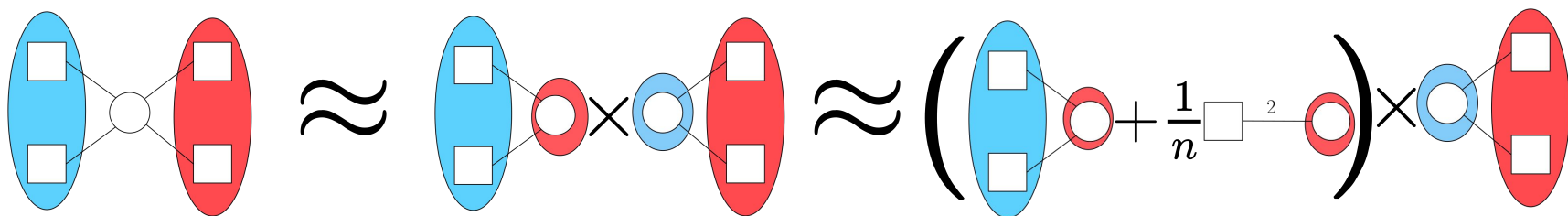


# Sum-of-Squares Lower Bounds for Sherrington-Kirkpatrick via Planted Affine Planes



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# Introduction

**Sherrington-Kirkpatrick (SK) problem:** Given  $W \in \mathbb{R}^{n \times n}$  sampled from GOE(n), find the maximum value of  $x^T W x$  over  $x \in \{+1, -1\}^n$

GOE(n): Gaussian Orthogonal Ensemble. A symmetric  $n \times n$  matrix where the off-diagonal entries are  $N(0, 1)$  and the diagonal entries are  $N(0, 2)$ .

Important problem considered before in

- Computer Science - Can be viewed as average-case MaxCut on random graphs where  $x$  encodes the partition
- Statistical Physics - Can be viewed as minimizing energy of a physical system, where  $x$  encodes spin values in a spin-glass model

# Facts about SK

Define  $\text{OPT}(W) = \max (x^T W x)$  over all  $x \in \{+1, -1\}^n$

$\text{OPT}(W) \leq (2 + o(1)) n^{1.5}$  whp over  $W \sim \text{GOE}(n)$ , because  $W$  has maximum eigenvalue  $(2 + o(1)) n^{0.5}$  whp

[Conjectured by Parisi '79, proved by Talagrand '06]

$\text{OPT}(W) \approx 2 P^* n^{1.5} \approx 1.52 n^{1.5}$  whp over  $W \sim \text{GOE}(n)$

Here,  $P^*$  is known as the **Parisi constant**

# Optimization vs Certification

**Optimization problem:** Given  $W \sim \text{GOE}(n)$ , maximize  $x^\top W x$  over  $x \in \{+1, -1\}^n$

Recently, a breakthrough result by Montanari provided a polynomial time optimization algorithm for SK!

[Montanari '19] There is a  $\text{poly}(n, 1/\varepsilon)$  time algorithm that w.h.p. outputs  $x \in \{+1, -1\}^n$  satisfying  $x^\top W x \geq (2P^* - \varepsilon) n^{1.5}$  (*under a statistical physics conjecture*)

In this work, we are concerned with the **certification problem**:

Given  $W \sim \text{GOE}(n)$ , output an upper bound on  $\max (x^\top W x)$  over  $x \in \{+1, -1\}^n$  that gets close to the true optimum whp.

# The Sum-of-Squares hierarchy

Sum-of-Squares is a powerful certification method for a polynomial objective/constraints.

The algorithm is parameterized by degree  $d$ , with larger  $d$  = more powerful SoS.

