

# Eigenvalues of Non-Regular Linear Quasirandom Hypergraphs

John Lenz \*

University of Illinois at Chicago  
lenz@math.uic.edu

Dhruv Mubayi †

University of Illinois at Chicago  
mubayi@math.uic.edu

September 5, 2013

## Abstract

In the paper “Eigenvalues and Linear Quasirandom Hypergraphs”, the authors defined a spectral quasirandom property for  $k$ -uniform hypergraphs which extends the well-known graph property of the separation of the first and second largest eigenvalues of the adjacency matrix of a graph. The authors proved this spectral property is equivalent to several other hypergraph quasirandom properties, but to simplify the presentation only proved the equivalence for so-called  $k$ -uniform, coregular hypergraphs with loops. This paper extends the results of “Eigenvalues and Linear Quasirandom Hypergraphs” by proving this equivalence for all  $k$ -uniform hypergraphs, not just the coregular ones.

## 1 Introduction

The study of quasirandom or pseudorandom graphs was initiated by Thomason [18, 19] and then refined by Chung, Graham, and Wilson [7], resulting in a list of equivalent (deterministic) properties of graph sequences which are inspired by  $G(n, p)$ . Almost immediately after proving their graph theorem, Chung and Graham [2, 3, 4, 5, 6] began investigating a  $k$ -uniform hypergraph generalization. Since then, many authors have studied hypergraph quasirandomness [1, 8, 9, 12, 13, 14, 15, 16, 17].

One important  $k$ -uniform hypergraph quasirandom property is **Disc**, which states that all sufficiently large vertex sets have the same edge density as the entire hypergraph. Kohayakawa, Nagle, Rödl, and Schacht [14] and Conlon, Hàn, Person, and Schacht [8] studied **Disc** and found several properties equivalent to it, but were not able to find a generalization of a graph property called **Eig**. In graphs, **Eig** states that the first and second largest (in

---

\*Research partly supported by NSA Grant H98230-13-1-0224.

†Research supported in part by NSF Grants 0969092 and 1300138.

absolute value) eigenvalues of the adjacency matrix are separated. The authors [16] answered this question by defining a property **Eig** for  $k$ -uniform hypergraphs and showed that it is equivalent to **Disc**, but only proved this for so-called coregular sequences. In this paper, we build up the additional algebra required to prove this equivalence for all  $k$ -uniform hypergraph sequences, not just the coregular ones. Before stating our result, we need some definitions.

Let  $k \geq 2$  be an integer and let  $\pi$  be a proper partition of  $k$ , by which we mean that  $\pi$  is an unordered list of at least two positive integers whose sum is  $k$ . For the partition  $\pi$  of  $k$  given by  $k = k_1 + \dots + k_t$ , we will abuse notation by saying that  $\pi = k_1 + \dots + k_t$ . A  $k$ -uniform hypergraph with loops  $H$  consists of a finite set  $V(H)$  and a collection  $E(H)$  of  $k$ -element multisets of elements from  $V(H)$ . Informally, every edge has size exactly  $k$  but a vertex is allowed to be repeated inside of an edge. If  $F$  and  $G$  are  $k$ -uniform hypergraphs with loops, a *labeled copy of  $F$  in  $H$*  is an edge-preserving injection  $V(F) \rightarrow V(H)$ , i.e. an injection  $\alpha : V(F) \rightarrow V(H)$  such that if  $E$  is an edge of  $F$ , then  $\{\alpha(x) : x \in E\}$  is an edge of  $H$ . The following is our main theorem.

**Theorem 1.** *Let  $0 < p < 1$  be a fixed constant and let  $\mathcal{H} = \{H_n\}_{n \rightarrow \infty}$  be a sequence of  $k$ -uniform hypergraphs with loops such that  $|V(H_n)| = n$  and  $|E(H_n)| \geq p \binom{n}{k}$ . Let  $\pi = k_1 + \dots + k_t$  be a proper partition of  $k$  and let  $\ell \geq 2$ . Assume that  $\mathcal{H}$  satisfies the property*

- **Cycle $_{4\ell}[\pi]$** : the number of labeled copies of  $C_{\pi, 4\ell}$  in  $H_n$  is at most  $p^{|E(C_{\pi, 4\ell})|} n^{|V(C_{\pi, 4\ell})|} + o(n^{|V(C_{\pi, 4\ell})|})$ , where  $C_{\pi, 4\ell}$  is the hypergraph cycle of type  $\pi$  and length  $4\ell$  defined in [16, Section 2].

Then  $\mathcal{H}$  satisfies the property

- **Eig $[\pi]$** :  $\lambda_{1,\pi}(H_n) = pn^{k/2} + o(n^{k/2})$  and  $\lambda_{2,\pi}(H_n) = o(n^{k/2})$ , where  $\lambda_{1,\pi}(H_n)$  and  $\lambda_{2,\pi}(H_n)$  are the first and second largest eigenvalues of  $H_n$  with respect to  $\pi$ , defined in Section 2.

When Theorem 1 is combined with [16], we obtain the following theorem.

**Theorem 2.** *Let  $0 < p < 1$  be a fixed constant and let  $\mathcal{H} = \{H_n\}_{n \rightarrow \infty}$  be a sequence of  $k$ -uniform hypergraphs with loops such that  $|V(H_n)| = n$  and  $|E(H_n)| \geq p \binom{n}{k} + o(n^k)$ . Let  $\pi = k_1 + \dots + k_t$  be a proper partition of  $k$ . The following properties are equivalent:*

- **Eig $[\pi]$** :  $\lambda_{1,\pi}(H_n) = pn^{k/2} + o(n^{k/2})$  and  $\lambda_{2,\pi}(H_n) = o(n^{k/2})$ , where  $\lambda_{1,\pi}(H_n)$  and  $\lambda_{2,\pi}(H_n)$  are the first and second largest eigenvalues of  $H_n$  with respect to  $\pi$ , defined in Section 2.
- **Expand $[\pi]$** : For all  $S_i \subseteq \binom{V(H_n)}{k_i}$  where  $1 \leq i \leq t$ ,

$$e(S_1, \dots, S_t) = p \prod_{i=1}^t |S_i| + o(n^k)$$

where  $e(S_1, \dots, S_t)$  is the number of tuples  $(s_1, \dots, s_t)$  such that  $s_1 \cup \dots \cup s_t$  is a hyperedge and  $s_i \in S_i$ .

- **Count** $[\pi$ -linear]: If  $F$  is an  $f$ -vertex,  $m$ -edge,  $k$ -uniform,  $\pi$ -linear hypergraph, then the number of labeled copies of  $F$  in  $H_n$  is  $p^m n^f + o(n^f)$ . The definition of  $\pi$ -linear appears in [16, Section 1].
- **Cycle** $_4[\pi]$ : The number of labeled copies of  $C_{\pi,4}$  in  $H_n$  is at most  $p^{|E(C_{\pi,4})|} n^{|V(C_{\pi,4})|} + o(n^{|V(C_{\pi,4})|})$ , where  $C_{\pi,4}$  is the hypergraph four cycle of type  $\pi$  which is defined in [16, Section 2].
- **Cycle** $_{4\ell}[\pi]$ : the number of labeled copies of  $C_{\pi,4\ell}$  in  $H_n$  is at most  $p^{|E(C_{\pi,4\ell})|} n^{|V(C_{\pi,4\ell})|} + o(n^{|V(C_{\pi,4\ell})|})$ , where  $C_{\pi,4\ell}$  is the hypergraph cycle of type  $\pi$  and length  $4\ell$  defined in [16, Section 2].

