 \leq i, j \leq M} c_i c_j S_0(p_{ij}), \quad (11)$$

5
where $d_i = \sum_{1 \leq j \leq M} c_j p_{ij}$ is the value of the degree function on the $i$th cluster. The problem of searching for entropy-maximizing graphons with fixed edge density $\epsilon$ and $k$-star density $\tau_k$ can now be formulated as

$$\max_{g \in [0,1]^N} s(g), \quad \text{subject to: } \epsilon(g) - \epsilon = 0, \quad \tau_k(g) - \tau = 0, \quad C(g) = 1. \quad (12)$$

where $C(g) = \sum_{1 \leq j \leq M} c_j$.

The following result says that the maximization problem (12) can be solved using the method of Lagrange multipliers. The existence of finite Lagrange multipliers was previously established in [6], treating the space of graphons as a linear space of functions $[0,1]^2 \rightarrow [0,1]$, intuitively considering perturbations of graphons localized about points in $[0,1]^2$. For star models we may restrict to $M$-podal graphons, as noted above, and thus consider perturbations in the relevant parameters $p_{ij}$ and $c_j$.

**Lemma 3.1.** Let $g$ be a local maximizer in (12). Then for constraints $\epsilon, \tau$ off the ER curve, there exist unique $\alpha, \beta, \gamma \in \mathbb{R}$ such that

$$\nabla s(g) - \alpha \nabla \epsilon(g) - \beta \nabla \tau_k(g) - \gamma \nabla C(g) = 0. \quad (13)$$

We do not include the proof, which follows easily from that of Lemma 3.5 in [6]. We also note that one can remove the variable $c_M$ and the constraint $C(g) = 1$, eliminating the multiplier $\gamma$.

For convenience later, we now write down the exact form of the Euler-Lagrange equation (13). We first verify that

$$\frac{\partial \epsilon}{\partial p_{ij}} = A_{ij}, \quad \frac{\partial \epsilon}{\partial c_i} = 2 \sum_{j=1}^{M} c_j p_{ij} = 2d_i, \quad (14)$$

$$\frac{\partial \tau_k}{\partial p_{ij}} = k \frac{d_i^{k-1} + d_j^{k-1}}{2} A_{ij}, \quad \frac{\partial \tau_k}{\partial c_i} = d_i^k + k \sum_{j=1}^{M} c_j d_j^{k-1} p_{ij}, \quad (15)$$

$$\frac{\partial C}{\partial p_{ij}} = 0, \quad \frac{\partial C}{\partial c_i} = 1, \quad (16)$$

$$\frac{\partial s}{\partial p_{ij}} = S_0'(p_{ij}) A_{ij}, \quad \frac{\partial s}{\partial c_i} = 2 \sum_{j=1}^{M} c_j S_0(p_{ij}), \quad (17)$$

where $A_{ij} = 2c_i c_j$ if $i \neq j$ and $A_{ij} = c_i^2$ if $i = j$. We can then write down (13) explicitly as

$$S_0'(p_{ij}) = \alpha + \beta \frac{k}{2} (d_i^{k-1} + d_j^{k-1}), \quad 1 \leq i \leq j \leq M \quad (18)$$

$$2 \sum_{j=1}^{M} c_j S_0(p_{ij}) = 2 \alpha d_i + \beta (d_i^k + k \sum_{j=1}^{M} c_j d_j^{k-1} p_{ij}) + \gamma, \quad 1 \leq i \leq M \quad (19)$$

These Euler-Lagrange equations, together with the constraints,

$$\epsilon(g) - \epsilon = 0, \quad \tau_k(g) - \tau = 0, \quad C(g) - 1 = 0, \quad (20)$$

We do not include the proof, which follows easily from that of Lemma 3.5 in [6]. We also note that one can remove the variable $c_M$ and the constraint $C(g) = 1$, eliminating the multiplier $\gamma$. For convenience later, we now write down the exact form of the Euler-Lagrange equation (13). We first verify that
are the optimality conditions for the maximization problem (12). In principle, we can solve this system to find the maximizer $g$.

Next we consider the significance of the Lagrange multipliers $\alpha$ and $\beta$. Suppose that $g_0$ is the unique entropy maximizer for $\epsilon = \epsilon_0$ and $\tau = \tau_0$. Then any sequence of graphons that maximize entropy for $(\epsilon, \tau)$ approaching $(\epsilon_0, \tau_0)$ must approach $g_0$: this follows from upper semicontinuity of the entropy and the fact that we can perturb $g_0$ to any nearby $(\epsilon, \tau)$ by changing some $p_{ij}$. But if $g = g_0 + \delta g$, then

$$s(g) = s(g_0) + ds_{g_0}(\delta g) + O(\delta g^2)$$

$$= s(g_0) + \alpha d\epsilon_{g_0}(\delta g) + \beta d\tau_{g_0}(\delta g) + O(\delta g^2)$$

$$= s(g_0) + \alpha(\epsilon - \epsilon_0) + \beta(\tau - \tau_0) + O(\delta g^2).$$

That is, $\partial s(\epsilon, \tau)/\partial \epsilon = \alpha$ and $\partial s(\epsilon, \tau)/\partial \tau = \beta$.

If $g_0$ is not a unique entropy maximizer, then we only have 1-sided (directional) derivatives:

Lemma 3.2. The function $s(\epsilon, \tau)$ admits directional derivatives in all directions at all points $(\epsilon, t)$ in the interior of the profile.

Proof. The change in entropy in a given direction is obtained by maximizing $ds = \alpha d\epsilon + \beta d\tau$ over all entropy maximizers at $(\epsilon_0, \tau_0)$. That is, when fixed $\epsilon$ and increasing $\tau$, we get the largest $\beta$ of all the graphons that maximize entropy at $(\epsilon_0, \tau_0)$, and when decreasing $\tau$ we get the smallest $\beta$. Likewise, when increasing or decreasing $\epsilon$ we get the largest or smallest values of $\alpha$, and when doing a directional derivative in the direction $(v_1, v_2)$, we get the largest value of $v_1 \alpha + v_2 \beta$. \qed

Existence of directional derivatives implies the fundamental theorem of calculus, so for fixed $\epsilon$ we can write

$$s(\epsilon, \tau) = s(\epsilon, \epsilon^k) + \int_{\epsilon}^{\tau} \beta(g_{\text{max}}(\epsilon, \tau)) d\tau,$$

where $g_{\text{max}}(\epsilon, \tau)$ is the entropy-maximizing graphon at $(\epsilon, \tau)$ that maximizes its right derivative (with respect to $\tau$).

Before proving Theorem 1.1 for $k$-stars, we record some properties of the function $\psi_k(\epsilon, \bar{\epsilon})$ of (4) and its critical points.

Theorem 3.3. For fixed $k$ and $\epsilon$, there is a unique solution to $\partial \psi'_k(\epsilon, \bar{\epsilon})/\partial \bar{\epsilon} = 0$, which we denote $\bar{\epsilon} = \zeta_k(\epsilon)$. The function $\zeta_k$ is a strictly decreasing, with nowhere-vanishing derivative and with fixed point at $\epsilon = (k - 1)/k$. Furthermore, $\zeta_k$ is an involution: $\bar{\epsilon} = \zeta_k(\epsilon)$ if and only if $\epsilon = \zeta_k(\bar{\epsilon})$.

Even though the proof is elementary we will need some parts of it later, so we give it here.

Proof. Fix $k \geq 2$ and let

$$N(\epsilon, \bar{\epsilon}) = 2[S_0(\bar{\epsilon}) - S_0(\epsilon) - S'_0(\epsilon)(\bar{\epsilon} - \epsilon)]$$

$$D(\epsilon, \bar{\epsilon}) = \bar{\epsilon}^k - \epsilon^k - k\epsilon^{k-1}(\bar{\epsilon} - \epsilon)$$

(23)
be the numerator and denominator of the function $\psi_k(\epsilon, \bar{\epsilon}) = N/D$. Note that these definitions make sense for all real values of $k$, not just for integers. When taking derivatives of $N$, $D$ and $\psi$, we will denote a derivative with respect to the first variable by a dot, and a derivative with respect to the second variable by $'$. That is, $D'(\epsilon, \bar{\epsilon}) = \partial D/\partial \bar{\epsilon}$ and $\dot{D}(\epsilon, \bar{\epsilon}) = \partial D/\partial \epsilon$. As noted earlier, this definition of $\psi_k$ has a removable singularity at $\bar{\epsilon} = \bar{\epsilon}$, which we fill in by defining

$$\psi_k(\epsilon, \epsilon) = \frac{N''(\epsilon, \epsilon)}{D''(\epsilon, \epsilon)} = \frac{2S_0''(\epsilon)}{[k(k - 1)\epsilon^{k-2}]}.$$  

The denominator $D$ vanishes only at $\bar{\epsilon} = \epsilon$.

Some useful explicit derivatives are:

$$N' = 2[S_0'(\bar{\epsilon}) - S_0'(\epsilon)], \quad N'' = 2S_0''(\bar{\epsilon}) = \frac{-1}{\epsilon(1 - \epsilon)},$$

$$\dot{N} = -2S_0''(\epsilon)(\bar{\epsilon} - \epsilon), \quad \dot{N}' = -2S_0''(\epsilon),$$

$$D' = k[\epsilon^{k-1} - \epsilon^{k-1}], \quad D'' = k(k - 1)\epsilon^{k-2},$$

$$\dot{D} = -k(k - 1)\epsilon^{k-2}(\bar{\epsilon} - \epsilon), \quad \dot{D}' = -k(k - 1)\epsilon^{k-2}.$$  

Note that $D$ and $N$ both vanish when $\bar{\epsilon} = \epsilon$, so we can write

$$N(\epsilon, \bar{\epsilon}) = \int_\epsilon^{\bar{\epsilon}} N'(\epsilon, x)dx = \int_\epsilon^{\bar{\epsilon}} \dot{N}(x, \bar{\epsilon})dx,$$  

and similarly for $D(\epsilon, \bar{\epsilon})$.

We proceed in steps:

Step 1. Analyzing $\psi$ near $\bar{\epsilon} = \epsilon$ to see that $\psi_k'(-\epsilon, \epsilon) = 0$ only when $\epsilon = (k - 1)/k$.

Step 2. Showing that we can never have $\psi_k' = \psi_k'' = 0$.

Step 3. Showing that the equation $\psi_k'(\epsilon, \bar{\epsilon})$ is symmetric in $\epsilon$ and $\bar{\epsilon}$, implying that $\zeta_k$ is an involution.

Step 4. Showing that $\psi_k$ has a unique critical point.

Step 5. Showing that $d\zeta_k/d\epsilon$ is never zero.