**Main question:** what bound can SoS certify for the SK problem?

[Montanari-Sen '16, Kunisky-Bandeira '19, Mohanty-Raghavendra-Xu '20, Kunisky '20]

Whp degree-6 SoS has value  $(2 - o(1)) n^{1.5}$

**This work:** For some constant  $\delta > 0$ , whp degree- $n^\delta$  SoS still has value  $(2 - o(1)) n^{1.5}$

Rules out degree- $O(1)$  SoS = polynomial time SoS

# Planted Affine Planes

**Planted Affine Planes (PAP) problem:** Let  $n \ll m$ . If  $d_1, \dots, d_m$  are random vectors in  $\mathbb{R}^n$  sampled from  $N(0, I_n)$ , can we prove that there is no vector  $v \in \mathbb{R}^n$  such that  $\langle v, d_u \rangle^2 = 1$  for all  $u = 1, 2, \dots, m$ ?

**This work:** For  $m \leq n^{1.5 - \epsilon}$ , w.h.p. over  $d_1, \dots, d_m \sim N(0, I_n)$ , degree- $n^\delta$  SoS thinks this system of equations is feasible.

Remainder of talk: SoS lower bound for PAP

# [BHKMP '16] Recipe for SoS lower bounds

Goal for SoS lower bounds: Construct degree- $D$  pseudodistribution of solutions, specified by **pseudomoments**  $\tilde{\mathbb{E}}[v^S]$  for all subsets  $S$  of  $\{1, \dots, n\}$  of size at most  $D$ .

Equivalently, construct a **moment matrix**  $\mathcal{M}$  with rows, columns indexed by subsets of  $\{1, \dots, n\}$  of size  $\leq D/2$  such that (1)  $\mathcal{M}$  obeys some linear constraints on its entries, and (2)  $\mathcal{M} \succeq 0$ .

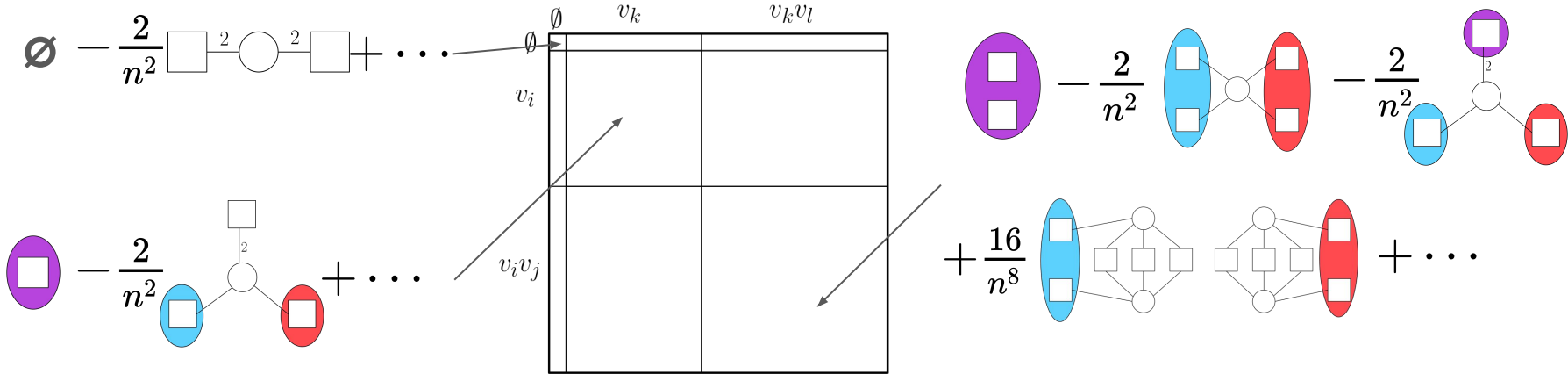
	$\emptyset$	$v_k$	$v_k v_l$
$\emptyset$			
$v_i$			
$v_i v_j$			$\tilde{\mathbb{E}}[v^I v^J]$

[BHKMP '16] Recipe for average-case problems

1. Construct a candidate  $\mathcal{M}$  via **pseudocalibration**
2. Decompose  $\mathcal{M}$  into **graph matrices** and use the decomposition to prove  $\mathcal{M} \succeq 0$