The remainder of this paper is organized as follows. Section 2 contains the definitions we will require from [16] and also an overview of the proof of Theorem 1. Section 3 contains the algebraic properties required for the proof of Theorem 1, and finally Section 4 contains the proof of Theorem 1.

## 2 Definitions and Overview

First, let us recall the proof of  $\text{Cycle}_4[1+1] \Rightarrow \text{Eig}[1+1]$  for regular graphs. If  $A$  is the adjacency matrix of a graph  $G$ , then  $A^4$  counts walks of length 4 in the sense that the  $(i, j)$ -th entry of  $A^4$  is the number of walks of length 4 between  $i$  and  $j$ . The trace of  $A^4$  is then the number of circuits of length 4 in  $G$ . The trace of a square real symmetric matrix is the sum of its eigenvalues. If  $G$  is  $d$ -regular, then the largest eigenvalue of  $A^4$  is  $d^4$  so if the number of circuits of length four in  $G$  is  $d^4 + o(n^4)$ , then the trace of  $A^4$  is  $d^4 + o(n^4)$  which implies that all the eigenvalues of  $A$  besides  $d$  are  $o(n)$ .

For non-regular graphs having density  $p$ , Chung, Graham, and Wilson [7] proved that in a graph sequence satisfying  $\text{Eig}[1+1]$ , the distance between the all-ones vector and the eigenvector corresponding to the largest eigenvalue is  $o(1)$  (see the bottom of page 350 in [7]). The reason for this is that if  $A$  is the adjacency matrix and  $v$  the unit length eigenvector corresponding to the largest eigenvalue, then the second largest eigenvalue of  $A$  is the spectral norm of  $A - \lambda_1 v v^T$ . But as the proof of the (non-regular) Expander Mixing Lemma shows,  $\text{Expand}[1+1]$  is related to a bound on the spectral norm of  $A - p n \hat{1} \hat{1}^T$ , where  $\hat{1}$  is the all-ones vector scaled to unit length (note  $n \hat{1} \hat{1}^T = J$ , the all-ones matrix). Indeed, if  $S, T \subseteq V(G)$  and  $\chi_S$  and  $\chi_T$  are the indicator vectors for  $S$  and  $T$  respectively, then  $e(S, T) - p|S||T|$  is exactly  $\chi_S^T (A - p n \hat{1} \hat{1}^T) \chi_T$ . Chung, Graham, and Wilson [7] proved that  $\|v - \hat{1}\| = o(1)$  to conclude that  $A - \lambda_1 v v^T$  and  $A - p n \hat{1} \hat{1}^T$  are almost the same matrix so their spectral norms are asymptotically equal, so a bound on  $\lambda_2(A)$  also bounds  $\|A - p n \hat{1} \hat{1}^T\|$ . Proposition 3 extends this to hypergraphs, and is our main result. Before stating this proposition, we recall several definitions from [16].

**Definition. (Friedman and Wigderson [10, 11])** Let  $H$  be a  $k$ -uniform hypergraph with loops. The *adjacency map* of  $H$  is the symmetric  $k$ -linear map  $\tau_H : W^k \rightarrow \mathbb{R}$  defined as

follows, where  $W$  is the vector space over  $\mathbb{R}$  of dimension  $|V(H)|$ . First, for all  $v_1, \dots, v_k \in V(H)$ , let

$$\tau_H(e_{v_1}, \dots, e_{v_k}) = \begin{cases} 1 & \{v_1, \dots, v_k\} \in E(H), \\ 0 & \text{otherwise,} \end{cases}$$

where  $e_v$  denotes the indicator vector of the vertex  $v$ , that is the vector which has a one in coordinate  $v$  and zero in all other coordinates. We have defined the value of  $\tau_H$  when the inputs are standard basis vectors of  $W$ . Extend  $\tau_H$  to all the domain linearly.

**Definition.** Let  $W$  be a finite dimensional vector space over  $\mathbb{R}$ , let  $\sigma : W^k \rightarrow \mathbb{R}$  be any  $k$ -linear function, and let  $\vec{\pi}$  be a proper ordered partition of  $k$ , so  $\vec{\pi} = (k_1, \dots, k_t)$  for some integers  $k_1, \dots, k_t$  with  $t \geq 2$ . Now define a  $t$ -linear function  $\sigma_{\vec{\pi}} : W^{\otimes k_1} \times \dots \times W^{\otimes k_t} \rightarrow \mathbb{R}$  by first defining  $\sigma_{\vec{\pi}}$  when the inputs are basis vectors of  $W^{\otimes k_i}$  and then extending linearly. For each  $i$ ,  $B_i = \{b_{i,1} \otimes \dots \otimes b_{i,k_i} : b_{i,j} \text{ is a standard basis vector of } W\}$  is a basis of  $W^{\otimes k_i}$ , so for each  $i$ , pick  $b_{i,1} \otimes \dots \otimes b_{i,k_i} \in B_i$  and define

$$\sigma_{\vec{\pi}}(b_{1,1} \otimes \dots \otimes b_{1,k_1}, \dots, b_{t,1} \otimes \dots \otimes b_{t,k_t}) = \sigma(b_{1,1}, \dots, b_{1,k_1}, \dots, b_{t,1}, \dots, b_{t,k_t}).$$

Now extend  $\sigma_{\vec{\pi}}$  linearly to all of the domain.  $\sigma_{\vec{\pi}}$  will be  $t$ -linear since  $\sigma$  is  $k$ -linear.

**Definition.** Let  $W_1, \dots, W_k$  be finite dimensional vector spaces over  $\mathbb{R}$ , let  $\|\cdot\|$  denote the Euclidean 2-norm on  $W_i$ , and let  $\phi : W_1 \times \dots \times W_k \rightarrow \mathbb{R}$  be a  $k$ -linear map. The *spectral norm of  $\phi$*  is

$$\|\phi\| = \sup_{\substack{x_i \in W_i \\ \|x_i\|=1}} |\phi(x_1, \dots, x_k)|.$$

**Definition.** Let  $H$  be a  $k$ -uniform hypergraph with loops and let  $\tau = \tau_H$  be the ( $k$ -linear) adjacency map of  $H$ . Let  $\pi$  be any (unordered) partition of  $k$  and let  $\vec{\pi}$  be any ordering of  $\pi$ . The *largest and second largest eigenvalues of  $H$  with respect to  $\pi$* , denoted  $\lambda_{1,\pi}(H)$  and  $\lambda_{2,\pi}(H)$ , are defined as

$$\lambda_{1,\pi}(H) := \|\tau_{\vec{\pi}}\| \quad \text{and} \quad \lambda_{2,\pi}(H) := \left\| \tau_{\vec{\pi}} - \frac{k!|E(H)|}{n^k} J_{\vec{\pi}} \right\|.$$

**Definition.** Let  $V_1, \dots, V_t$  be finite dimensional vector spaces over  $\mathbb{R}$  and let  $\phi, \psi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be  $t$ -linear maps. The *product* of  $\phi$  and  $\psi$ , written  $\phi * \psi$ , is a  $(t-1)$ -linear map defined as follows. Let  $u_1, \dots, u_{t-1}$  be vectors where  $u_i \in V_i$ . Let  $\{b_1, \dots, b_{\dim(V_i)}\}$  be any orthonormal basis of  $V_t$ .

$$\begin{aligned} \phi * \psi &: (V_1 \otimes V_1) \times (V_2 \otimes V_2) \times \dots \times (V_{t-1} \otimes V_{t-1}) \rightarrow \mathbb{R} \\ \phi * \psi(u_1 \otimes v_1, \dots, u_{t-1} \otimes v_{t-1}) &:= \sum_{j=1}^{\dim(V_t)} \phi(u_1, \dots, u_{t-1}, b_j) \psi(v_1, \dots, v_{t-1}, b_j) \end{aligned}$$

Extend the map  $\phi * \psi$  linearly to all of the domain to produce a  $(t-1)$ -linear map.