The following calculus fact will be used repeatedly. When $D \neq 0$, $\psi_k' = 0$ is equivalent to $N/D = N'/D'$, and $\psi_k'' = 0$ is equivalent to $N/D = N'/D' = N''/D''$. This follows from the quotient rule:

$$\psi' = \frac{DN' - ND'}{D^2},$$

$$\psi'' = \frac{DN'' - ND''}{D^2} - 2\frac{D'(DN' - ND')}{D^3}.$$  

(27)
Explicitly, this becomes

\[ \psi_k(\epsilon, \tilde{\epsilon}) = \frac{N''(\epsilon, \epsilon)(\epsilon - \epsilon)^2/2 + N'''(\epsilon, \epsilon)(\epsilon - \epsilon)^3/6 + \cdots}{D''(\epsilon, \epsilon)(\epsilon - \epsilon)^2/2 + D'''(\epsilon, \epsilon)(\epsilon - \epsilon)^3/6 + \cdots} = \frac{N''(\epsilon, \epsilon) + N'''(\epsilon, \epsilon)(\epsilon - \epsilon)/3 + \cdots}{D''(\epsilon, \epsilon) + D'''(\epsilon, \epsilon)(\epsilon - \epsilon)/3 + \cdots}. \]  

(28)

\( \psi_k'(\epsilon, \epsilon) = 0 \) is then equivalent to

\[ \frac{-k(k-1)(k-2)e^{k-3}}{\epsilon(1-\epsilon)} = \frac{-k(k-1)e^{k-2}(1-2\epsilon)}{\epsilon^2(1-\epsilon)^2} \]

\[ (k-2)(1-\epsilon) = 1 - 2\epsilon \]

\[ k\epsilon = k - 1. \]  

(29)

**Step 1.** Since \( N \) and \( D \) have double roots at \( \epsilon = \epsilon \), we can do a Taylor series for both of them near \( \epsilon = \epsilon \):

\[ 0 = N' D'' - D' N'' = \int_{\tilde{\epsilon}}^{\epsilon} D''(\epsilon, \tilde{\epsilon})\dot{N}(x, \tilde{\epsilon}) - N''(\epsilon, \tilde{\epsilon})\dot{D}'(x, \tilde{\epsilon})dx. \]  

(30)

Explicitly, this becomes

\[ 0 = \int_{\tilde{\epsilon}}^{\epsilon} \frac{k(k-1)}{\epsilon(1-\epsilon)x(1-x)} \left[ \tilde{\epsilon}^{k-1}(1-\tilde{\epsilon}) - x^{k-1}(1-x) \right] dx. \]  

(31)

The function \( x^{k-1}(1-x) \) has a single maximum at \( x = (k-1)/k \). If both \( \epsilon \) and \( \tilde{\epsilon} \) are on the same side of this maximum, then the integrand will have the same sign for all \( x \) between \( \tilde{\epsilon} \) and \( \epsilon \), and the integral will not be zero. Thus we must have \( \epsilon < (k-1)/k < \tilde{\epsilon} \), or vice-versa, and we must have \( \epsilon^{k-1}(1-\epsilon) < \tilde{\epsilon}^{k-1}(1-\tilde{\epsilon}) \). Note that in this case the integrand changes sign exactly once.

Now we apply the same sort of analysis to the other equation:

\[ 0 = N D'' - D N'' = \int_{\tilde{\epsilon}}^{\epsilon} D''(x, \tilde{\epsilon})\dot{N}(x, \tilde{\epsilon}) - N''(x, \tilde{\epsilon})\dot{D}(x, \tilde{\epsilon})dx. \]  

(32)

Explicitly, this becomes

\[ 0 = \int_{\tilde{\epsilon}}^{\epsilon} \frac{k(k-1)}{\epsilon(1-\epsilon)x(1-x)} \left[ \tilde{\epsilon}^{k-1}(1-\tilde{\epsilon}) - x^{k-1}(1-x) \right] (\epsilon - x)dx. \]  

(33)

This is the same integral as before, only with an extra factor of \((\epsilon - x)\). If we view the first integral (31) as a mass distribution (with total mass zero), then the second integral is (minus) the first moment of this mass distribution relative to the endpoint \( \tilde{\epsilon} \). But we have already seen that the distribution changes sign exactly once, and so must have a non-zero first moment. This is a contradiction.
Step 3. If \( ND' = DN' \), then \( N/D = N'/D' \). Call this common ration \( r \). Then

\[
N = rD \quad \text{and} \quad N' = rD'.
\]

(34)

Note that \( N' \) and \( D' \) are odd under interchange of \( \epsilon \) and \( \bar{\epsilon} \), so the second equation is invariant under this interchange. Furthermore, we have \((\bar{\epsilon} - \epsilon)N' - N = r[(\bar{\epsilon} - \epsilon)D' - D]\). However, \((\bar{\epsilon} - \epsilon)N' - N\) is the same as \( N \) with the roles of \( \epsilon \) and \( \bar{\epsilon} \) reversed, while \((\bar{\epsilon} - \epsilon)D' - D\) is the same as \( D \) with the roles of \( \epsilon \) and \( \bar{\epsilon} \) reversed. Thus the two equations are satisfied for \((\epsilon, \bar{\epsilon})\) if and only if they are satisfied for \((\bar{\epsilon}, \epsilon)\).

Step 4. For \( k = 2 \) we explicitly compute that \( \psi' \) = 0 only at \( \bar{\epsilon} = 1 - \epsilon \). If \( k_{min} \) is the infimum of all values of \( k \) for which \( \psi_k \) has multiple critical points, then at a critical point of \( \psi_{k_{min}} \) we must have \( \psi_k' = \psi_k'' = 0 \), which is a contradiction. Thus \( k_{min} \) does not exist, and \( \psi_k \) has a unique critical point for all \( k \geq 2 \). In particular, \( \zeta_k \) is a well-defined function.

Step 5. The function \( \zeta_k \) is defined by the condition that \( DN' - ND' = 0 \) (and \( \bar{\epsilon} \neq \epsilon \), except when \( \epsilon = (k - 1)/k \)). Let \( f(\bar{\epsilon}, \epsilon) = DN' - ND' = D^2\psi' \). Moving along the curve \( \bar{\epsilon} = \zeta_k(\epsilon) \) (that is, \( f = 0 \)), we differentiate implicitly:

\[
0 = df = \dot{f}d\epsilon + f'd\bar{\epsilon},
\]

(35)

so

\[
\frac{d\bar{\epsilon}}{d\epsilon} = -\frac{\dot{f}}{f'}.
\]

(36)

We compute \( f' = DN'' - ND'' \). This is nonzero by Step 2. We also have

\[
\dot{f} = D\dot{N}' - \dot{N}D' + \dot{D}N' - N\dot{D}'
\]

\[
= -2S_0''(\epsilon)(D - (\bar{\epsilon} - \epsilon)D) + k(k - 1)\epsilon^{k-2}(N - (\bar{\epsilon} - \epsilon)N')
\]

\[
= 2S_0''(\epsilon)(\epsilon^{k-2} + (\bar{\epsilon} - \epsilon)\epsilon^{k-1}) - 2k(k - 1)\epsilon^{k-2}[S_0(\epsilon) - S_0(\bar{\epsilon}) + (\bar{\epsilon} - \epsilon)S_0'(\bar{\epsilon})]
\]

\[
= \dot{D}(\bar{\epsilon}, \epsilon)N''(\bar{\epsilon}, \epsilon) - N(\bar{\epsilon}, \epsilon)D''(\bar{\epsilon}, \epsilon).
\]

(37)

The arguments in the last line are written in the correct order! That is, \( \dot{f} \) is the same as \( f' \), only with the roles of \( \epsilon \) and \( \bar{\epsilon} \) reversed. Since the equation \( f = 0 \) is symmetric in \( \epsilon \) and \( \bar{\epsilon} \), the argument of Step 2 can be repeated to show that \( \dot{f} \neq 0 \).

Since \( d\bar{\epsilon}/d\epsilon \) is never zero, and since \( d\bar{\epsilon}/d\epsilon = -1 \) at the fixed point (by symmetry), \( \zeta_k'(\epsilon) = d\bar{\epsilon}/d\epsilon \) must always be negative.

\[\square\]

4 Theorem 1.1 for \( k \)-stars

Theorem 4.1. Let \( H \) be a \( k \)-star and suppose that \( \epsilon \neq (k - 1)/k \). Then there exists a number \( \tau_0 > \epsilon^k \) such that for all \( \tau \in (\epsilon^k, \tau_0) \), the entropy-optimizing graphon at \((\epsilon, \tau)\) is unique and bipodal. The parameters \((c, p_{11}, p_{12}, p_{22})\) are analytic functions of \( \epsilon \) and \( \tau \). As \( \tau \) approaches \( \epsilon^k \) from above, \( p_{22} \to \epsilon \), \( p_{12} \to \zeta_k(\epsilon) \), \( p_{11} \) satisfies \( S_0'(p_{11}) = 2S_0'(p_{12}) - S_0(p_{22}) \) and \( c = O(\tau - \epsilon^k) \).
Proof. The entropy-maximizing graphon for each \((\epsilon, \tau)\) is multipodal \([6]\), and the parameters \(\{c_j\}\) and \(\{p_{12}\}\) must satisfy the optimality conditions \((18), (19)\). The first step of the proof is to estimate the terms in the optimality equations to within \(o(1)\). This will determine the solutions to within \(o(1)\) and demonstrate that our optimizing graphon is close to bipodal of the desired form. The second step, based on a separate argument, will show that the optimizer is exactly bipodal. The third step shows that the optimizer is in fact unique.

In doing our asymptotic analysis, our small parameter is \(\Delta \tau := \tau - \epsilon^k\). But we could just as well use \(\Delta s := s(g) - S_0(\epsilon)\) or the squared \(L^2\) norm of \(\Delta g := g - g_0\), where \(g_0(x, y) = \epsilon\) (here \(g\) denotes the graphon as a function \([0, 1]^2 \rightarrow [0, 1]\), not a vector of multipodal parameters.) We claim that these are all of the same order. Through arguments found in \([14]\), one can bound \(\Delta \tau\) above by a constant times \(\|\Delta g\|^2\), and bound \(|\Delta s|\) below by a multiple of \(\|\Delta g\|^2\). By considering a bipodal graphon with \(p_{11} = p_{12} = \zeta_k(\epsilon)\) and \(p_{22}\) close to \(\epsilon\), we can bound \(|\Delta s|\) above by a constant times \(\Delta \tau\). This shows that \(O(\Delta s) = O(\Delta \tau)\), and \(O(\|\Delta g\|^2)\) is trapped in between.