# Graph matrix decomposition

$$\mathcal{M} = \sum_{\text{shapes } \beta} c_{\beta} M_{\beta} = \sum_{\substack{\text{shapes } \beta: \\ \beta \text{ satisfies parity constraints}}} \prod_{\circ \in V(\beta)} h_{\deg(\circ)}(1) \cdot \frac{M_{\beta}}{n^{|E(\beta)|/2}}$$



Theorem [AMP '20]: Whp for all shapes  $\beta$ : let  $S$  be a **minimum-weight vertex separator** of the left and right sides of  $\beta$ ,

$$\|M_{\beta}\| \leq \tilde{O}(\sqrt{m}^{\#\circ \text{ not in } S} \cdot \sqrt{n}^{\#\square \text{ not in } S})$$

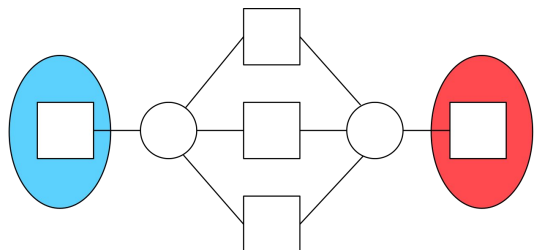


# Graph matrix decomposition

$$\mathcal{M} = \sum_{\text{shapes } \beta} c_\beta M_\beta = \sum_{\substack{\text{shapes } \beta: \\ \beta \text{ satisfies parity constraints}}} \prod_{o \in V(\beta)} h_{\deg(o)}(1) \cdot \frac{M_\beta}{n^{|E(\beta)|/2}}$$

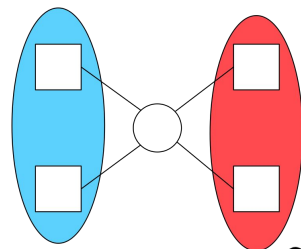
Naive idea: show that  $\sum_{\beta \neq \text{identity}} |c_\beta| \|M_\beta\| = o(1)$

Some terms do have  $o(1)$  norm...



$$|c_\beta| \cdot \|M_\beta\| = o(1)$$

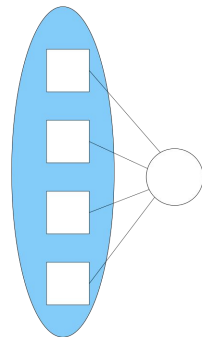
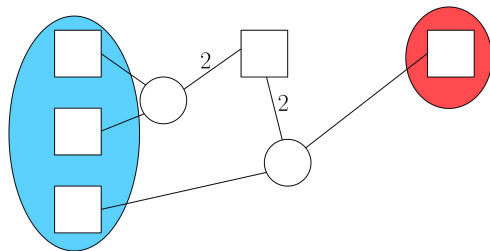
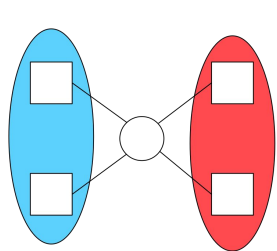
...but others do not



$$|c_\beta| \cdot \|M_\beta\| = \tilde{\Omega}(1)$$

# Spiders

Def: a **spider** has two degree-1 squares in left side or right side adjacent to same circle



Theorem:  $\sum_{\substack{\text{non-spider,} \\ \text{non-identity } \beta}} |c_\beta| \cdot \|M_\beta\| = o(1)$