Lemma 5 shows that the maps are well defined: the map is the same for any choice of orthonormal basis by the linearity of  $\phi$  and  $\psi$ .

**Definition.** Let  $V_1, \dots, V_t$  be finite dimensional vector spaces over  $\mathbb{R}$  and let  $\phi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map and let  $s$  be an integer  $0 \leq s \leq t - 1$ . Define

$$\phi^{2^s} : V_1^{\otimes 2^s} \times \dots \times V_{t-s}^{\otimes 2^s} \rightarrow \mathbb{R}$$

where  $\phi^{2^0} := \phi$  and  $\phi^{2^s} := \phi^{2^{s-1}} * \phi^{2^{s-1}}$ .

Note that we only define this for exponents which are powers of two because the product  $*$  is only defined when the domains of the maps are the same. An expression like  $\phi^3 = \phi * (\phi * \phi)$  does not make sense because  $\phi$  and  $\phi * \phi$  have different domains. This defines the power  $\phi^{2^{t-1}}$ , which is a linear map  $V_1^{\otimes 2^{t-1}} \rightarrow \mathbb{R}$ .

**Definition.** Let  $V_1, \dots, V_t$  be finite dimensional vector spaces over  $\mathbb{R}$  and let  $\phi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map and define  $A[\phi^{2^{t-1}}]$  to be the following square matrix/bilinear map. Let  $u_1, \dots, u_{2^{t-2}}, v_1, \dots, v_{2^{t-2}}$  be vectors where  $u_i, v_i \in V_1$ .

$$A[\phi^{2^{t-1}}] : V_1^{\otimes 2^{t-2}} \times V_1^{\otimes 2^{t-2}} \rightarrow \mathbb{R}$$

$$A[\phi^{2^{t-1}}](u_1 \otimes \dots \otimes u_{2^{t-2}}, v_1 \otimes \dots \otimes v_{2^{t-2}}) := \phi^{2^{t-1}}(u_1 \otimes v_1 \otimes u_2 \otimes v_2 \otimes \dots \otimes u_{2^{t-2}} \otimes v_{2^{t-2}}).$$

Extend the map linearly to the entire domain to produce a bilinear map.

Lemma 7 below proves that  $A[\phi^{2^{t-1}}]$  is a square symmetric real valued matrix. The following is the main result of this note.

**Proposition 3.** *Let  $\{\psi_r\}_{r \rightarrow \infty}$  be a sequence of symmetric  $k$ -linear maps, where  $\psi_r : V_r^k \rightarrow \mathbb{R}$ ,  $V_r$  is a vector space over  $\mathbb{R}$  of finite dimension, and  $\dim(V_r) \rightarrow \infty$  as  $r \rightarrow \infty$ . Let  $\hat{1}$  denote the all-ones vector in  $V_r$  scaled to unit length and let  $J : V_r^k \rightarrow \mathbb{R}$  be the  $k$ -linear all-ones map. Let  $\pi$  be a proper (unordered) partition of  $k$ , and assume that for every ordering  $\vec{\pi}$  of  $\pi$ ,*

$$\begin{aligned} \lambda_1(A[\psi_{\vec{\pi}}^{2^{t-1}}]) &= (1 + o(1))\psi(\hat{1}, \dots, \hat{1})^{2^{t-1}}, \\ \lambda_2(A[\psi_{\vec{\pi}}^{2^{t-1}}]) &= o\left(\lambda_1(A[\psi_{\vec{\pi}}^{2^{t-1}}])\right). \end{aligned}$$

Then for every ordering  $\vec{\pi}$  of  $\pi$ ,

$$\|\psi_{\vec{\pi}} - qJ_{\vec{\pi}}\| = o(\psi(\hat{1}, \dots, \hat{1})),$$

where  $q = \dim(V_r)^{-k/2}\psi(\hat{1}, \dots, \hat{1})$ .

For graphs,  $A[\tau^2]$  is the adjacency matrix squared so Proposition 3 states that the spectral norm of  $A - \frac{2|E(G)|}{n^2}J$  is little- $o$  of the square root of the largest eigenvalue of  $A^2$ , exactly what is proved by Chung, Graham, and Wilson (see the bottom of page 350 in [7]). The proof appears in the next section.

### 3 Algebraic properties of multilinear maps

In this section we prove several algebraic facts about multilinear maps, including Proposition 3. Throughout this section,  $V$  and  $V_i$  are finite dimensional vector spaces over  $\mathbb{R}$ . Also in this section we make no distinction between bilinear maps and matrices, using whichever formulation is convenient. We will use a symbol  $\cdot$  to denote the input to a linear map; for example, if  $\phi : V_1 \times V_2 \times V_3 \rightarrow \mathbb{R}$  is a trilinear map and  $x_1 \in V_1$  and  $x_2 \in V_2$ , then by the expression  $\phi(x_1, x_2, \cdot)$  we mean the linear map from  $V_3$  to  $\mathbb{R}$  which takes a vector  $x_3 \in V_3$  to  $\phi(x_1, x_2, x_3)$ . Lastly, we use several basic facts about tensors, all of which follow from the fact that for finite dimensional spaces, the tensor product of  $V$  and  $W$  is the vector space over  $\mathbb{R}$  of dimension  $\dim(V) \dim(W)$ . For example, if  $x$  and  $y$  are unit length, then  $x \otimes y$  is also unit length.