Order the clusters so that the largest cluster is the last cluster (of length \(c_M\)). By subtracting the equation \((19)\) for \(c_M\) from the equations for \(c_j\), we eliminate \(\gamma\) from our equations:

\[
S'_0(p_{ij}) = \alpha + \frac{k}{2} \beta (d_i^{-1} + d_j^{-1}) \\
2 \sum_{j=1}^{M} c_j (S_0(p_{ij}) - S_0(p_{Mj})) = 2\alpha(d_i - d_M) + \beta \left(d_i^{k} - d_M^{k} + k \sum_{j=1}^{M} c_j d_j^{-1}(p_{ij} - p_{Mj})\right).
\]

**Step 1.** Since \(\|\Delta g\|\) is small, the area of the region where \(g(x, y)\) differs substantially from \(\epsilon\) must be small. Thus all clusters must either have \(d_i\) close to \(\epsilon\) or \(c_i\) close to zero (or both). We call a cluster Type I if \(c_i\) is close to 0 and Type II if \(d_i\) is close to \(\epsilon\). (If a cluster meets both conditions, we arbitrarily throw it into one camp of the other). The first equation in \((38)\) implies that, for fixed \(i\), the values of \(p_{ij}\) are nearly constant for all \(j\) of Type II. Since the \(c_j\)’s are small for \(j\) of Type I, this common value must be close to \(d_i\). To within \(o(1)\), our equations then simplify to

\[
S'_0(d_i) = \alpha + \frac{k}{2} \beta (d_i^{k-1} + \epsilon^{k-1}), \\
S_0(d_i) - S_0(\epsilon) = \alpha(d_i - \epsilon) + \beta[d_i^{k} - \epsilon^{k} + k\epsilon^{k-1}(d_i - \epsilon)].
\]

Since \(d_M = \epsilon + o(1)\), the first of those equations applied to \(d_M\) implies that

\[
\alpha + k\epsilon^{k-1}\beta = S'_0(\epsilon) + o(1).
\]

We can thus replace \(\alpha\) with \(S'_0(\epsilon) - k\epsilon^{k-1}\beta + o(1)\) throughout. This gives the equations (again with \(o(1)\) errors):

\[
2(S'_0(d_i) - S'_0(\epsilon)) = k\beta(d_i^{k-1} - \epsilon^{k-1}), \\
2[S_0(d_i) - S_0(\epsilon) - S'_0(\epsilon)(d_i - \epsilon)] = \beta[d_i^{k} - \epsilon^{k} - k\epsilon^{k-1}(d_i - \epsilon)].
\]

There are two solutions to these equations. One is simply to have \(d_i = \epsilon\), in which case both equations say \(0 = 0\). Indeed, we already know that there must be clusters with \(d_i\) close to \(\epsilon\). In looking for solutions with \(d_i \neq \epsilon\), the second equation says that \(\beta = \psi_k(\epsilon, d_i)\).
We can also divide the first equation by the second to eliminate $\beta$. This gives an equation that is algebraically equivalent to $\partial\psi_k(\epsilon, d_i)/\partial d_i = 0$. In other words, $d_i$ must be the unique critical point $\zeta_k(\epsilon)$ of $\psi_k$, and $\beta$ must be the critical value. In fact, the critical point is a maximum of $\psi_k$. Remember that $s(\epsilon, \tau) = s(\epsilon, \epsilon^k) + \int_0^1 \beta$ from (22). Since the computation of $\beta$ is independent of $\Delta \tau$ (to lowest order), we have $s(\epsilon, \tau) - s(\epsilon, \epsilon^k) = \beta \Delta \tau + o(\Delta \tau)$, so maximizing $\beta$ is tantamount to maximizing $s$.

**Step 2.** We have shown so far that the optimizing graph is multipodal, with all of the clusters either having $d_i$ close to $\zeta_k(\epsilon)$ or close to $\epsilon$. We refine our definitions of Type I and Type II so that all the clusters with $d_i$ close to $\zeta_k(\epsilon)$ are Type I and all the clusters with $d_i$ close to $\epsilon$ are Type II. Since the value of $g(x, y)$ is determined by $d(x)$ and $d(y)$ (and $\alpha$ and $\beta$), this means that the optimizing graph is nearly constant (i.e. with pointwise small fluctuations) on each quadrant. We order the clusters so that the Type I clusters come before Type II.

Let $g_b$ be the bipodal graphon obtained by averaging over each quadrant. Let $\Delta g_f = g - g_b$. (The $f$ stands for “further”.) We will show that having $\Delta g_f$ non-zero is an inefficient way to increase $\tau$, that is, $(s(g) - s(g_b))/\Delta \tau(g) - \Delta \tau(g_b)$ is less than $\beta$. This will imply that $\Delta g_f = 0$ and so $g = g_b$.

Since $\tau = \int_0^1 d(x)^k dx$, the changes in $\tau$ are a function only of the marginal distributions of $\Delta g_f$. Once these are fixed, the values of $\Delta g_f$ on each quadrant must take the form

$$\Delta g_f(x, y) = \text{(function of } x) + \text{(function of } y).$$

(42)

The reason is that we can write the entropy on each quadrant as $\int \int S_0(g_b + \Delta g_f) = \int \int S_0(g_b) + S'_0(g_b) \Delta g_f + (1/2) S''_0(g_b) \Delta g_f^2 + \cdots$. The first term is independent of $\Delta g_f$ and the second is zero (since $g_b$ was assumed to equal the average value of $g = g_b + \Delta g_f$ on the quadrant). Since the changes to the graphon are pointwise small, we can ignore terms past the second, so we are basically left with $S''_0(g_b)/2$ times the squared $L^2$ norm of $\Delta g_f$ on the quadrant, which we then minimize subject to the constraint that the marginal distributions are fixed. We can write $\Delta g_f(x, y) = \phi_1(x) + \phi_2(y) + \phi_3(x, y)$, where $\phi_1$ and $\phi_2$ give the two fixed marginals, and $\phi_3$ has zero marginals. But then $\int \Delta g_f^2 = \int \phi_1^2 + \phi_2^2 + \phi_3^2$, since all of the cross terms integrate to zero. (Integrating $\phi_2(y)\phi_3(x, y)$ over $x$ or $\phi_1(x)\phi_3(x, y)$ over $y$ gives zero since $\phi_3$ has zero marginals, and integrating $\phi_1(x)\phi_2(y)$ over either $x$ or $y$ gives zero since $\phi_1$ and $\phi_2$ have mean zero). The way to minimize $\int \Delta g_f^2$ is simply to take $\phi_3 = 0$. This establishes (42).

Furthermore, to maximize $\tau(g) - \tau(g_b)$, the functions of $x$ should be the same (up to scale) in the I-I and I-II quadrants, and the same (up to scale) in the II-I and II-II quadrants. This is because $\tau(g) - \tau(g_b) \approx \int_1^c k(k - 1)\epsilon^{k-2} \delta d(x)^2 dx$ involves a cross term between the contributions to $\delta d(x)$ from two quadrants, and this cross term is maximized when the corresponding functions point in the same direction.

The upshot is that there are functions $F_1(x)$ on $[0, c]$ and $F_2(x)$ on $[c, 1]$, each with mean zero and normalized to have root-mean-squared 1, and constants $\mu$, $\nu$, $\kappa$, $\lambda$, such that
\[ \Delta g_f(x, y) = \mu F_1(x) + \mu F_1(y) \text{ on the I-I square.} \]
\[ \Delta g_f(x, y) = \nu F_1(x) + \kappa F_2(y) \text{ on the I-II rectangle.} \]
\[ \Delta g_f(x, y) = \lambda F_2(x) + \lambda F_2(y) \text{ on the II-II square.} \] (43)

Now we compute the changes in \( \tau \) and in \( s \), to second order in \((\mu, \nu, \kappa, \lambda)\), noting that all of the first-order changes are zero, and that the integral of \( \Delta q_f^2 \) over the I-I square, the two rectangles, and the II-II square are \( 2c^2 \mu^2 \), \( 2c(1-c)(\nu^2 + \kappa^2) \), and \( 2(1-c)^2 \lambda^2 \), respectively.

\[ s(g) - s(g_b) = \mu^2 c^2 S_0''(p_{11}) + \nu^2 c(1-c)S_0''(p_{12}) \]
\[ + \kappa^2 c(1-c)S_0''(p_{12}) + \lambda^2 (1-c)^2 S_0''(p_{22}), \]
\[ \tau(g) - \tau(g_b) = ck(k-1)d_1^{k-2}(\mu c + \nu(1-c))^2/2 + \]
\[ + (1-c)k(k-1)d_2^{k-2}(\kappa c + \lambda(1-c))^2/2. \] (44)

Both the change in \( s \) and the change in \( \tau \) are the sum of two terms, one involving \( \mu \) and \( \nu \), and the other involving \( \kappa \) and \( \lambda \). Let:

\[ A_1 = \mu^2 c^2 S_0''(p_{11}) + \nu^2 c(1-c)S_0''(p_{12}), \]
\[ A_2 = \kappa^2 c(1-c)S_0''(p_{12}) + \lambda^2 (1-c)^2 S_0''(p_{22}), \]
\[ B_1 = ck(k-1)d_1^{k-2}(\mu c + \nu(1-c))^2/2, \]
\[ B_2 = (1-c)k(k-1)d_2^{k-2}(\kappa c + \lambda(1-c))^2/2, \] (45)

so to lowest order,

\[ \frac{S(g) - S(g_b)}{\tau_k(g) - \tau_k(g_b)} = \frac{A_1 + A_2}{B_1 + B_2}, \] (46)

For the perturbations involving only \( \kappa \) and \( \lambda \), the ratio \( A_2/B_2 \) depends only on \( r = \kappa/\lambda \):

\[ \frac{A_2}{B_2} = \frac{2[r^2 cS_0''(p_{12}) + (1-c)S_0''(p_{22})]}{k(k-1)d_2^{k-2}(rc + (1-c))^2/2}. \] (47)

We optimize by taking a derivative w.r.t. \( r \) and setting it equal to zero, with the result that \( r = S_0''(p_{22})/S_0''(p_{12}) \), independent of \( c \). Since \( r \) does not diverge as \( c \to 0 \), the limit of \( A_2/B_2 \) as \( c \to 0 \) can be obtained by simply setting \( c = 0 \), giving a limiting ratio of \( 2S_0''(\epsilon)/[k(k-1)d_2^{k-2}] = \psi_k(\epsilon, \epsilon) < \beta \). Since the limit is less than \( \beta \), the ratio must be smaller than \( \beta \) for all sufficiently small values of \( c \).

Almost identical arguments apply to the perturbations involving only \( \mu \) and \( \nu \). The optimal ratio \( \mu/\nu \) is then \( S_0''(p_{12})/S_0''(p_{11}) \), which again cannot diverge as \( c \to 0 \). Thus for small values of \( c \) the dominant terms are those involving \( \nu \), and the ratio \( A_1/B_1 \) approaches \( 2S_0''(p_{12})/[k(k-1)d_1^{k-2}] \). But \( d_1 \approx p_{12} \approx \hat{\epsilon} \), so our ratio goes to \( 2S_0''(\hat{\epsilon})/[k(k-1)d_1^{k-2}] = \psi_k(\hat{\epsilon}, \hat{\epsilon}) < \beta \).