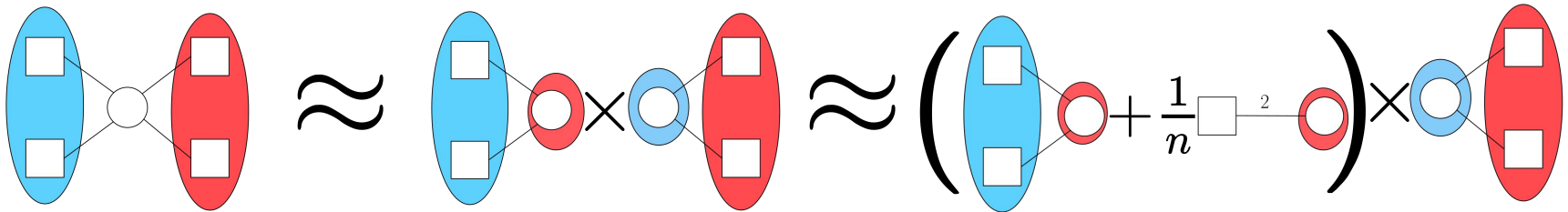
So the only large-norm shapes are the spiders.

# Spiders

It suffices to prove that  $\mathcal{M}$  is PSD on  $\text{Null}(\mathcal{M})^\perp$

Spiders (the large-norm shapes) are approximately zero on  $\text{Null}(\mathcal{M})^\perp$ !

$$\mathcal{M} \times \left( \text{Spider} + \frac{1}{n} \text{Square} \right) = 0$$



Theorem: There is  $\mathcal{M}'$  such that  $x^\top \mathcal{M} x = x^\top \mathcal{M}' x$  for all  $x \in \text{Null}(\mathcal{M})^\perp$  and all eigenvalues of  $\mathcal{M}'$  are  $1 \pm o(1)$

# Open Problems

Interpret Montanari's algorithm as rounding SoS?

Average-case sparse MaxCut

Improve PAP assumption  $m \leq n^{1.5}$  to  $m \leq n^2$

Improve SoS degree to  $n/\log n$

# Proof outline

Reduce the SK problem to the Planted Boolean Vector problem  
[Mohanty-Raghavendra-Xu '19]

Consider the dual version of Planted Boolean Vector problem, which we term  
**Planted Affine Planes (PAP)**

Directly prove an SoS lower bound for PAP via the following steps

1. Construct a candidate solution using pseudocalibration
2. Prove positive-semidefiniteness of the candidate moment matrix  $M$

The PSDness proof is the most innovative part of this work

# Outline

1. The Sherrington-Kirkpatrick problem
2. Reduction to Planted Affine Planes
3. Pseudocalibration + graph matrices
4. PSD-ness sketch
5. Open Problems

# The Sum-of-Squares hierarchy

Sum-of-Squares looks for a refutation proof of a system of polynomial constraints.

A degree- $D$  **pseudodistribution** of solutions exists iff no degree- $D$  SoS refutation exists  
In other words, SoS thinks the system is feasible.

Dual view: A series of convex relaxations to a polynomial program, parameterized by  $D$ , called degree of SoS.

Obtains state-of-the-art algorithms for many problems such as Max  $k$ -CSPs, Tensor PCA, etc.

**Question restated:** How do SoS relaxations for the Sherrington-Kirkpatrick problem perform?

# SoS lower bounds for PAP

**Main theorem restated:** For  $m \leq n^{1.5 - \epsilon}$ , w.h.p. over  $d_1, \dots, d_m \sim N(0, I_n)$ , degree- $n^\delta$  SoS thinks this system of equations is feasible.

More concretely, given input  $d_1, \dots, d_m$  in  $\mathbb{R}^n$ , specify  $\tilde{\mathbb{E}}[v_1]$ ,  $\tilde{\mathbb{E}}[v_2 v_4]$ , etc., such that

1.  $\tilde{\mathbb{E}}[1] = 1$  and  $\tilde{\mathbb{E}}[\langle v, d_u \rangle^2] = \tilde{\mathbb{E}}[1]$  for all  $u = 1, \dots, m$  and other constraints.
2.  $\tilde{\mathbb{E}}[g^2] \geq 0$  for all polynomials of degree at most  $D/2$ . **[Positivity condition]**