**Lemma 4.** *Let  $\phi : V \rightarrow \mathbb{R}$  be a linear map. There exists a vector  $v$  such that  $\phi = \langle v, \cdot \rangle$ .*

*Proof.*  $v$  is the vector dual to  $\phi$  in the dual of the vector space  $V$ . Alternatively, let the  $i$ th coordinate of  $v$  be  $\phi(e_i)$ , since then for any  $x$ ,

$$\phi(x) = \phi\left(\sum \langle x, e_i \rangle e_i\right) = \sum \langle x, e_i \rangle \phi(e_i) = \sum \langle x, e_i \rangle \langle v, e_i \rangle = \langle x, v \rangle.$$

□

**Lemma 5.** *Let  $\phi, \psi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be  $t$ -linear maps. The maps  $\phi * \psi$  and  $A[\phi^{2^t-1}]$  are well defined. Also,  $\phi * \psi$  is basis independent in the sense that the definition of  $\phi * \psi$  is independent of the choice of orthonormal basis  $b_1, \dots, b_t$  of  $V_t$ .*

*Proof.* First, extending the definitions of  $\phi * \psi$  and  $A[\phi^{2^t-1}]$  linearly to the entire domain (non-simple tensors) is well defined, since  $\phi$  and  $\psi$  are linear. That is, write each  $u_i$  and  $v_i$  in terms of some orthonormal basis and expand each tensor in  $V_i \otimes V_i$  also in terms of this basis. The linearity of  $\phi$  and  $\psi$  then shows that the definitions of  $\phi * \psi$  and  $A[\phi^{2^t-1}]$  are well defined and linear. To see basis independence of  $\phi * \psi$ , by Lemma 4 the linear map  $\phi(u_1, \dots, u_{t-1}, \cdot) : V_t \rightarrow \mathbb{R}$  equals  $\langle u', \cdot \rangle$  for some vector  $u'$ . Similarly,  $\psi(v_1, \dots, v_t, \cdot)$  equals  $\langle v', \cdot \rangle$  for some vector  $v'$ . Then

$$(\phi * \psi)(u_1 \otimes v_1, \dots, u_{t-1} \otimes v_{t-1}) = \sum_{i=1}^{\dim(V_t)} \langle u', b_i \rangle \langle v', b_i \rangle = \langle u', v' \rangle.$$

The last equality is valid for any orthonormal basis, since the dot product of  $u'$  and  $v'$  sums the product of the  $i$ th coordinate of  $u'$  in the basis  $\{b_1, \dots, b_{\dim(V_t)}\}$  with the  $i$ th coordinate of  $v'$  in the basis  $\{b_1, \dots, b_{\dim(V_t)}\}$ . □

**Definition.** For  $s \geq 0$  and  $V$  a finite dimensional vector space over  $\mathbb{R}$ , define the vector space isomorphism  $\Gamma_{V,s} : V^{\otimes 2^s} \rightarrow V^{\otimes 2^s}$  as follows. If  $s = 0$ , define  $\Gamma_{V,0}$  to be the identity map. If  $s \geq 1$ , let  $\{b_1, \dots, b_{\dim(V)}\}$  be any orthonormal basis of  $V$  and define for all  $(i_1, \dots, i_{2^s-1}, j_1, \dots, j_{2^s-1}) \in [\dim(V)]^{2^s}$ ,

$$\Gamma_{V,s}(b_{i_1} \otimes b_{j_1} \otimes \dots \otimes b_{i_{2^s-1}} \otimes b_{j_{2^s-1}}) = b_{j_1} \otimes b_{i_1} \otimes \dots \otimes b_{j_{2^s-1}} \otimes b_{i_{2^s-1}}. \quad (1)$$

Extend  $\Gamma_{V,s}$  linearly to all of  $V^{\otimes 2^s}$ .

**Remarks.**  $\Gamma_{V,s}$  is a vector space isomorphism since it restricts to a bijection of an orthonormal basis to itself. Also, it is easy to see that  $\Gamma_{V,s}$  is well defined and independent of the choice of orthonormal basis, since each  $b_i$  can be written as a linear combination of an orthonormal basis  $\{b'_1, \dots, b'_{\dim(V)}\}$  and (1) can be expanded using linearity. For notational convenience, we will usually drop the subscript  $V$  and write  $\Gamma_s$  for  $\Gamma_{V,s}$ .

**Lemma 6.** *Let  $\phi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map, let  $0 \leq s \leq t-1$ , and let  $x_1 \in V_1^{\otimes 2^s}, \dots, x_{t-s} \in V_{t-s}^{\otimes 2^s}$ . Then*

$$\phi^{2^s}(x_1, \dots, x_{t-s}) = \phi^{2^s}(\Gamma_s(x_1), \dots, \Gamma_s(x_{t-s})).$$

*Proof.* By induction on  $s$ . The base case is  $s = 0$  where  $\Gamma_0$  is the identity map. Expand the definition of  $\phi^{2^{s+1}}$  and use induction to obtain

$$\begin{aligned} \phi^{2^{s+1}}(x_1 \otimes y_1, \dots, x_{t-s-1} \otimes y_{t-s-1}) &= \sum_{j=1}^{\dim(V_{t-s}^{\otimes 2^s})} \phi^{2^s}(x_1, \dots, x_{t-s-1}, b_j) \phi^{2^s}(y_1, \dots, y_{t-s-1}, b_j) \\ &= \sum_{j=1}^{\dim(V_{t-s}^{\otimes 2^s})} \phi^{2^s}(\Gamma_s(x_1), \dots, \Gamma_s(x_{t-s-1}), \Gamma_s(b_j)) \phi^{2^s}(\Gamma_s(y_1), \dots, \Gamma_s(y_{t-s-1}), \Gamma_s(b_j)). \end{aligned}$$

But since  $\Gamma_s$  is a vector space isomorphism,  $\{\Gamma_s(b_1), \dots, \Gamma_s(b_{\dim(V_{t-s}^{\otimes 2^s})})\}$  is an orthonormal basis of  $V_{t-s}^{\otimes 2^s}$ . Thus Lemma 5 shows that

$$\begin{aligned} \sum_{j=1}^{\dim(V_{t-s}^{\otimes 2^s})} \phi^{2^s}(\Gamma_s(x_1), \dots, \Gamma_s(x_{t-s-1}), \Gamma_s(b_j)) \phi^{2^s}(\Gamma_s(y_1), \dots, \Gamma_s(y_{t-s-1}), \Gamma_s(b_j)) \\ = \phi^{2^{s+1}}(\Gamma_s(x_1) \otimes \Gamma_s(y_1), \dots, \Gamma_s(x_{t-s-1}) \otimes \Gamma_s(y_{t-s-1})) \end{aligned}$$

Finally,  $\Gamma_s(x_i) \otimes \Gamma_s(y_i) = \Gamma_{s+1}(x_i \otimes y_i)$  (write  $x_i$  and  $y_i$  as linear combinations, expand  $\Gamma_{s+1}(x_i \otimes y_i)$  using linearity, and apply (1)). Thus  $\phi^{2^{s+1}}(x_1 \otimes y_1, \dots, x_{t-s-1} \otimes y_{t-s-1}) = \phi^{2^{s+1}}(\Gamma_{s+1}(x_1 \otimes y_1), \dots, \Gamma_{s+1}(x_{t-s-1} \otimes y_{t-s-1}))$ , completing the proof.  $\square$

**Lemma 7.** *Let  $V_1, \dots, V_t$  be finite dimensional vector spaces over  $\mathbb{R}$ . If  $\phi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  is a  $t$ -linear map, then  $A[\phi^{2^{t-1}}]$  is a square symmetric real valued matrix.*