Thus there is a constant \( \beta_0 < \beta \) such that \( A_1 \leq \beta_0 B_1 \) and \( A_2 \leq \beta_0 B_2 \), so \( A_1 + A_2 < \beta_0 (B_1 + B_2) \), so

\[ s(g) - s(g_b) \leq \beta_0 (\tau(g) - \tau(g_b)). \] (48)
However \( ds/d\tau \approx \beta \) for changes in \( c \) that preserve the bipodal structure. This means if we perturb a bipodal graphon to maximize \( s \), it is better to perturb \( c \) than to make \((\mu, \nu, \kappa, \lambda)\) nonzero. Thus \( \kappa \) and \( \lambda \) must both be zero, implying that there is only one Type II cluster, and \( \mu \) and \( \nu \) must be zero, implying that there is only one Type I cluster.

**Step 3.** We have established that the minimizing graphon is bipodal, with \( p_{22} \approx \epsilon \) and \( p_{12} \approx \zeta_k(\epsilon) \). We now show that the form of this graphon is unique. Since the equation is bipodal, we consider the exact optimality equations. After eliminating \( \gamma \), we have

\[
S'_0(p_{11}) = \alpha + k\beta d_k^{k-1}, \\
S'_0(p_{12}) = \alpha + \frac{k}{2} \beta (d_1^{k-1} + d_2^{k-1}), \\
S'_0(p_{22}) = \alpha + k\beta d_k^{k-1}, \\
\frac{\partial S}{\partial c} = \frac{\partial \epsilon}{\partial c} + \beta \frac{\partial \tau}{\partial c}, \\
\tau = \tau_0. \tag{49}
\]

We use the second and third equations to solve for \( \alpha \) and \( \beta \):

\[
\alpha = -\frac{S'_0(p_{22})(d_2^{k-1} + d_1^{k-1}) + 2d_2^{k-1}S'_0(p_{12})}{d_2^{k-1} - d_1^{k-1}}, \\
\beta = \frac{2 S'_0(p_{22}) - S'_0(p_{12})}{k(d_2^{k-1} - d_1^{k-1})}. \tag{50}
\]

Plugging this into the first equation then gives

\[
S'_0(p_{11}) - 2S'_0(p_{12}) + S'_0(p_{22}) = 0. \tag{51}
\]

This leaves four equations in four unknowns, which we write as

\[
\vec{f} = \begin{pmatrix} 0 \\ 0 \\ \epsilon_0 \\ t_0 \end{pmatrix}, \tag{52}
\]

where

\[
f_1 = S'_0(p_{11}) - 2S'_0(p_{12}) + S'_0(p_{22}), \\
f_2 = \frac{\partial S}{\partial c} - \alpha \frac{\partial \epsilon}{\partial c} - \beta \frac{\partial \tau}{\partial c}, \\
f_3 = c^2 p_{11} + 2c(1-c)p_{12} + (1-c)^2 p_{22}, \\
f_4 = cd_k^1 + (1-c)d_k^2. \tag{53}
\]

and where \( \alpha \) and \( \beta \) are given by (50).

We know a solution when \( \tau_0 = \epsilon_0 \), namely \( p_{22} = \epsilon_0 \), \( p_{12} = \zeta_k(\epsilon_0) \), \( c = 0 \) and \( p_{11} = S'_0^{-1}(2S'_0[\zeta_k(\epsilon_0)] - S'_0(\epsilon_0)) \). We will show that \( df \) has non-zero determinant at this point. By the inverse function theorem, this implies that, when \( \tau_0 \) is close to \( \epsilon_0 \), there is only one value
of \((p_{11}, p_{12}, p_{22}, c)\) close to this point for which \(f(p_{11}, p_{12}, p_{22}, c) = (0, 0, \epsilon_0, \tau_0)^T\). Moreover, the parameters \((p_{11}, p_{12}, p_{22}, c)\) depend analytically on \(\epsilon_0\) and \(\tau_0\). This will complete the proof. (Note that we have reordered the variables by listing \(c\) last.)

The derivatives of \(f_1, f_3,\) and \(f_4\) are:

\[
\begin{align*}
  df_1 &= (S''_0(p_{11}), -2S''_0(p_{12}), S''_0(p_{22}), 0), \\
  df_3 &= (c^2, 2c(1-c), (1-c)^2, 2c\epsilon_{11} + 2(1-2c)p_{12} - 2(1-c)p_{22}), \\
  df_4 &= (k^2d_{12}^{-1}k c(1-c)(d_{12}^{-1} + d_{22}^{-1}), k(1-c)^2d_{22}^{-1}, \\
  &\quad d_{12}^{-1} - d_{22}^{-1} + kcd_{12}^{-1}(p_{11} - p_{12}) + k(1-c)d_{22}^{-1}(p_{12} - p_{22})).
\end{align*}
\]

(54)

Evaluating at \(c = 0\) gives

\[
\begin{align*}
  df_1 &= (S''_0(p_{11}), -2S''_0(p_{12}), S''_0(p_{22}), 0), \\
  df_3 &= (0, 0, 1, 2p_{12} - 2p_{22}), \\
  df_4 &= (0, 0, kp_{22}^{-1}p_{12} - p_{22}^{-1}, kp_{22}^{-1}(p_{12} - p_{22})).
\end{align*}
\]

(55)

d\(f\) is block triangular, with \(2 \times 2\) blocks. The lower right block has determinant \(p_{12}^k - p_{22}^k - kp_{22}^{-1}(p_{12} - p_{22}) = D(p_{22}, p_{12})\), which is non-zero when \(p_{12} \neq p_{22}\), i.e. when \(\epsilon_0 \neq (k - 1)/k\).

Also \(\partial f_2/\partial p_{11} = 0\) when \(c = 0\), since \(\alpha\) and \(\beta\) are independent of \(p_{11}\) (when \(c = 0\)) and since \(\partial^2 S/\partial c \partial p_{11}, \partial^2 \epsilon/\partial c \partial p_{11}\) and \(\partial^2 t/\partial c \partial p_{11}\) are all \(O(c)\). As a result,

\[
\det(df) = S''_0(p_{11})\partial f_2/\partial p_{12}D(p_{22}, p_{12}).
\]

(56)

So as long as \(p_{12} \neq p_{22}\) (i.e. as long as \(\epsilon_0 \neq (k - 1)/k\)), everything boils down to computing \(\partial f_2/\partial p_{12}\) at \(c = 0\) and seeing that it is nonzero. We compute

\[
\frac{\partial \beta}{\partial p_{12}} = \frac{2}{k} \frac{(p_{12}^k - p_{22}^k)(-S''_0(p_{12}) - S''_0(p_{22}) - S''_0(p_{12}))(-(k - 1)p_{12}^{k-2})}{(p_{22}^k - p_{12}^k)^2}
\]

(57)

at \(c = 0\). We will show separately that this quantity is nonzero.

Since \(\alpha = S''_0(p_{22}) - k\beta d_{22}^{-1},\)

\[
\frac{\partial \alpha}{\partial p_{12}} = -kd_{22}^{-1} \frac{\partial \beta}{\partial p_{12}} - k(k - 1)\beta d_{22}^{-2} \frac{\partial d_{12}}{\partial p_{12}} - k(k - 1)d_{22}^{-2}c\beta \Rightarrow -kd_{22}^{-1} \frac{\partial \beta}{\partial p_{12}},
\]

(58)

where \(\Rightarrow\) denotes a limit as \(c \to 0\). We also compute

\[
\frac{\partial^2 S}{\partial c \partial p_{12}} = 2(1 - 2c)S'_0(p_{12}) \Rightarrow 2S'_0(p_{12})
\]

\[
\frac{\partial^2 \epsilon}{\partial c \partial p_{12}} = 2(1 - 2c) \Rightarrow 2
\]

\[
\frac{\partial^2 t}{\partial c \partial p_{12}} = k(1 - 2c)(d_{12}^{-1} + d_{22}^{-1}) \Rightarrow k(p_{12}^{-1} + p_{22}^{-1})
\]

(59)
Finally we combine everything:

\[
\frac{\partial f_2}{\partial p_{12}}|_{c=0} = \frac{\partial^2 S}{\partial c \partial p_{12}} - \frac{\partial \alpha}{\partial p_{12}} \frac{\partial c}{\partial \epsilon} - \alpha \frac{\partial^2 \epsilon}{\partial c \partial p_{12}} - \frac{\partial \beta}{\partial p_{12}} \frac{\partial \tau}{\partial c} - \beta \frac{\partial^2 \tau}{\partial c \partial p_{12}} \\
= 2S_0'(p_{12}) - 2\alpha - \beta k(p_{12}^{k-1} + p_{22}^{k-1}) + (kp_{22}^{k-1}(2p_{12} - 2p_{22}) - (p_{12}^k - p_{22}^k + kp_{22}^{k-1}(p_{12} - p_{22}))) \frac{\partial \beta}{\partial p_{12}}. 
\]

The terms not involving \(\partial \beta/\partial p_{12}\) all cancel, by the second variational equation, and we are left with

\[
\frac{\partial f_2}{\partial p_{12}} = -D(p_{12}, p_{22}) \frac{\partial \beta}{\partial p_{12}}. 
\]

Finally, we need to show that \(\partial \beta/\partial p_{12} \neq 0\). Since \(p_{12}\) maximizes \(\psi_k(p_{22}, p_{12})\) (for fixed \(p_{22}\)), we must have (referring to the notation of the proof of Theorem 3.3) \((N/D)' = 0\), or equivalently \(N'/D' = N/D\), where we write \(\psi_k = N/D\), as above. But \(\beta = N'/D'\). If \(\partial \beta/\partial p_{12}\) were equal to zero, then we would have \(N''/D'' = N'/D'\). But we have previously shown that it is impossible to simultaneously have \(N/D = N'/D' = N''/D''\), except at \(p_{12} = p_{22} = (k - 1)/k\), so \(\partial \beta/\partial p_{12}\) must be nonzero whenever \(\epsilon_0 \neq (k - 1)/k\). This makes \(\det(d\tilde{f})\) nonzero at \((p_{11}, \zeta_k(\epsilon_0), \epsilon_0, 0)\), so the solutions near this point are unique and analytic in \((\epsilon, \tau)\).

\[\square\]

5 Theorem 1.1 for \(k\)-starlike graphs.

Now suppose that \(H\) is a \(k\)-starlike graph with \(\ell\) edges, and with \(n_k\) vertices of degree \(k\), and let \(\tau\) be the density of \(H\) and \(\tau_k\) be the density of \(k\)-stars. Our first result relates \(\Delta \tau := \tau - \epsilon^\ell\) to \(\Delta \tau_k := \tau_k - \epsilon^k\).

Lemma 5.1. If \(g\) is an entropy-maximizing graphon for \((\epsilon, \tau)\) with \(\tau > \epsilon^\ell\), then \(\Delta \tau = n_k \epsilon^{\ell-k} \Delta \tau_k + O(\Delta \tau_k^{3/2})\).