**Moment matrix:** Matrix  $\mathcal{M}$  with rows, columns indexed by subsets of  $\{1, \dots, n\}$  such that

$$\mathcal{M}[I, J] = \tilde{\mathbb{E}}[v^I \cdot v^J] \text{ for subsets } I, J \text{ of } \{1, \dots, n\} \text{ of size } \leq D/2$$

**Positivity condition:**  $\mathcal{M} \succeq 0$

	$\emptyset$	$v_k$	$v_k v_l$
$\emptyset$			
$v_i$			
$v_i v_j$			$\tilde{\mathbb{E}}[v^I v^J]$



# Pseudocalibration

The **pseudocalibration** heuristic introduced by [BHKMP '16] gives a candidate  $\tilde{\mathbb{E}}$  that satisfies the constraints (approximately).

$$\text{For any subset } S \subseteq [n], \tilde{\mathbb{E}}[v^S] = \sum_{\alpha \in \mathbb{N}^{m \times n}} c_\alpha h_\alpha(d_1, \dots, d_m)$$

$h_\alpha$  - basis of Hermite polynomials

$c_\alpha$  - real coefficients

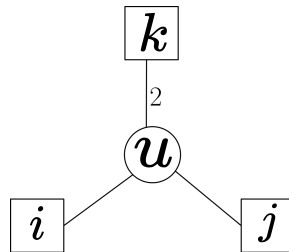
$c_\alpha = 0$  unless  $\alpha$  satisfies simple parity conditions, in which case,  $c_\alpha \approx n^{-|\alpha|/2}$

Main difficulty: Checking the positivity condition  $\mathcal{M} \geq 0$

# Graph Matrices

Hermite polynomials on  $\{d_u\}$  are in 1-to-1 correspondence with  $\mathbb{N}$ -edge-labeled graphs on  $[m] \cup [n]$

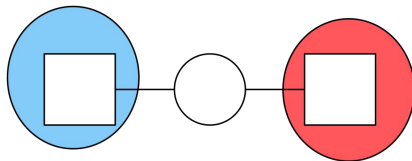
$$h_1(d_{u,i})h_1(d_{u,j})h_2(d_{u,k})$$



If we look at  $\mathcal{M}$ , the same “Fourier shape” appears in lots of different entries  $\mathcal{M}[I, J]$ . And in fact the coefficient  $c_\alpha$  only depends on (1) the shape and (2) the sets  $I, J$

Collect all such entries together into a matrix  $M_\beta$  encoded by a graph  $\beta$

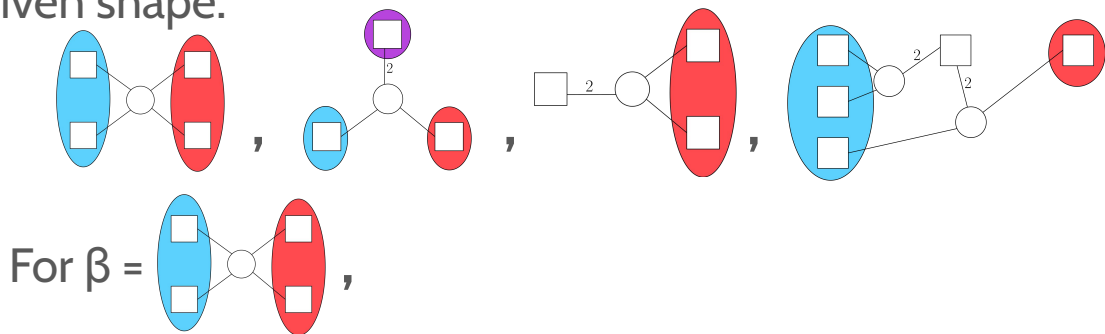
This will be a **graph matrix**.



# Graph Matrices

**Graph matrix [BHKKMP '16, AMP '20]:** A graph matrix  $M_\beta$  is defined by a **shape**  $\beta =$  bipartite  $\mathbb{N}$ -edge-labeled graph with special subsets of vertices  $U, V$ . The matrix sums up all Hermite characters with the given shape.