*Proof.* Let  $\phi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map.  $A[\phi^{2^{t-1}}]$  is a bilinear map from  $V_1^{\otimes 2^{t-2}} \times V_1^{\otimes 2^{t-2}} \rightarrow \mathbb{R}$  and so is a square matrix of dimension  $\dim(V_1)^{2^{t-2}}$ . Lemma 6 shows that  $A[\phi^{2^{t-1}}]$  is a symmetric matrix, since

$$\begin{aligned} A[\phi^{2^{t-1}}](x_1 \otimes \dots \otimes x_{2^{t-2}}, y_1 \otimes \dots \otimes y_{2^{t-2}}) &= \phi^{2^{t-1}}(x_1 \otimes y_1 \otimes \dots \otimes x_{2^{t-2}} \otimes y_{2^{t-2}}) \\ &= \phi^{2^{t-1}}(\Gamma(x_1 \otimes y_1 \otimes \dots \otimes x_{2^{t-2}} \otimes y_{2^{t-2}})) \\ &= \phi^{2^{t-1}}(y_1 \otimes x_1 \otimes \dots \otimes y_{2^{t-2}} \otimes x_{2^{t-2}}) \\ &= A[\phi^{2^{t-1}}](y_1 \otimes \dots \otimes y_{2^{t-2}}, x_1 \otimes \dots \otimes x_{2^{t-2}}). \end{aligned}$$

The above equation is valid for all  $x_i, y_i \in V_1$ , in particular for all basis elements of  $V_1$  which implies that  $A[\phi^{2^{t-1}}](w, z) = A[\phi^{2^{t-1}}](z, w)$  for all basis vectors  $w, z$  of  $V_1^{\otimes 2^{t-2}}$ . Thus  $A[\phi^{2^{t-1}}]$  is a square symmetric real-valued matrix.  $\square$

**Lemma 8.** *Let  $\phi : V_1 \times \cdots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map and let  $x_1 \in V_1, \dots, x_t \in V_t$  be unit length vectors. Then*

$$|\phi(x_1, \dots, x_t)|^2 \leq |\phi^2(x_1 \otimes x_1, \dots, x_{t-1} \otimes x_{t-1})|.$$

*Proof.* Consider the linear map  $\phi(x_1, \dots, x_{t-1}, \cdot)$  which is a linear map from  $V_t$  to  $\mathbb{R}$ . By Lemma 4, there exists a vector  $w \in V_t$  such that  $\phi(x_1, \dots, x_{t-1}, \cdot) = \langle w, \cdot \rangle$ . Now expand out the definition of  $\phi^2$ :

$$\phi^2(x_1 \otimes x_1, \dots, x_{t-1} \otimes x_{t-1}) = \sum_j |\phi(x_1, \dots, x_{t-1}, b_j)|^2 = \sum_j |\langle w, b_j \rangle|^2 = \langle w, w \rangle$$

where the last equality is because  $\{b_j\}$  is an orthonormal basis of  $V_t$ . Since  $\|w\| = \sqrt{\langle w, w \rangle}$ ,

$$|\phi^2(x_1 \otimes x_1, \dots, x_{t-1} \otimes x_{t-1})| = |\langle w, w \rangle| = \left| \left\langle w, \frac{w}{\|w\|} \right\rangle \right|^2.$$

But since  $x_t$  is unit length and  $\langle w, \cdot \rangle$  is maximized over the unit ball at vectors parallel to  $w$  (so maximized at  $w/\|w\|$ ),  $\left| \left\langle w, \frac{w}{\|w\|} \right\rangle \right| \geq |\langle w, x_t \rangle|$ . Thus

$$|\phi^2(x_1 \otimes x_1, \dots, x_{t-1} \otimes x_{t-1})| = \left| \left\langle w, \frac{w}{\|w\|} \right\rangle \right|^2 \geq |\langle w, x_t \rangle|^2 = |\phi(x_1, \dots, x_t)|^2.$$

The last equality used the definition of  $w$ , that  $\phi(x_1, \dots, x_{t-1}, \cdot) = \langle w, \cdot \rangle$ .  $\square$

**Lemma 9.** *Let  $\phi : V_1 \times \cdots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map and let  $x_1 \in V_1, \dots, x_t \in V_t$  be unit length vectors. Then for  $0 \leq s \leq t-1$ ,*

$$|\phi(x_1, \dots, x_t)|^{2^s} \leq \left| \phi^{2^s}(\underbrace{x_1 \otimes \cdots \otimes x_1}_{2^s}, \dots, \underbrace{x_{t-s} \otimes \cdots \otimes x_{t-s}}_{2^s}) \right|$$

which implies that

$$|\phi(x_1, \dots, x_t)|^{2^{t-1}} \leq \left| A[\phi^{2^{t-1}}](\underbrace{x_1 \otimes \cdots \otimes x_1}_{2^{t-2}}, \underbrace{x_1 \otimes \cdots \otimes x_1}_{2^{t-2}}) \right|.$$

*Proof.* By induction on  $s$ . The base case is  $s = 0$  where both sides are equal and the induction step follows from Lemma 8. By definition of  $A[\phi^{2^{t-1}}]$ ,

$$\left| A[\phi^{2^{t-1}}](\underbrace{x_1 \otimes \cdots \otimes x_1}_{2^{t-2}}, \underbrace{x_1 \otimes \cdots \otimes x_1}_{2^{t-2}}) \right| = \left| \phi^{2^{t-1}}(\underbrace{x_1 \otimes \cdots \otimes x_1}_{2^{t-1}}) \right|,$$

completing the proof.  $\square$



**Lemma 10.** Let  $V_1, \dots, V_t$  be vector spaces over  $\mathbb{R}$  and let  $\phi : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be a  $t$ -linear map. Then  $\|\phi\|^{2^{t-1}} \leq \lambda_1(A[\phi^{2^{t-1}}])$ .

*Proof.* Pick  $x_1, \dots, x_t$  unit length vectors to maximize  $\phi$ , so  $\phi(x_1, \dots, x_t) = \|\phi\|$ . Then Lemma 9 shows that

$$\|\phi\|^{2^{t-1}} = |\phi(x_1, \dots, x_t)|^{2^{t-1}} \leq \left| A[\phi^{2^{t-1}}](\underbrace{x_1 \otimes \dots \otimes x_1}_{2^{t-2}}, \underbrace{x_1 \otimes \dots \otimes x_1}_{2^{t-2}}) \right|$$

Since  $x_1 \otimes \dots \otimes x_1$  is unit length, the above expression is upper bounded by the spectral norm of  $A[\phi^{2^{t-1}}]$ .  $\square$

**Lemma 11.** Let  $\{M_r\}_{r \rightarrow \infty}$  be a sequence of square symmetric real-valued matrices with dimension going to infinity where  $\lambda_2(M_r) = o(\lambda_1(M_r))$ . Let  $u_r$  be a unit length eigenvector corresponding to the largest eigenvalue in absolute value of  $M_r$ . If  $\{x_r\}$  is a sequence of unit length vectors such that  $|x_r^T M_r x_r| = (1 + o(1))\lambda_1(M_r)$ , then

$$\|u_r - x_r\| = o(1).$$

Consequently, for any unit length sequence  $\{y_r\}$  where each  $y_r$  is perpendicular to  $x_r$ ,

$$|y_r^T M_r y_r| = o(\lambda_1(M_r)).$$

*Proof.* Throughout this proof, the subscript  $r$  is dropped; all terms  $o(\cdot)$  should be interpreted as  $r \rightarrow \infty$ . This exact statement was proved by Chung, Graham, and Wilson [7], although they don't clearly state it as such. We give a proof here for completeness using slightly different language but the same proof idea: if  $x$  projected onto  $u^\perp$  is too big then the second largest eigenvalue is too big. Write  $x = \alpha v + \beta u$  where  $v$  is a unit length vector perpendicular to  $u$  and  $\alpha, \beta \in \mathbb{C}$  and  $\alpha^2 + \beta^2 = 1$  (since  $u$  is an eigenvector it might have complex entries). Let  $\phi(x, y) = x^T M y$  be the bilinear map corresponding to  $M$ . Since  $u^T M v = \lambda_1 u^T v = \lambda_1 \langle u, v \rangle = 0$ , we have  $\phi(u, v) = 0$ . This implies that

$$\begin{aligned} \phi(x, x) &= \phi(\alpha v + \beta u, \alpha v + \beta u) = \alpha^2 \phi(v, v) + \beta^2 \phi(u, u) + 2\alpha\beta \phi(u, v) \\ &= \alpha^2 \phi(v, v) + \beta^2 \phi(u, u). \end{aligned}$$