Proof. Writing \(g(x, y) = \epsilon + \Delta g(x, y)\), we expand \(\tau\) as a polynomial in \(\Delta g\):

\[
\tau = \int dx \prod g(x_i, x_j) = \int dx \prod (\epsilon + \Delta g(x_i, x_j)), 
\]

where there is a variable \(x_i\) for each vertex of \(H\) and the product is over all edges in \(H\).

The 0-th order term is \(\epsilon^\ell\). The first-order term is identically zero, since \(\int \Delta g(x, y) dx \, dy = \Delta \epsilon = 0\). When looking at higher-order expansions, there are some terms that come from having all \(\Delta g\)'s along edges that share a single vertex of degree \(k\). These terms also appear in the expansion of \(\tau_k\), so the sum of those terms is exactly \(\epsilon^{\ell-k} \Delta \tau_k\). Since all vertices have degree \(k\) or 1, summing these terms gives \(n_k \epsilon^{\ell-k} \Delta \tau_k\).

What remains are terms where the \(\Delta g\)'s refer to edges that do not all share a vertex. We bound these in turn. In each case, let \(\{e_i\}\) be the set of edges that correspond to factors of \(\Delta g\).

- If one of the \(e_i\)'s is disconnected from the rest, then the integral is exactly zero. So we can assume that all connected components of \(\{e_i\}\) contain at least two edges.

\[\]
• If there is more than one connected component, then we get a product of factors, one for each connected component. Each factor is bounded by a constant times $\|\Delta g\|^2$, so the product is $O(\|\Delta g\|^4)$.

• If there is only one connected component, whose edges do not all share a vertex, then \{e_i\} either contains a triangle or a chain of three consecutive edges. We bound such terms by taking absolute values of all the $\Delta g$’s and setting all terms other than the three edges in the triangle or 3-chain to 1. The result is either a constant times $\int\int\int\int |\Delta g(w,x)||\Delta g(x,y)||\Delta g(y,z)|dwdxdydz$, or by a constant times $\int\int\int |\Delta g(x,y)||\Delta g(y,z)||\Delta g(z,x)|dxdydz$, either of which in turn is bounded by a constant times $\|\Delta g\|^3$. (If we then think of $|\Delta g|$ as the integral kernel of an operator $L$ on $L^2(0,1)$, then the integral for a 3-chain is the expectation of $L^3$ in a particular state, and the integral for a triangle is the trace of $L^3$. Both are bounded by $Tr(L^3)/2 = \|\Delta g\|^3$.)

Since $\Delta \tau_k$ scales as $\|\Delta g\|^2$, all the corrections to the approximation $\Delta \tau \approx n_k \epsilon \Delta \tau_k$ are $O(\Delta \tau_k^3)$ or smaller.

5.1 Proof of Theorem 1.1

Since $\Delta \tau$ is proportional to $\Delta \tau_k$ (plus small errors), the problem of optimizing $\Delta s/\Delta \tau$ is a small perturbation of the problem of optimizing $\Delta s/\Delta \tau_k$, or equivalently optimizing $\Delta s$ for fixed $\Delta \tau_k$, which we solved in the last section. Since that problem has a unique optimizer, any optimizer for $\Delta s/\Delta \tau$ must come close to optimizing $\Delta s/\Delta \tau_k$, and so must be close to the bipodal graphon derived in Theorem 4.1.

We can thus write $g = g_b + \Delta g_f$, as in the last steps of the proof of Theorem 4.1, where $g_b = \epsilon + \Delta g_b$ is a bipodal graphon with $p_{22} \approx \epsilon$ and $p_{12} \approx \zeta_k(\epsilon)$ and where $\Delta g_f$ is a function that averages to zero on each quadrant of $g_b$.

Lemma 5.2. The function $\Delta g_f$ is pointwise small. That is, as $\tau \to \epsilon \ell$, $\Delta g_f$ goes to zero in sup-norm.

Proof of lemma. Since we no longer in the setting where the entropy maximizer is proven to be multipodal, we cannot use the equations (38) directly. However, we can still apply the method of Lagrange multipliers to pointwise variations of the graphon. (See [6] for a rigorous justification.) These variational equations are

$$\frac{1}{2} \ln \left( \frac{1}{g(x,y)} - 1 \right) = \frac{\delta s}{\delta g(x,y)} = \alpha + \beta \frac{\delta \tau}{\delta g(x,y)}.$$  \hfill (63)

We need to compute $\delta \tau/\delta g$ and show that it is nearly constant on each quadrant. Since $\alpha$ and $\beta$ are constants, this would imply that $g(x,y)$ is nearly constant on each quadrant, and hence that $\Delta g_f$ is pointwise small. Let $g_0(x,y) = \epsilon$.

Since $\|\Delta g\|$ is small (where $\Delta g = g - g_0 = \Delta g_b + \Delta g_f$), we can find a small constant $a = o(1)$ such that, for all $x$ outside a set $U \subset [0,1]$ of measure $a$, $\int_0^1 |\Delta g(x,y)|dy < a$. (This set $U$ is essentially what we previously called the Type I clusters, but at this stage of the
argument we are not assuming a multipodal structure. Rather, we are just using the fact that \( \tau - e^\ell = O(\|\Delta g\|^2) \).

The functional derivative \( \delta \tau / \delta g \) has a diagrammatic expansion similar to the expansion of \( \tau \). For each edge of \( H \), we get a contribution by deleting the edge, assigning the values \( x \) and \( y \) to the endpoints of the edge, and integrating over the values of all other vertices. Since \( U \) is small, we can estimate \( \delta \tau / \delta g \) to within \( o(1) \) by restricting the integral to \((U^c)^{v-2}\), where \( v \) is the number of vertices in \( H \) and \( U^c \) is the complement of \( U \). This implies that terms involving \( \Delta g \) can only contribute non-negligibly on edges connected to \( x \) or to \( y \). Furthermore, they can only contribute when attached to \( x \) if \( x \in U \), and can only contribute when attached to \( y \) if \( y \in U \).

We now begin a bootstrap. We will show that \( \delta \tau / \delta g \) is nearly constant on each quadrant \( U^c \times U^c, U \times U^c, U \times U \) in turn. This will show that \( g \) is nearly constant on that quadrant, which will help us prove that \( \delta \tau / \delta g \) is nearly constant on the next quadrant.

If \( x \) and \( y \) are both in \( U^c \), then the contributions of the terms involving \( \Delta g \) are negligible, so \( \delta \tau / \delta g(x,y) \) can be computed, to within a small error, using the approximation \( g(x,y) \approx \epsilon \). But when \( g(x,y) = \epsilon \), \( \delta \tau / \delta g(x,y) \) is independent of \( x \) and \( y \). Since \( \delta \tau / \delta g(x,y) \) is nearly constant on \( U^c \times U^c \), equation (63) implies that \( g \) is nearly constant on \( U^c \times U^c \). In particular, \( \Delta g \) is pointwise small on \( U^c \times U^c \).

Next suppose that \( y \in U^c \) and \( x \in U \). Then \( \delta \tau / \delta g(x,y) \) is nearly independent of \( y \), so \( g(x,y) \) is nearly independent of \( y \), and is nearly equal to \( d(x) \). But then the integrals involved in computing \( \delta \tau / \delta g(x,y) \) are easy, where we use \( g_0 + \Delta g \) on the edges connected to \( x \), \( g_0 \) on all other edges, and only integrate over \((U^c)^{v-2}\). If the degree of \( x \) is \( k \), then the edges connected to \( x \) contribute \( d(x)^{k-1}e^{\ell-k} \). Summing over edges, and symmetrizing over the assignment of \( x \) and \( y \) to the two endpoints, we obtain the approximation

\[
\frac{\delta \tau}{\delta g(x,y)} \approx \frac{kn_k e^{\ell-k}}{2} (d(x)^{k-1} + d(y)^{k-1}).
\]

(64)

Up to an overall factor of \( n_k e^{\ell-k} \), this is the same functional derivative as for a \( k \)-star. This also applies if \( x \in U^c \), except that in the latter case \( d(x) \approx \epsilon \), and also applies if \( x \in U^c \) and \( y \in U \).

In other words, we can use the approximation (64) in (63) whenever either \( x \) or \( y \) (or both) is in \( U^c \). This implies that the integrated equations (39) apply for all \( x \) (with \( d_i \) replaced by \( d(x) \), and with \( \beta \) scaled up by \( n_k e^{\ell-k} \)). Following the exact same reasoning as in the proof of Theorem 4.1, we obtain that \( d(x) \) only takes on 2 possible values (up to \( o(1) \) errors). We then define Type I and Type II points, depending on whether the degree function is close to \( \zeta_k(\epsilon) \) or \( \epsilon \), respectively, so that \( U \) is precisely the set of Type I points. Our graphon is then nearly constant on the I-II, II - I and II-II quadrants.

We still need to show that the graphon is nearly constant in the I-I quadrant. Suppose that \( x \) and \( y \) are in \( U \). In computing \( \delta \tau / \delta g(x,y) \), we approximate our integral by integrating over \((U^c)^{v-2}\). But if \( z \in U^c \), then \( g(x,z) \) is (nearly) independent of \( x \), since we have just established that \( g \) is nearly constant on the I-I-II quadrant. Thus \( \delta \tau / \delta g \) (which is obtained by integrating products of terms \( g(x,z) \)) is nearly independent of \( x \). Likewise, it is nearly independent of \( y \), implying that \( g(x,y) \) is nearly constant on the I-I quadrant.

Note, by the way, that the approximation (64) does not apply in the I-I quadrant; in that case \( \delta \tau / \delta g \) contains terms with powers of both \( d(x) \) and \( d(y) \). However, that approximation
is not needed for our proof, since the $I-I$ quadrant only contributes $O(c)$ to the integrated equations (39).

Returning to the proof of Theorem 1.1, we need to compare $s(g_b + \Delta g_f) - s(g_b)$ to
\[ \tau(g_b + \Delta g_f) - \tau(g_b). \]

As before, we expand $\tau(g)$ as the integral of a polynomial in $g$, obtained by assigning $g_0 + \Delta g_b + \Delta g_f$ to each edge of $H$ and integrating. The difference between $\tau(g_b + \Delta g_f)$ and $\tau(g_b)$ consists of terms with at least one $\Delta g_f$. However, the terms with exactly one $\Delta g_f$ are identically zero, since $g_b$ is constant on quadrants, and $\Delta g_f$ averages to zero on each quadrant. Furthermore, terms for which all of the $\Delta g_b$’s and $\Delta g_f$’s share a vertex are exactly what we would get from the approximation $\Delta \tau \approx n_k \epsilon^{\ell-k} \tau_k$. Any term that distinguishes between $\Delta \tau$ and $n_k \epsilon^{\ell-k} \Delta \tau_k$ must have at least two $\Delta g_f$’s and either a third $\Delta g_f$ or a $\Delta g_b$, forming either a 3-chain, a triangle, or two connected $\Delta g_f$’s and a disconnected $\Delta g_b$.