$\emptyset$	$k$	$k, l$
$\emptyset$		
$i$	<b>0</b>	<b>0</b>
$i, j$	<b>0</b>	



$$M_\beta[\{i, j\}, \{k, l\}] = \sum_{u=1}^m h_1(d_{u,i}) h_1(d_{u,j}) h_1(d_{u,k}) h_1(d_{u,l})$$

$$\mathcal{M} = \sum_{\substack{\text{shapes } \beta: \\ \beta \text{ satisfies parities constraints}}} \prod_{\circ \in V(\beta)} h_{\deg(\circ)}(1) \cdot \frac{M_\beta}{n^{|E(\beta)|/2}}$$

# Graph Matrices

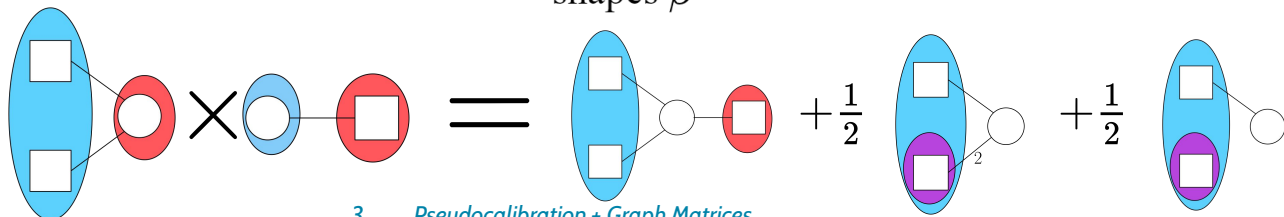
Theorem [AMP '20]: W.h.p. for all shapes  $\beta$ : let  $S$  be a **minimum-weight vertex separator** of left and right sides of  $\beta$ ,

$$\|M_\beta\| \leq \tilde{O}(\sqrt{m}^{\#\text{O not in } S} \cdot \sqrt{n}^{\#\square \text{ not in } S})$$

In fact, this is the only property of the random input that we need

Graph matrices are a “functional matrix algebra”:

$$M_\beta \cdot M_\gamma = \sum_{\text{shapes } \beta'} c_{\beta'} M_{\beta'}$$



Norm bounds:

$$\|M_\beta\| \leq$$

$$\tilde{O}(\sqrt{m}^{\#\circ \text{ not in } S} \cdot \sqrt{n}^{\#\square \text{ not in } S})$$

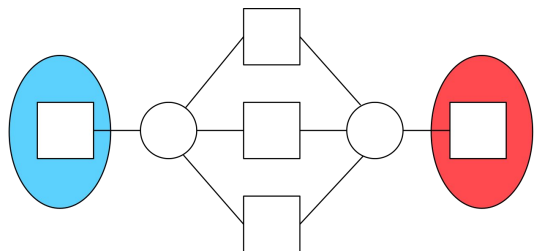
where  $S$  is a min-weight vertex separator of  $\beta$

## PSD-ness proof sketch

$$\mathcal{M} = \sum_{\text{shapes } \beta} c_\beta M_\beta = \sum_{\substack{\text{shapes } \beta: \\ \beta \text{ satisfies parity constraints}}} \prod_{\circ \in V(\beta)} h_{\deg(\circ)}(1) \cdot \frac{M_\beta}{n^{|E(\beta)|/2}}$$

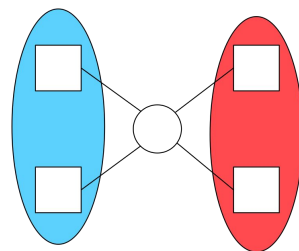
Naive idea: show that  $\sum_{\beta \neq \text{identity}} |c_\beta| \|M_\beta\| = o(1)$

Some terms do have  $o(1)$  norm...