The second largest eigenvalue of  $M$  is the largest eigenvalue of  $M - \lambda_1(M)uu^T$  which is the spectral norm of  $M - \lambda_1(M)uu^T$ . Thus

$$|\phi(v, v)| = |v^T M v| = |v^T (M - \lambda_1(M)uu^T)v| \leq \lambda_2(M). \quad (2)$$

Using that  $\phi(u, u) = \lambda_1(M)$  and the triangle inequality, we obtain

$$|\phi(x, x)| \leq \alpha^2 \lambda_2(M) + \beta^2 \lambda_1(M). \quad (3)$$

Since  $\alpha^2 + \beta^2 = 1$ ,  $|\alpha|$  and  $|\beta|$  are between zero and one. Combining this with (3) and  $|\phi(x, x)| = (1 + o(1))\lambda_1(M)$  and  $\lambda_2(M) = o(\lambda_1(M))$ , we must have  $|\beta| = 1 + o(1)$  which in turn implies that  $|\alpha| = o(1)$ . Consequently,

$$\|u - x\|^2 = \langle u - x, u - x \rangle = \langle u, u \rangle + \langle x, x \rangle - 2\langle u, x \rangle = 2 - 2\beta = o(1).$$

Now consider some  $y$  perpendicular to  $x$  and similarly to the above, write  $y = \gamma w + \delta u$  for some unit length vector  $w$  perpendicular to  $u$  and  $\gamma, \delta \in \mathbb{C}$  with  $\gamma^2 + \delta^2 = 1$ . Then

$$\phi(y, y) = \phi(\gamma w + \delta u, \gamma w + \delta u) = \gamma^2 \phi(w, w) + \delta^2 \phi(u, u)$$

and as in (2), we have  $|\phi(w, w)| \leq \lambda_2(M)$ . Thus

$$|\phi(y, y)| \leq \gamma^2 \lambda_2(M) + \delta^2 \lambda_1(M).$$

We want to conclude that the above expression is  $o(\lambda_1(M))$ . Since  $\lambda_2(M) = o(\lambda_1(M))$ , we must prove that  $|\delta| = o(1)$  to complete the proof.

$$\delta = \langle y, u \rangle = \left\langle y, \frac{x - \alpha v}{\beta} \right\rangle = \frac{1}{\beta} (\langle y, x \rangle - \alpha \langle y, v \rangle) = \frac{-\alpha \langle y, v \rangle}{\beta}.$$

But  $|\alpha| = o(1)$ ,  $|\beta| = 1 + o(1)$ , and  $\|y\| = \|v\| = 1$  so  $|\delta| = o(1)$  as required.  $\square$

**Lemma 12.** *Let  $J : V_1 \times \dots \times V_t \rightarrow \mathbb{R}$  be the all-ones map and let  $\vec{1}_i$  be the all-ones vector in  $V_i$ . Then for all  $x_1, \dots, x_t$  with  $x_i \in V_i$ ,*

$$J(x_1, \dots, x_t) = \left\langle \vec{1}_1, x_1 \right\rangle \cdots \left\langle \vec{1}_t, x_t \right\rangle. \quad (4)$$

*Proof.* If  $x_1, \dots, x_t$  are standard basis vectors, then the left and right hand side of (4) are the same. By linearity, (4) is then the same for all  $x_1, \dots, x_t$ .  $\square$

*Proof of Proposition 3.* Again throughout this proof, the subscript  $r$  is dropped; all terms  $o(\cdot)$  should be interpreted as  $r \rightarrow \infty$ . Let  $\hat{1}$  denote the all-ones vector scaled to unit length in the appropriate vector space. Pick an ordering  $\vec{\pi} = (k_1, \dots, k_t)$  of  $\pi$ . The definition of spectral norm is independent of the choice of the ordering for the entries of  $\vec{\pi}$ , so  $\|\psi_{\vec{\pi}} - qJ_{\vec{\pi}}\|$  is the same for all orderings. Let  $w_1, \dots, w_t$  be unit length vectors where  $(\psi_{\vec{\pi}} - qJ_{\vec{\pi}})(w_1, \dots, w_t) = \|\psi_{\vec{\pi}} - qJ_{\vec{\pi}}\|$  and write  $w_i = \alpha_i y_i + \beta_i \hat{1}$  where  $y_i$  is a unit length vector perpendicular to the all-ones vector and  $\alpha_i, \beta_i \in \mathbb{R}$  with  $\alpha_i^2 + \beta_i^2 = 1$ . Then

$$\begin{aligned} \|\psi_{\vec{\pi}} - qJ_{\vec{\pi}}\| &= (\psi_{\vec{\pi}} - qJ_{\vec{\pi}})(w_1, \dots, w_t) = (\psi_{\vec{\pi}} - qJ_{\vec{\pi}})(\alpha_1 y_1 + \beta_1 \hat{1}, \dots, \alpha_t y_t + \beta_t \hat{1}) \\ &= \psi_{\vec{\pi}}(\alpha_1 y_1 + \beta_1 \hat{1}, \dots, \alpha_t y_t + \beta_t \hat{1}) - q \dim(V_r)^{k/2} \prod \beta_i. \end{aligned} \quad (5)$$

The last equality used that  $y_i$  is perpendicular to  $\hat{1}$ , so Lemma 12 implies that if  $y_i$  appears as input to  $J_{\vec{\pi}}$  then the outcome is zero no matter what the other vectors are. Thus the only non-zero term involving  $J_{\vec{\pi}}$  is  $J_{\vec{\pi}}(\hat{1}, \dots, \hat{1}) = \dim(V_r)^{k/2}$ . Note that  $\psi(\hat{1}, \dots, \hat{1}) = \psi_{\vec{\pi}}(\hat{1}, \dots, \hat{1})$

since the all-ones vector scaled to unit length in  $V^{\otimes k_i}$  is the tensor product of the all-ones vector scaled to unit length in  $V$ . Inserting  $q = \dim(V_r)^{-k/2} \psi_{\vec{\pi}}(\hat{1}, \dots, \hat{1})$  in (5), we obtain

$$\|\psi_{\vec{\pi}} - qJ_{\vec{\pi}}\| = \psi_{\vec{\pi}}(\alpha_1 y_1 + \beta_1 \hat{1}, \dots, \alpha_t y_t + \beta_t \hat{1}) - \left( \prod \beta_i \right) \psi_{\vec{\pi}}(\hat{1}, \dots, \hat{1}). \quad (6)$$

Now consider expanding  $\psi_{\vec{\pi}}$  in (6) using linearity; the term  $(\prod \beta_i) \psi_{\vec{\pi}}(\hat{1}, \dots, \hat{1})$  cancels, so all terms include at least one  $y_i$ . We claim that each of these terms is small; the following claim finishes the proof, since  $\|\psi_{\vec{\pi}} - qJ_{\vec{\pi}}\|$  is the sum of terms each of which  $o(\psi(\hat{1}, \dots, \hat{1}))$ .