Let $\Delta g_f(x, y) = |\Delta g_f(x, y)|$, and let
\[
\Delta g_b(x, y) = \begin{cases} 2c & x, y \in \Pi, \\ 1 & \text{otherwise.} \end{cases}
\]  

This is conveniently expressed in terms of outer products. Let $|1\rangle \in L^2([0, 1])$ be the constant function 1, and let $|\omega\rangle$ be the function
\[
\omega(x) = \begin{cases} 0 & x < c, \\ 1 & x > c. \end{cases}
\]

Then
\[
\Delta g_b' = |1\rangle\langle 1| - |\omega\rangle\langle \omega| + 2c|\omega\rangle\langle \omega| - |1\rangle\langle 1| - |1 - \omega\rangle\langle 1 - \omega| + 2c|\omega\rangle\langle \omega|.
\]

Note that $|\Delta g_b(x, y)| \leq \Delta g_b'(x, y)$ for all $x, y \in (0, 1)$. To see this, the only issue is what happens when $(x, y)$ is in the $II-\Pi$ quadrant, since otherwise we trivially have $|\Delta g_b| \leq 1$. Since $e(g)$ is fixed, $(1 - c)^2$ times $\Delta g_b(x, y)$ for $x, y > c$ equals minus the integral of $\Delta g_b$ over the other three quadrants. But the area of those three quadrants is $2c - c^2 < 2c$, and the biggest possible value of $|\Delta g_b|$ is $\max(e, 1 - e) < 1$, so $\frac{1}{(1-c)^2} \int |\Delta g_b|$ (integrated over the $I-I$, $I-II$ and $II-I$ quadrants) is strictly less than $2c + O(c^2)$, and so is bounded by $2c$ for small $c$ (note that $O(c^2)$ errors are negligible).

We obtain upper bounds on the contributions of the relevant terms in the expansion of $\tau$ by replacing three $\Delta g_f(x, y)$’s and $\Delta g_b(x, y)$’s with $\Delta g_f'(x, y)$ and $\Delta g_b'(x, y)$, respectively, and replacing all other terms with 1.

Since all graphons are symmetric, hence Hermitian, their operator norms are bounded by their $L^2$ norms, so for any 3-chain
\[
\langle 1|\Delta g_1^* \Delta g_2^* \Delta g_3|1\rangle \leq \|\Delta g_1^*\|\|\Delta g_2^*\|\|\Delta g_3\|.
\]

Since $\|\Delta g_b^*\|$ and $\|\Delta g_f^*\|$ are both $o(1)$ (more precisely, $O(\sqrt{\tau - e^\ell})$), the contribution of any 3-chain is bounded by an $o(1)$ constant times $\|\Delta g_f^*\|^2$.
As for triangles, $\text{Tr}(\Delta g_f^3) \leq \|\Delta g_f^3\| = \|\Delta g_f\|^3$. Finally, we must estimate the trace of $\Delta g_f'\Delta g_f\Delta g_f'$. But this trace is

$$
(1 - \omega|\Delta g_f'\Delta g_f|1) + (\omega|\Delta g_f'\Delta g_f|1 - \omega) + 2c(\omega|\Delta g_f'\Delta g_f|\omega).
$$

Since $\|1 - \omega\| = \sqrt{c}$, the total is bounded by $(2\sqrt{c} + 2c^2)\|\Delta g_f\|^2$.

The upshot is that the ratio of $s(g_b + \Delta g_f) - s(g_b)$ and $\tau(g_b + \Delta g_f) - \tau(g_b)$ is the same as that computed for $k$-stars (up to an overall factor of $n_k\epsilon^{k-1}$), plus an $o(1)$ correction. But that ratio was bounded by a constant $\beta_0 < \beta$. Restricting attention to values of $\tau$ for which the correction is smaller than $(\beta - \beta_0)/2$, we still obtain the result that having a non-zero $\Delta g_f$ is a less efficient way of generating additional $\tau$ than simply changing $c$. Thus the optimizing graphon is exactly bipodal.

Once bipodality is established, uniqueness follows exactly as in the proof of Theorem 4.1. The difference between $\Delta \tau$ and $n_k\epsilon^{k-1}\Delta \tau_k$ is of order $c^{3/2}$, and so does not affect the linearization of the optimality equations at $c = 0$.

### 6 Linear combinations of $k$-stars

We proved Theorem 1.1 by first showing that $k$-star models have the desired behavior, and then showing that, for an arbitrary $k$-starlike graph $H$, $\Delta \tau$ is well-approximated by a multiple of $\Delta \tau_k$, so the model with densities of edges and $H$ behaves essentially the same as a model with densities of edges and $k$-stars.

To prove Theorem 1.2, we consider in this section a family of models in which we can prove bipodality and uniqueness of entropy maximizers directly, as we did for $k$-stars. In the next section, we will show how to approximate a model with an arbitrary $H$ with a model in this family.

Let $h(x) = \sum_{k \geq 1} a_k x^k$ be a polynomial with non-negative coefficients and degree $\geq 2$. Let $\tau = \sum a_k \tau_k$, and consider graphs with fixed edge density $\epsilon$ and fixed $\tau$. In [6] it was proved that the entropy-maximizing graphons in such models are always multipodal.

Most of the analysis of $k$-star models carries over to positive linear combinations, and so will only be sketched briefly. We will provide complete details where the arguments differ.

In analogy to our earlier development, let $\psi(\epsilon, \bar{\epsilon}) = N/D$, where

$$
N(\epsilon, \bar{\epsilon}) = 2[S_0(\bar{\epsilon}) - S_0(\epsilon) - (\bar{\epsilon} - \epsilon)S_0'(\epsilon)],
$$

$$
D(\epsilon, \bar{\epsilon}) = h(\bar{\epsilon}) - h(\epsilon) - (\bar{\epsilon} - \epsilon)h'(\epsilon).
$$

(70)

Since $h''(x)$ is positive for $x > 0$, $D$ is only zero when $\bar{\epsilon} = \epsilon$, and we fill in that removable singularity in $\psi$ by defining $\psi(\epsilon, \epsilon) = 2S_0'(\epsilon)/h''(\epsilon)$.

**Theorem 6.1.** For all but finitely many values of $\epsilon$, there is a $\tau_0 > h(\epsilon) > 0$ such that, for $\tau \in (h(\epsilon), \tau_0)$, the entropy-optimizing graphon is bipodal and unique, with data varying analytically with $\epsilon$ and $\tau$. As $\tau$ approaches $h(\epsilon)$ from above, $p_{22} \to \epsilon$, $p_{12}$ approaches a point $\bar{\epsilon}$ where $\psi'(\epsilon, \bar{\epsilon}) = 0$, $p_{11}$ satisfies $S_0'(p_{11}) = 2S_0'(p_{12}) - S_0'(p_{22})$ and $c \to 0$ as $O(\Delta \tau)$. 

20
Proof. For a multipodal graphon, \( \tau(g) = \sum c_i h(d_i) \). After eliminating \( \gamma \), the optimality equations become

\[
S'_0(p_{ij}) = \alpha + \beta (h'(d_i) + h'(d_j))/2, \tag{71}
\]

\[
2 \sum_{j=1}^{M} c_j (S_0(p_{ij}) - S_0(p_{Mj})) = 2\alpha (d_i - d_M) + \beta \left[ h(d_i) - h(d_M) + \sum_{j=1}^{M} c_j h'(d_j)(p_{ij} - p_{Mj}) \right]. \tag{72}
\]

As before, we distinguish between Type I clusters that are small and Type II clusters that have \( d_i \approx \epsilon \). Summing the optimality equations over \( j \) of Type II, and approximating \( d_j \) by \( \epsilon \), we obtain the equations

\[
S'_0(d_i) = \alpha + \beta (h'(d_i) + h'(\epsilon))/2, \tag{73}
\]

\[
S_0(d_i) - S_0(\epsilon) = \alpha (d_i - \epsilon) + \beta \left[ h(d_i) - h(\epsilon) + h'(\epsilon)(d_i - \epsilon) \right], \tag{74}
\]

that are accurate to within \( o(1) \). We use the first equation, with \( i = M \) (a type II cluster), to solve for \( \alpha \), and plug it into the equations for \( i < M \) to get

\[
2(S'_0(d_i) - S'_0(\epsilon)) = \beta (h'(d_i) - h'(\epsilon)), \tag{75}
\]

\[
2[S_0(d_i) - S_0(\epsilon) - S'_0(\epsilon)(d_i - \epsilon)] = \beta [h(d_i) - h(\epsilon) - h'(\epsilon)(d_i - \epsilon)], \tag{76}
\]

again to within \( o(1) \). As before in the proof of Theorem 4.1, this implies that either \( d_i \approx \epsilon \) or that \( \psi(\epsilon, d_i) \) is maximized with respect to \( d_i \).

Unlike in the \( k \)-star case, it is not true that \( \psi'(\epsilon, \bar{\epsilon}) \) has a unique solution for each \( \epsilon \). However, it remains true that \( \psi(\epsilon, \bar{\epsilon}) \) has a unique global maximizer (w.r.t. \( \bar{\epsilon} \)) for all but finitely many values of \( \epsilon \). Since the equations defining multiple maxima are analytic, they must be satisfied either for all \( \epsilon \) or for only finitely many \( \epsilon \). But it is straightforward to check that there is only one maximizer when \( \epsilon \) is sufficiently small, since then \( h(\epsilon) \) and \( h'(\epsilon) \) are dominated by the lowest order term in the polynomial.

Thus, for all but finitely many values of \( \epsilon \), the values of \( d_i \) must all either approximate \( \epsilon \) or the unique value of \( \bar{\epsilon} \) that maximizes \( \psi(\epsilon, \bar{\epsilon}) \). This allows for a re-segregation of the clusters into Type I (with \( d_i \) close to \( \bar{\epsilon} \)) and Type II (with \( d_i \) close to \( \epsilon \)) and yields a graphon that is approximately bipodal. Step 2 of the proof of Theorem 4.1, proving that the optimizing graphon is exactly bipodal with data of the desired form, then proceeds exactly as before.

What remains is showing that the optimizing graphon is unique by linearizing the exact optimality equations for bipodal graphons near \( c = 0 \). These equations are:

\[
S'_0(p_{11}) = \alpha + \beta h'(d_1),
\]

\[
S'_0(p_{12}) = \alpha + \beta (h'(d_1) + h'(d_2))/2,
\]

\[
S'_0(p_{22}) = \alpha + \beta h'(d_2),
\]

\[
\frac{\partial S}{\partial \epsilon} = \frac{\alpha}{\partial \epsilon} + \beta \frac{\partial \epsilon}{\partial \epsilon},
\]

\[
\frac{\partial c}{\partial \epsilon} = \epsilon_0, \frac{\partial c}{\partial \epsilon}, \tau = \tau_0. \tag{77}
\]

Using the second and third equations to eliminate \( \alpha \) and \( \beta \) gives:
\[
\alpha = \frac{2h'(d_2)S'_0(p_{12}) - S'_0(p_{22})(h'(d_2) + h'(d_1))}{h'(d_2) - h'(d_1)}, \]
\[
\beta = \frac{2(S'_0(p_{22}) - S'_0(p_{12}))}{h'(d_2) - h'(d_1)}. \tag{78}
\]

We also have \( \alpha = S'_0(p_{22}) - \beta h'(d_2) \) and \( S'_0(p_{11}) = 2S'_0(p_{12}) - S'_0(p_{22}) \). Note that
\[
\frac{\partial \alpha}{\partial p_{12}} = -\beta ch''(d_2) - h'(d_2) \frac{\partial \beta}{\partial p_{12}} \Rightarrow -h'(p_{22}) \frac{\partial \beta}{\partial p_{12}} \tag{79}
\]
as \( c \downarrow 0 \).