$$\begin{aligned} |c_\beta| \cdot \|M_\beta\| &= O(1/n^4) \cdot \tilde{O}(n^2 m) \\ &= \tilde{O}(m/n^2) = o(1) \end{aligned}$$

...but others do not



$$\begin{aligned} |c_\beta| \cdot \|M_\beta\| &= O(1/n^2) \cdot \tilde{O}(n^2) \\ &= \tilde{\Omega}(1) \end{aligned}$$

# PSD-ness proof sketch

Non-identity terms with  $\Omega(1)$  norm *must* exist because  $\mathcal{M}$  has a nontrivial null space  
 Null space is induced by constraints " $\langle v, d_u \rangle^2 = 1$ "

$$\mathcal{M} \times \left( \left( \begin{array}{c} \text{[blue oval with 2 squares]} \cdot \mathbf{u} + \frac{1}{n} \text{[square]} \cdot \mathbf{u} \end{array} \right) \right)$$

$$\times \left( \begin{array}{c} \text{[vector u]} \\ \text{[vector u]} \end{array} \right)$$

$$\begin{array}{c} \emptyset \\ v_i \\ v_i v_j \end{array} \begin{array}{c|c} \emptyset & v_k \quad v_k v_l \\ \hline & \end{array}$$

$$\begin{array}{c} \emptyset \\ k \\ k, l \end{array} \begin{array}{c} 0 \\ 0 \\ h_1(d_{u,k})h_1(d_{u,l}) \end{array}$$

$$+ \frac{1}{n} \begin{array}{c} \emptyset \\ k \\ k, l \end{array} \begin{array}{c} \sum_k h_2(d_{u,k}) \\ 0 \\ 0 \end{array}$$

After some simplifications, top row =  $\tilde{\mathbb{E}}[\langle v, d_u \rangle^2 - 1]$

# Certification

**Certification problem:** Given  $W \sim \text{GOE}(n)$ , output an upper bound on  $\max (x^T W x)$  over  $x \in \{+1, -1\}^n$  that gets close to the true optimum whp.

Spectral certificate: Just output the maximum eigenvalue of  $W$ , which is  $(2 + o(1)) n^{1.5}$  whp.

Can other certificates get closer to the true optimum, which is  $\approx 1.52 n^{1.5}$  whp?

In particular, the **Sum-of-Squares hierarchy** offers a natural series of convex relaxations for this problem.

# Results on SoS for SK

Degree-2 SoS has value  $(2 - o(1)) n^{1.5}$

[Montanari-Sen '16]

Degree-4 SoS has value  $(2 - o(1))n^{1.5}$

[Mohanty-Raghavendra-Xu '20; Kunisky-Bandeira '19]

Degree-6 SoS has value  $(2 - o(1)) n^{1.5}$

[Kunisky '20]

**This work:** For some constant  $\delta > 0$ , whp degree- $n^\delta$  SoS still has value  $(2 - o(1)) n^{1.5}$

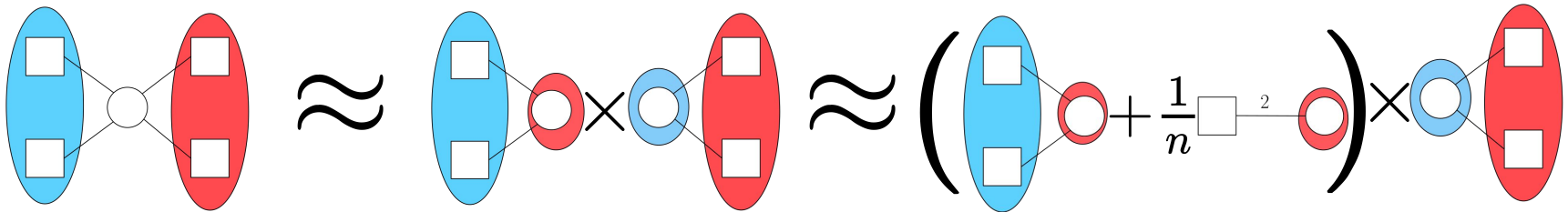
Rules out degree- $O(1)$  SoS = polynomial time SoS