*Claim:* If  $z_1, \dots, z_{i-1}, z_{i+1}, \dots, z_t$  are unit length vectors, then

$$|\psi_{\vec{\pi}}(z_1, \dots, z_{i-1}, y_i, z_{i+1}, \dots, z_t)| = o(\psi(\hat{1}, \dots, \hat{1})).$$

*Proof.* Change the ordering of  $\vec{\pi}$  to an ordering  $\vec{\pi}'$  that differs from  $\vec{\pi}$  by swapping 1 and  $i$ . Since  $\psi$  is symmetric,

$$\psi_{\vec{\pi}}(z_1, \dots, z_{i-1}, y_i, z_{i+1}, \dots, z_t) = \psi_{\vec{\pi}'}(y_i, z_2, \dots, z_{i-1}, z_1, z_{i+1}, \dots, z_t). \quad (7)$$

Therefore proving the claim comes down to bounding  $\psi_{\vec{\pi}'}(y_i, z_2, \dots, z_{i-1}, z_1, z_{i+1}, \dots, z_t)$ , which is a combination of Lemma 9 and Lemma 11 as follows. For the remainder of this proof, denote by  $A$  the matrix  $A[\psi_{\vec{\pi}'}^{2^{t-1}}]$ . By assumption, we have  $\lambda_2(A) = o(\lambda_1(A))$  so Lemma 11 can be applied to the matrix sequence  $A$ . Next we would like to show that we can use  $\hat{1}$  for  $x$  in the statement of Lemma 11; i.e. that  $A(\hat{1}, \hat{1}) = (1 + o(1))\lambda_1(A)$ . By Lemma 9 and the assumption  $\lambda_1(A) = (1 + o(1))\psi(\hat{1}, \dots, \hat{1})^{2^{t-1}}$ , we have

$$|\psi_{\vec{\pi}'}(\hat{1}, \dots, \hat{1})|^{2^{t-1}} \leq |A(\hat{1}, \hat{1})| \leq \lambda_1(A) = (1 + o(1))\psi(\hat{1}, \dots, \hat{1})^{2^{t-1}}.$$

Using the definition of  $\psi_{\vec{\pi}'}$ , we have  $\psi_{\vec{\pi}'}(\hat{1}, \dots, \hat{1}) = \psi(\hat{1}, \dots, \hat{1})$ , which implies asymptotic equality through the above equation. In particular,  $|A(\hat{1}, \hat{1})| = (1 + o(1))\lambda_1(A)$  which is the condition in Lemma 11 for  $x = \hat{1}$ . Lastly, to apply Lemma 11 we need a vector  $y$  perpendicular to  $\hat{1}$ . The vector  $y_i \otimes \dots \otimes y_i \in V^{\otimes k_i 2^{t-2}}$  is perpendicular to  $\hat{1}$  (in  $V^{\otimes k_i 2^{t-2}}$ ) since  $y_i$  itself is perpendicular to  $\hat{1}$  (in  $V^{\otimes k_i}$ ). Thus Lemma 11 implies that

$$\left| A(\underbrace{y_i \otimes \dots \otimes y_i}_{2^{t-2}}, \underbrace{y_i \otimes \dots \otimes y_i}_{2^{t-2}}) \right| = o(\lambda_1(A)). \quad (8)$$

Using Lemma 9 again shows that

$$|\psi_{\vec{\pi}'}(y_i, z_2, \dots, z_{i-1}, z_1, z_{i+1}, \dots, z_t)|^{2^{t-1}} \leq \left| A(\underbrace{y_i \otimes \dots \otimes y_i}_{2^{t-2}}, \underbrace{y_i \otimes \dots \otimes y_i}_{2^{t-2}}) \right|.$$

Combining this equation with (7) and (8) shows that  $|\psi_{\vec{\pi}}(z_1, \dots, z_{i-1}, y_i, z_{i+1}, \dots, z_t)|^{2^{t-1}} = o(\lambda_1(A))$ . By assumption,  $\lambda_1(A) = (1 + o(1))\psi(\hat{1}, \dots, \hat{1})^{2^{t-1}}$ , completing the proof of the claim.  $\square$

## 4 $\text{Cycle}_{4\ell}[\pi] \Rightarrow \text{Eig}[\pi]$

In this section, we prove that  $\text{Cycle}_{4\ell}[\pi] \Rightarrow \text{Eig}[\pi]$  for any hypergraph sequence  $\mathcal{H}$  with  $|V(H_n)| = n$  and  $|E(H_n)| \geq p\binom{n}{k} + o(n^k)$ . We require the following result from [16].

**Proposition 13.** [16, Proposition 6] *Let  $H$  be a  $k$ -uniform hypergraph, let  $\vec{\pi}$  be a proper ordered partition of  $k$ , and let  $\ell \geq 2$  be an integer. Let  $\tau$  be the adjacency map of  $H$ . Then  $\text{Tr} \left[ A[\tau_{\vec{\pi}}^{2^{t-1}}]^\ell \right]$  is the number of labeled circuits of type  $\vec{\pi}$  and length  $2\ell$  in  $H$ .*

Propositions 3 and 13 and Lemma 10 combine to prove that  $\text{Cycle}_{4\ell}[\pi] \Rightarrow \text{Eig}[\pi]$  for any proper partition  $\pi$  as follows.

*Proof that  $\text{Cycle}_{4\ell}[\pi] \Rightarrow \text{Eig}[\pi]$ .* Let  $\mathcal{H} = \{H_n\}_{n \rightarrow \infty}$  be a sequence of hypergraphs and let  $\tau_n$  be the adjacency map of  $H_n$ . For notational convenience, the subscript on  $n$  is dropped below. Throughout this proof, we use  $\hat{1}$  to denote the all-ones vector scaled to unit length. Wherever we use the notation  $\hat{1}$ , it is the input to a multilinear map and so  $\hat{1}$  denotes the all-ones vector in the appropriate vector space corresponding to whatever space the map is expecting as input. This means that in the equations below  $\hat{1}$  can stand for different vectors in the same expression, but attempting to subscript  $\hat{1}$  with the vector space (for example  $\hat{1}_{V_3}$ ) would be notationally awkward.