We define \( \tilde{f} \) as before, with \( f_3 = \epsilon \) and \( f_4 = \tau \), and compute
\[
df_3 = (c^2, 2c(1 - c), (1 - c)^2, 2cp_{11} + 2(1 - 2c)p_{12} - 2(1 - c)p_{22}) \]
\( \Rightarrow (0, 0, 1, 2(p_{12} - p_{22})) \),
\[
df_4 = (c^2h'(d_1), c(1 - c)(h'(d_1) + h'(d_2)), (1 - c)^2h'(d_2), \]
\( h(d_1) - h(d_2) + ch'(d_1)(p_{11} - p_{12}) + h'(d_2)(p_{12} - p_{22})) \)
\( \Rightarrow (0, 0, h'(p_{22}), h(p_{12}) - h(p_{22}) + h'(p_{22})(p_{12} - p_{22})) \). \tag{80}

The lower right block of \( \tilde{d}f \) then gives a contribution of \( h(p_{12}) - h(p_{22}) + h'(p_{22})(p_{12} - p_{22}) - 2h'(p_{22})(p_{12} - p_{22}) = h(p_{12}) - h(p_{22}) - h'(p_{22})(p_{12} - p_{22}) = D(p_{22}, p_{12}). \)

As before, \( \partial f_2/\partial p_{11} = 0 \) when \( c = 0 \), so \( \det(\tilde{f}) = S''_0(p_{11})(h(p_{11}) - h(p_{22}) - h'(p_{22})(p_{12} - p_{22})) \partial f_2/\partial p_{11} \). Now
\[
\frac{\partial f_2}{\partial p_{12}} = \frac{\partial^2 S}{\partial c \partial p_{12}} - \alpha \frac{\partial^2 \epsilon}{\partial c \partial p_{12}} - \beta \frac{\partial^2 \tau}{\partial c \partial p_{12}} - \frac{\partial \alpha}{\partial p_{12}} \frac{\partial \epsilon}{\partial c} - \frac{\partial \beta}{\partial p_{12}} \frac{\partial \tau}{\partial c}. \tag{81}
\]

Since \( \alpha \) and \( \beta \) are independent of \( c \), the first three terms are
\[
\frac{\partial}{\partial c} \left( \frac{\partial S}{\partial p_{12}} - \alpha \frac{\partial \epsilon}{\partial p_{12}} - \beta \frac{\partial \tau}{\partial p_{12}} \right) = \frac{\partial}{\partial c}(0) = 0, \tag{82}
\]
by the second variational equation. This leaves
\[
\frac{\partial f_2}{\partial p_{12}} = (h'(p_{22})(2p_{12} - p_{22}) - (h(p_{12}) - h(p_{22}) + h'(p_{22})(p_{12} - p_{22}))) \frac{\partial \beta}{\partial p_{12}}. \tag{83}
\]

Combining with our earlier results, we have:
\[
\det(\tilde{f}) = -S''_0(p_{11})D(p_{22}, p_{12})^2 \frac{\partial \beta}{\partial p_{12}}. \tag{84}
\]
The expression \( D(p_{22}, p_{12}) = h(p_{12}) - h(p_{22}) - h'(p_{22})(p_{12} - p_{22}) \) has a double root at \( p_{12} = p_{22} \) and is nonzero elsewhere, thanks to the monotonicity of \( h' \).

As a last step, we consider when \( \partial \beta / \partial p_{12} \) can be zero. Since \( \beta = N'/D' \), we are interested in when \( (N'/D')' = 0 \). But that is equivalent to having \( N''/D'' = N'/D' \). Since we already have \( N/D = N'/D' \), this means that \( \psi'' = (N/D)'' = 0 \). Since we are looking at the value of \( \epsilon \) that maximizes \( \psi \), having \( \psi' = \psi'' = 0 \) would imply \( \psi''' = 0 \) (or else \( \tilde{\epsilon} \) would only be
a point of inflection, and not a local maximum). But if \((N/D)' = (N/D)' = (N/D)'' = 0\), then \(N/D = N'/D' = N''/D'' = N'''/D'''\). Note that \(N'', N''', D''\) and \(D'''\) are functions of \(\tilde{e}\) only, and are rational functions:

\[
\begin{align*}
N'' &= 2S''_0(\tilde{e}) = \frac{-1}{\tilde{e}} - \frac{1}{1 - \tilde{e}}, \\
N''' &= 2S'''_0(\tilde{e}) = \frac{1}{\tilde{e}^2} - \frac{1}{(1 - \tilde{e})^2}, \\
D'' &= h''(\tilde{e}), \\
D''' &= h'''(\tilde{e}).
\end{align*}
\]

Setting \(D''N''' = D'''N''\) gives a polynomial equation for \(\tilde{e}\), which has only finitely many roots. Since the equation \(\psi' = 0\) is symmetric in \(\epsilon\) and \(\tilde{e}\), \(\tilde{e}\) determines \(\epsilon\), so there are only finitely many values of \(\epsilon\) for which \(\partial \beta / \partial p_{12}\) is zero.

In summary, we exclude the finitely many values of \(\epsilon\) for which \(\psi\) achieves its maximum more than once, and the finitely many values of \(\epsilon\) for which \(\partial \beta / \partial p_{12} = 0\). For all other values of \(\epsilon\), the optimizing graphon is bipodal of the prescribed form and unique.

\[\square\]

7 Proof of Theorem 1.2

The proof has three steps.

Step 1. Showing that, for fixed \(\epsilon\), \(\Delta \tau\) can be approximated by the change in a positive linear combination of \(\tau_k\)'s, as studied in the last section.

Step 2. Defining a set \(B_H \subset (0,1)\) of “bad values”, determined by analytic equations, such that for all \(\epsilon \not\in B_H\) and for \(\tau\) close enough to \(\epsilon^\ell\), the optimizing graphon is unique and bipodal and of the desired form.

Step 3. Showing that \(B_H\) is finite.

**Step 1.** This is a repetition of the proof of Lemma 5.1. In the expansion of \(\Delta \tau\), we get a contribution \(n_k \epsilon^{\ell-k} \Delta \tau_k\) from diagrams where all the edges associated with \(\Delta g\) are connected to a vertex of degree \(k\), where \(n_k\) is the number of vertices of \(H\) of degree \(k\). Summing over \(k\), and bounding the remaining terms by \(O(\|\Delta g\|^3)\), as before, we have

\[
\Delta \tau = \sum_k n_k \epsilon^{\ell-k} \Delta \tau_k + O(\Delta \tau^{3/2}).
\]  

**Step 2.** For fixed \(\epsilon\), we consider a model whose density is \(\sum_k n_k \epsilon^{\ell-k} \tau_k\). As long as \(\psi(\epsilon, \tilde{e})\) for this model achieves its maximum at a unique value of \(\tilde{e}\), and as long as \(\partial \beta / \partial p_{12} \neq 0\) when \(p_{12}\) equals this value of \(\tilde{e}\), the proofs of Theorems 1.1 and 6.1 carry over almost verbatim.

That is, the model problem has a unique bipodal maximizer by the reasoning of Theorem 6.1. The entropy maximizer for the actual problem involving \(H\) must approximate the entropy maximizer for the model problem, and in particular must be approximately bipodal,
and so can be written as $g_h + \Delta g_f$, where $\Delta g_f$ averages to zero on each quadrant. The same arguments as in the proof of Theorem 1.1 show that $\Delta g_f$ is pointwise small. By a power series expansion, $(s(g_h + \Delta g_f) - s(g_h)) / (\tau(g_h + \Delta g_f) - \tau(g_h)) < \beta$, so for small $c$ we can increase the entropy by setting $\Delta g_f$ to zero and varying the bipodal data to achieve the correct value of $\tau$.

### Step 3.

For any fixed $\epsilon$, the model problem has only a finite number of bad values of $\epsilon$, but this is not enough to prove that $B_H$ is finite. Rather

$$B_H = \{ \epsilon | \epsilon \text{ is one of the bad points for the model with } a_k = n_k \epsilon^{\ell - k} \}, \quad (87)$$

where a value of $\epsilon$ is bad for a model if either $\psi$ has multiple maxima or if $\partial \beta / \partial p_{12} = 0$. Since the bad points for any linear combination of $k$-stars depends analytically on the coefficients of that linear combination, and since these coefficients are powers of $\epsilon$, the set $B_H$ is cut out by analytic equations in $\epsilon$.

As such, $B_H$ is either the entire interval $(0,1)$, or a finite set, or a countable set with limit points only at $0$ and/or $1$. We will show that neither $0$ nor $1$ is a limit point of $B_H$, implying that $B_H$ is finite.

Let $k_{\text{max}}$ be the largest degree of any vertex in $H$, and consider the model problem with $h(x) = \sum_{k=2}^{k_{\text{max}}} a_k x^k$, where $a_k = n_k \epsilon^{\ell - k}$. We begin with some constraints on the values of $\tilde{\epsilon}$ for which $\psi'(\tilde{\epsilon}) = 0$.

**Lemma 7.1.** Suppose that $\psi'(\epsilon, \tilde{\epsilon}) = 0$. If $\tilde{\epsilon} = \epsilon$, or if $\partial \beta / \partial p_{12} = 0$ when $p_{22} = \epsilon$ and $p_{12} = \tilde{\epsilon}$, then $(1/2) \leq \tilde{\epsilon} \leq (k_{\text{max}} - 1)/k_{\text{max}}$.