# PSD-ness proof sketch

It suffices to prove that  $\mathcal{M}$  is PSD on  $\text{Null}(\mathcal{M})^\perp$

Spiders approximately factor into a matrix with columns from  $\text{Null}(\mathcal{M})!$



Thus spiders are approximately zero on  $\text{Null}(\mathcal{M})^\perp$

$$x^\top \mathcal{M} x \approx x^\top (\mathcal{M} - \text{spider}) x \text{ for all } x \in \text{Null}(\mathcal{M})^\perp$$

Theorem: There is  $\mathcal{M}'$  such that  $x^\top \mathcal{M} x = x^\top \mathcal{M}' x$  for all  $x \in \text{Null}(\mathcal{M})^\perp$  and all eigenvalues of  $\mathcal{M}'$  are  $1 \pm o(1)$

# PSD-ness proof sketch

Due to  $\approx$ , killing a spider introduces smaller terms into the moment matrix

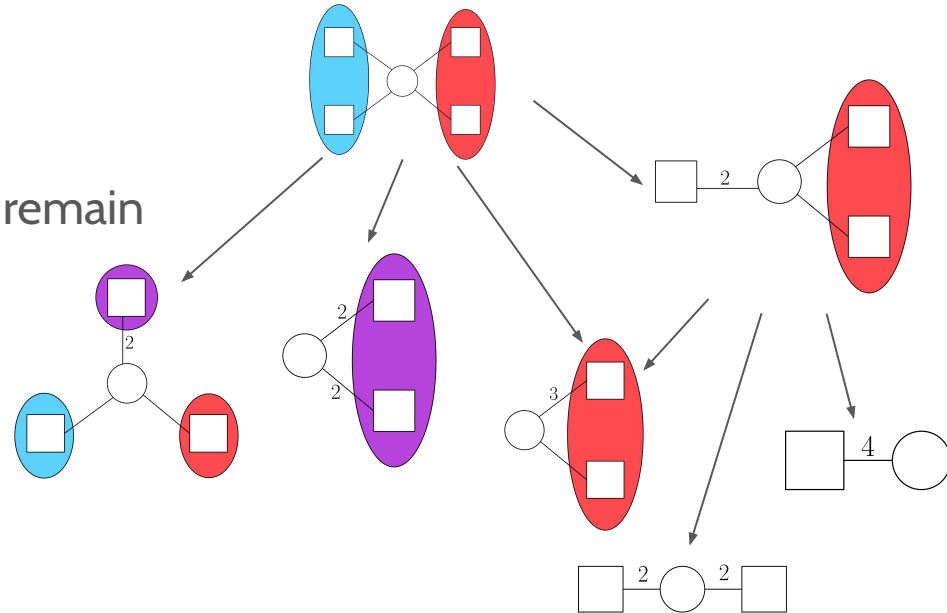
$$x^T \mathcal{M} x = x^T (\mathcal{M} - \text{spider} + \text{intersection terms}) x \text{ for all } x \in \text{Null}(\mathcal{M})^\perp$$

Some of these may be smaller spiders!

Recursively kill these until only non-spiders remain

Theorem: For  $c'_\beta$  the new coefficients on non-spiders,

$$\sum_{\substack{\text{non-spider,} \\ \text{non-identity } \beta}} |c'_\beta| \cdot \|M_\beta\| = o(1)$$



A wooden mallet with a rounded head and a handle, and a rectangular wooden block, both resting on a grey surface. The mallet is on the left, and the block is on the right. The text is overlaid on the image.

## Open Problems

Interpret Montanari's algorithm as rounding SoS?

Average case sparse MaxCut

Improve PAP assumption  $m \leq n^{1.5}$  to  $m \leq n^2$

Improve SoS degree to  $n/\log n$