The proof that  $\text{Cycle}_{4\ell}[\pi] \Rightarrow \text{Eig}[\pi]$  comes down to checking the conditions of Proposition 3. Let  $\vec{\pi}$  be any ordering of the entries of  $\pi$ . We will show that the first and second largest eigenvalues of  $A = A[\tau_{\vec{\pi}}^{2^{t-1}}]$  are separated. Let  $m = |E(C_{\pi,4\ell})| = 2\ell 2^{t-1}$  and note that  $|V(C_{\pi,4\ell})| = mk/2$  since  $C_{\pi,4\ell}$  is two-regular.  $A$  is a square symmetric real valued matrix, so let  $\mu_1, \dots, \mu_d$  be the eigenvalues of  $A$  arranged so that  $|\mu_1| \geq \dots \geq |\mu_d|$ , where  $d = \dim(A)$ . The eigenvalues of  $A^{2\ell}$  are  $\mu_1^{2\ell}, \dots, \mu_d^{2\ell}$  and the trace of  $A^{2\ell}$  is  $\sum_i \mu_i^{2\ell}$ . Since all  $\mu_i^{2\ell} \geq 0$ , Proposition 13 and  $\text{Cycle}_{4\ell}[\pi]$  implies that

$$\mu_1^{2\ell} + \mu_2^{2\ell} \leq \text{Tr} [A^{2\ell}] = \#\{\text{possibly degenerate } C_{\pi,4\ell} \text{ in } H_n\} \leq p^m n^{mk/2} + o(n^{mk/2}). \quad (9)$$

We now verify the conditions on  $\mu_1$  and  $\mu_2$  in Proposition 3, and to do that we need to compute  $\tau(\hat{1}, \dots, \hat{1})$ . Simple computations show that

$$\tau(\hat{1}, \dots, \hat{1}) = \tau_{\vec{\pi}}(\hat{1}, \dots, \hat{1}) = \frac{k!E(H)}{n^{k/2}}. \quad (10)$$

Using that  $|E(H_n)| \geq p\binom{n}{k} + o(n^k)$ , Lemma 10, and  $\mu_1^{2\ell} \leq p^m n^{mk/2} + o(n^{mk/2})$  from (9),

$$pn^{k/2} + o(n^{k/2}) \leq \frac{k!E(H)}{n^{k/2}} = \tau_{\vec{\pi}}(\hat{1}, \dots, \hat{1}) \leq \|\tau_{\vec{\pi}}\| \leq \mu_1^{1/2^{t-1}} \leq pn^{k/2} + o(n^{k/2}). \quad (11)$$

This implies equality up to  $o(n^{k/2})$  throughout the above expression, so  $\tau(\hat{1}, \dots, \hat{1}) = pn^{k/2} + o(n^{k/2})$ ,  $\lambda_{1,\pi}(H_n) = \|\tau_{\vec{\pi}}\| = pn^{k/2} + o(n^{k/2})$ , and  $\mu_1 = p^{2^{t-1}} n^{k2^{t-2}} + o(n^{k2^{t-2}})$ , so  $\mu_1 = (1 + o(1))\tau(\hat{1}, \dots, \hat{1})^{2^{t-1}}$ .

Insert  $\mu_1 = p^{2^{t-1}} n^{k2^{t-2}} + o(n^{k2^{t-2}})$  into (9) to show that  $\mu_2 = o(n^{k2^{t-2}})$ . Therefore, the conditions of Proposition 3 are satisfied, so

$$\|\tau_{\vec{\pi}} - qJ_{\vec{\pi}}\| = o(\tau(\hat{1}, \dots, \hat{1})) = o(n^{k/2}),$$

where  $q = n^{-k/2}\tau(\hat{1}, \dots, \hat{1})$ . Using (10),  $q = k!|E(H)|/n^k$ . Thus  $\|\tau_{\vec{\pi}} - qJ_{\vec{\pi}}\| = \lambda_{2,\pi}(H_n)$  and the proof is complete.  $\square$

The above proof can be extended to even length cycles in the case when  $\vec{\pi} = (k_1, k_2)$  is a partition into two parts. For these  $\vec{\pi}$ , the matrix  $A[\tau_{\vec{\pi}}^2]$  can be shown to be positive semidefinite since  $A[\tau_{\vec{\pi}}^2]$  will equal  $MM^T$  where  $M$  is the matrix associated to the bilinear map  $\tau_{\vec{\pi}}$ . Since  $A[\tau_{\vec{\pi}}^2]$  is positive semidefinite, each  $\mu_i \geq 0$  so any power of  $\mu_i$  is non-negative. For partitions into more than two parts, we don't know if the matrix  $A[\tau_{\vec{\pi}}^{2^{t-1}}]$  is always positive semidefinite or not.

## References

- [1] T. Austin and T. Tao. Testability and repair of hereditary hypergraph properties. *Random Structures Algorithms*, 36(4):373–463, 2010.
- [2] F. Chung. Quasi-random hypergraphs revisited. *Random Structures Algorithms*, 40(1):39–48, 2012.
- [3] F. R. K. Chung. Quasi-random classes of hypergraphs. *Random Structures Algorithms*, 1(4):363–382, 1990.
- [4] F. R. K. Chung and R. L. Graham. Quasi-random hypergraphs. *Random Structures Algorithms*, 1(1):105–124, 1990.
- [5] F. R. K. Chung and R. L. Graham. Quasi-random set systems. *J. Amer. Math. Soc.*, 4(1):151–196, 1991.
- [6] F. R. K. Chung and R. L. Graham. Cohomological aspects of hypergraphs. *Trans. Amer. Math. Soc.*, 334(1):365–388, 1992.
- [7] F. R. K. Chung, R. L. Graham, and R. M. Wilson. Quasi-random graphs. *Combinatorica*, 9(4):345–362, 1989.
- [8] D. Conlon, H. Hàn, Y. Person, and M. Schacht. Weak quasi-randomness for uniform hypergraphs. *Random Structures Algorithms*, 40(1):1–38, 2012.
- [9] P. Frankl and V. Rödl. The uniformity lemma for hypergraphs. *Graphs Combin.*, 8(4):309–312, 1992.
- [10] J. Friedman. Some graphs with small second eigenvalue. *Combinatorica*, 15(1):31–42, 1995.

- [11] J. Friedman and A. Wigderson. On the second eigenvalue of hypergraphs. *Combinatorica*, 15(1):43–65, 1995.
- [12] W. T. Gowers. Quasirandomness, counting and regularity for 3-uniform hypergraphs. *Combin. Probab. Comput.*, 15(1-2):143–184, 2006.
- [13] P. Keevash. A hypergraph regularity method for generalized Turán problems. *Random Structures Algorithms*, 34(1):123–164, 2009.
- [14] Y. Kohayakawa, B. Nagle, V. Rödl, and M. Schacht. Weak hypergraph regularity and linear hypergraphs. *J. Combin. Theory Ser. B*, 100(2):151–160, 2010.
- [15] Y. Kohayakawa, V. Rödl, and J. Skokan. Hypergraphs, quasi-randomness, and conditions for regularity. *J. Combin. Theory Ser. A*, 97(2):307–352, 2002.
- [16] J. Lenz and D. Mubayi. Eigenvalues and linear quasirandom hypergraphs. submitted. <http://arxiv.org/abs/1208.4863>.
- [17] J. Lenz and D. Mubayi. The poset of hypergraph quasirandomness. accepted in *Random Structures and Algorithms*. <http://arxiv.org/abs/1208.5978>.
- [18] A. Thomason. Pseudorandom graphs. In *Random graphs '85 (Poznań, 1985)*, volume 144 of *North-Holland Math. Stud.*, pages 307–331. North-Holland, Amsterdam, 1987.
- [19] A. Thomason. Random graphs, strongly regular graphs and pseudorandom graphs. In *Surveys in combinatorics 1987 (New Cross, 1987)*, volume 123 of *London Math. Soc. Lecture Note Ser.*, pages 173–195. Cambridge Univ. Press, Cambridge, 1987.