**Proof of lemma.** In both cases we are looking for solutions to $N''D''' = N'''D''$. Since $N'' = 2S_0''(\tilde{\epsilon})$, $N''' = 2S_0''''(\tilde{\epsilon})$, $D'' = h''(\tilde{\epsilon})$ and $D''' = h''''(\tilde{\epsilon})$, this equation does not involve $\epsilon$ (except insofar as the coefficients of $h$ depend on $\epsilon$). We have

\[
\begin{align*}
\frac{2S_0''''(\tilde{\epsilon})}{2S_0''(\tilde{\epsilon})} &= \frac{h''''(\tilde{\epsilon})}{h''(\tilde{\epsilon})}, \\
\frac{1}{1 - \tilde{\epsilon} - \tilde{\epsilon}^2} &= \frac{1 - \tilde{\epsilon} + \tilde{\epsilon}^2}{h''(\tilde{\epsilon})}, \\
\frac{2\tilde{\epsilon} - 1}{1 - \tilde{\epsilon}} &= \frac{\tilde{\epsilon} h'''(\tilde{\epsilon})}{h''(\tilde{\epsilon})}, \\
\frac{1}{1 - \tilde{\epsilon} - 2} &= \frac{\sum k(k - 1)(k - 2)a_k \tilde{\epsilon}^{k - 2}}{\sum k(k - 1)a_k \tilde{\epsilon}^{k - 2}}. \quad (88)
\end{align*}
\]

The right hand side of the last line is a weighted average of $k - 2$ with weights $k(k - 1)a_k \tilde{\epsilon}^{k - 2}$, and so is at least zero and at most $k_{\text{max}} - 2$. Thus $(1 - \tilde{\epsilon})^{-1}$ is between $2$ and $k_{\text{max}}$ and $\tilde{\epsilon}$ is between $1/2$ and $(k_{\text{max}} - 1)/k_{\text{max}}$. \hfill \Box

**Lemma 7.2.** If $\psi'(\epsilon, \tilde{\epsilon}) = 0$, and if $\epsilon$ is sufficiently close to $1$, then $\tilde{\epsilon}$ is uniquely defined and approaches $0$ as $\epsilon \to 1$. Likewise, if $\epsilon$ is sufficiently close to $0$, then $\tilde{\epsilon}$ is uniquely defined and approaches $1$ as $\epsilon \to 0$. 

24
Proof. When \( \epsilon < 1/2, \) or when \( \epsilon > (k_{\text{max}} - 1)/k_{\text{max}}, \) we cannot have \( \tilde{\epsilon} = \epsilon, \) so the equation \( \psi' = 0 \) is equivalent to \( ND' = D\psi' \) and \( \tilde{\epsilon} \neq \epsilon. \) Writing \( D\psi' - ND' = 0 \) explicitly, and doing some simple algebra, yields the equation

\[
S'_0(\epsilon)[h(\tilde{\epsilon}) - h(\epsilon) - (\tilde{\epsilon} - \epsilon)h'(\epsilon)] - S'(\tilde{\epsilon})[[h(\tilde{\epsilon}) - h(\epsilon) - (\tilde{\epsilon} - \epsilon)h'(\tilde{\epsilon})] + (S_0(\tilde{\epsilon}) - S_0(\epsilon))(h'(\tilde{\epsilon}) - h'(\epsilon))] = 0.
\]  

(89)

If \( \epsilon \) approaches 0 or 1 and \( \tilde{\epsilon} \) does not, then the first term diverges, while the other terms do not, insofar as \( S'_0 \) has singularities at 0 and 1 but \( S_0, h \) and \( h' \) do not. Thus \( \tilde{\epsilon} \) must go to 0 or 1 as \( \epsilon \) goes to 0 or 1.

We next rule out the possibility that both \( \epsilon \) and \( \tilde{\epsilon} \) approach 1. Suppose that \( \epsilon \) is close to 1. We expand both \( N \) and \( D \) in powers of \( (\tilde{\epsilon} - \epsilon) \):

\[
N = \sum_{m=2}^{\infty} \frac{2S_0^{(m)}(\epsilon)}{m!} (\tilde{\epsilon} - \epsilon)^m
\]

\[
= - \sum_{m=2}^{\infty} \left( \frac{1}{(1-\epsilon)^{m-1}} + \frac{(-1)^m}{\epsilon^{m-1}} \right) \frac{(\tilde{\epsilon} - \epsilon)}{m(m-1)},
\]

\[
D = \sum_{m=2}^{k_{\text{max}}} \frac{h^{(m)}(\epsilon)}{m!} (\tilde{\epsilon} - \epsilon)^m,
\]

(90)

where \( S_0^{(m)} \) and \( h^{(m)} \) denote \( m \)th derivatives. Note that the coefficients of the numerator grow rapidly with \( m \), while the growth of the coefficients of the denominator depend only on the degree of \( h \). For \( \tilde{\epsilon} > \epsilon > (k_{\text{max}} - 1)/k_{\text{max}}, \) \( \psi = N/D \) is a decreasing function of \( \tilde{\epsilon} \) (that is, negative and increasing in magnitude), so we cannot have \( \psi' = 0 \). Since the equation \( \psi' = 0 \) is symmetric in \( \epsilon \) and \( \tilde{\epsilon} \) (apart from the dependence of the coefficients of \( h \) on \( \epsilon \)), we also cannot have \( \epsilon > \tilde{\epsilon} > (k_{\text{max}} - 1)/k_{\text{max}}. \)

When \( \epsilon \) is close to 1, we must thus have \( \tilde{\epsilon} \) close to 0. But then \( N \approx 2S'_0(\epsilon), \ D \approx h'(\epsilon) - h(\epsilon), \ D' \approx -h'(E), \) and the equation

\[
2S'_0(\epsilon) = N' + 2S'_0(\epsilon) = 2S'_0(\epsilon) + ND'/D
\]

(91)

determines \( S'_0(\tilde{\epsilon}), \) and therefore \( \tilde{\epsilon}, \) uniquely as a function of \( \epsilon. \)

Next we consider \( \epsilon \to 0. \) If \( H \) is 2-starlike, then \( \psi \) is a multiple of \( \psi_2, \) and the result is already known. Otherwise, it is convenient to define a new polynomial \( \tilde{h}(z) = \sum n_k z^k, \) so that \( h(x) = \tilde{h}(x/\epsilon). \) Then

\[
D = h(\tilde{x}) - h(\epsilon) - h'(\epsilon)(\tilde{\epsilon} - \epsilon) = \epsilon^2[h(\epsilon) - \tilde{h}(1) - \tilde{h}'(1)(r - 1)]
\]

(92)

where \( r := \tilde{\epsilon}/\epsilon. \) Likewise,

\[
N = -[\tilde{\epsilon} \ln(\epsilon) - \epsilon \ln(\epsilon) + (1 - \tilde{\epsilon}) \ln(1 - \epsilon) - (1 - \epsilon)(1 - \tilde{\epsilon}) - (\tilde{\epsilon} - \epsilon)(\ln(\epsilon) - \ln(1 - \epsilon))]
\]

(93)

Since \( \epsilon \) and \( \tilde{\epsilon} \) are small, we can approximate \( \ln(1 - \epsilon) \) and \( \ln(1 - \tilde{\epsilon}) \) as \( -\epsilon \) and \( -\tilde{\epsilon}, \) respectively, giving

\[
N \approx -\epsilon[r \ln r - r + 1] + \epsilon^2(r - r^2)
\]

(94)
Note that the ratio $\psi = N/D$ is negative. Since $\bar{h}$ is a polynomial of degree at least 3, $D$ grows faster than $N$ as $r \to \infty$, so we can always increase $\psi$ by taking larger and larger values of $r = \bar{\epsilon}/\epsilon$. This argument only breaks down when the approximation $\ln(1 - \tilde{E}) \approx -\bar{\epsilon}$ breaks down, i.e. at values of $\bar{\epsilon}$ that are no longer close to 0. Thus we cannot have $\bar{\epsilon}$ and $\epsilon$ both close to zero.

Finally, if $\epsilon$ is close to 0 and $\bar{\epsilon}$ is close to 1, then $h(\epsilon)$ and $h'(\epsilon)$ are close to zero, while $h(\bar{\epsilon})$ is close to a multiple of $x^{k_{\text{max}}}$, since the coefficient of $x^{k_{\text{max}}}$ is $O(1/\epsilon)$ larger than any other coefficient. Thus $\psi$ behaves like $\psi_{k_{\text{max}}}$, and has a unique maximizer.

We have shown that when $\epsilon$ is close to 0 or 1, $\psi$ has a unique maximizer. Furthermore, $\bar{\epsilon}$ is not between $1/2$ and $(k_{\text{max}} - 1)/k_{\text{max}}$, so $\partial \beta / \partial p_{12} \neq 0$. So $\epsilon \not\in B_H$, completing Step 3 and the proof of Theorem 1.2.

8 Conclusions

We have shown that just above the ER curve, entropy maximizing graphons, constrained by the densities of edges and any one other subgraph $H$, exhibit the same qualitative behavior for all $H$ and for (almost) all values of $\epsilon$. The optimizing graphon is unique and bipodal.

These results were proven by perturbation theory, using the fact that the optimizing graphon has to be $L^2$-close to a constant (Erdős-Rényi) graphon. Surprisingly, the optimizing graphon is not pointwise close to constant. Rather, it is bipodal, with a small cluster of size $O(\Delta \tau)$. As $\Delta \tau$ approaches 0, the size of the small cluster shrinks, but the values of the graphon on each quadrant do not approach one another. Rather, $p_{22}$ approaches $\epsilon$, $p_{12}$ approaches the value of $\bar{\epsilon}$ that maximizes a specific function $\psi(\epsilon, \bar{\epsilon})$, and $p_{11}$ satisfies $S'_0(p_{11}) - 2S'_0(p_{12}) + S'_0(p_{22}) = 0$.

Finally, the asymptotic behavior of these graphons as $\tau \to \epsilon^\ell$ depends only on the degree sequence of $H$. In particular, the cases where $H$ is a triangle and when $H$ is a 2-star are asymptotically the same. This is illustrated in Figure 2. Since $\Delta \tau_{\text{triangle}} \approx 3\epsilon \Delta \tau_2$, the optimizing graphon for the 2-star model with $\epsilon = 0.4$ and $\Delta \tau_2 = 0.002$ should resemble the optimizing graphon for the triangle model with $\epsilon = 0.4$ and $\Delta \tau_{\text{triangle}} = 0.0024$. These optimizing graphons are obtained using the algorithms we developed in [12] without assuming bipodality. Numerical estimates indicate that the optimizing graphons are not exactly the same, thanks to $O(\Delta \tau_{2}^{3/2})$ corrections to $\Delta \tau_{\text{triangle}}$, but are still qualitatively similar.

Acknowledgments

The computational results shown in this work were obtained on the computational facilities in the Texas Super Computing Center (TACC). We gratefully acknowledge this computational support. R. Kenyon was partially supported by the Simons Foundation grant 327929. This work was also partially supported by NSF grants DMS-1208191, DMS-1509088, DMS-1321018 and DMS-1101326.
Figure 2: Numerical estimates of the optimizing graphon for the 2-star model with $\epsilon = 0.4$ and $\tau_2 = 0.1620$ (left) and the optimizing graphon for the triangle model with $\epsilon = 0.4$ and $\tau_{\text{triangle}} = 0.0664$ (right). (Although theoretically we have not tried to prove that $\Delta \tau_2 = 0.002$ is small enough to fit into the interval provided by Theorem 1.1, numerically it appears to be the case.)

References


