A Ihara-Bass Formula for Non-Boolean Matrices and Strong Refutations of Random CSPs

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## SAT solvers and average-case complexity

- SAT solvers work well on very large-scale instances coming from program verification, VLSI, etc
- For most applications, it is important to be able to *certify unsatisfiability of unsatisfiable formulas*
- Average-case complexity does not provide an explanation for feasibility of solving SAT in practice

# Refuting random k-SAT

- Pick a random k-SAT formula with n variables, m clauses
- If  $m > c_k n$ , formula is unsatisfiable whp
- Seems hard to find proof of unsatisfiability when m is, say, O(n log n)
- Feige proposed it as a complexity assumption
- Problem becomes easier for larger m. When is it poly-time?

# Refuting random k-SAT

- Easy to see: if  $m > c_k n^{k-1}$  there is, whp, an efficiently constructable refutation by *tree-like resolution* 
  - More work: same if  $m > c_k n^{k-1} / \log n$
- By spectral methods: whp efficiently constructable refutation if  $m > c_k n^{\lceil k/2 \rceil + o(1)}$ [Goerdt, Krivelevich 2001]

• By more sophisticated spectral methods: whp strong refutation if

$$\begin{split} m &> \frac{1}{\epsilon^2} c_k n^{\frac{k}{2}} \text{ if k is even} \\ m &> \frac{1}{\epsilon^2} c_k n^{\frac{k}{2}} \text{ polylog } n \text{ if k is odd} \end{split}$$

[Friedman, Goerdt 2001] . . . [Allen, O'Donnell, Witmer 2015]

# Our result

• Efficiently computable strong refutation if

$$m > \frac{1}{\epsilon^2} c_k n^{k/2}$$
 if k is even  
 $m > \frac{1}{\epsilon^2} c_k n^{k/2}$  polylog  $n$  if k is odd  
[Friedman, Goerdt 2001] . . . [Allen, O'Donnell, Witmer 2015]

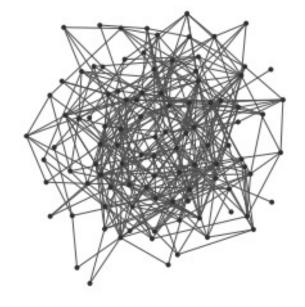
• Our result:

$$m > \frac{1}{\epsilon^2} c_k n^{k/2}$$
 even if k odd

• Sample 
$$G \sim \mathcal{G}_{n,\frac{d}{n}}$$

• Whp, max cut  $\leq \frac{1}{2} + \frac{c}{\sqrt{d}}$ 

Proof: Chernoff bounds + union bound



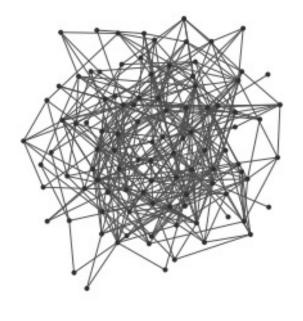
• Whp, there is efficiently computable proof that max cut  $\leq \frac{1}{2} + \frac{c'}{\sqrt{d}}$ 

Proof: [Feige, Ofek 2005] or Grothendieck's inequality + Chernoff bounds



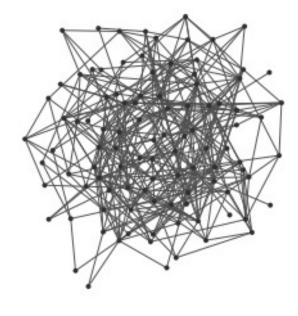
- Sample *G* so that each edge has probability  $\frac{a}{n}$  and edges are polylogn-wise independent
- Can we certify whp that max cut  $\leq \frac{1}{2} + \frac{c}{\sqrt{d}}$ ?
- Is it even true whp?

(If distribution has entropy o(n), we cannot take union bounds!)



- Sample G so that each edge has probability  $\frac{d(n)}{n}$ and edges are polylogn-wise independent
- By trace methods, whp non-trivial eigenvalues of adjacency matrix  $\leq \sqrt{d(n) \log n}$  in magnitude
- Trace calculation needs only O(log n)-wise independence of edges

• Whp, max cut is certifiably 
$$\frac{1}{2} + c \frac{\sqrt{\log n}}{\sqrt{d(n)}}$$



# Strong refutations of k-SAT

- Refuting random 4-SAT formula with n variables, m clauses reduces to a problem similar to
  - Find a certificate that a given random graph with  $n^2$  vertices and m independent random edges has a max cut  $\leq \frac{1}{2} + \epsilon$

- Refuting random 3-SAT formula with n variables, m clauses reduces to a problem similar to
  - Find a certificate that a given random graph with  $n^2$  vertices and  $\frac{m^2}{n}$  random-but-correlated edges has a max cut  $\leq \frac{1}{2} + \epsilon$

# Strong refutations of k-SAT, k even

- Refuting random 4-SAT formula with n variables, m clauses reduces to a problem similar to
  - Find a certificate that a given random graph with  $n^2$  vertices and m independent random edges has a max cut  $\leq \frac{1}{2} + \epsilon$
- Refuting random k-SAT (k even) formula with n variables, m clauses reduces to a problem similar to
  - Find a certificate that a given random graph with  $n^{k/2}$  vertices and m independent random edges has a max cut  $\leq \frac{1}{2} + \epsilon$
  - Can do if  $m > \frac{c}{\varepsilon^2} n^{k/2}$

# Strong refutations of k-SAT

• Refuting random 3-SAT formula with *n* variables, *m* clauses reduces to a problem similar to

• Find a certificate that a given random graph with  $n^2$  vertices and  $\frac{m^2}{n}$  random-but-correlated edges has a max cut  $\leq \frac{1}{2} + \epsilon$ 

- Refuting random k-SAT formula (k odd) with n variables, m clauses reduces to a problem similar to
  - Find a certificate that a given random graph with  $n^{(k+1)/2}$  vertices and  $\frac{m^2}{n^{(k-1)/2}}$  random-but-correlated edges has a max cut  $\leq \frac{1}{2} + \epsilon$

# Strong refutations of k-SAT

- Refuting random 3-SAT formula with n variables, m clauses reduces to a problem similar to
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• Can do if 
$$m > \frac{c}{\varepsilon^2} n^{k/2} \operatorname{polylog} n$$

# Strong refutations of random 4-SAT

- Enough to provide strong refutation of random 4-XOR [Feige 2002] +...
- To find strong refutation of random 4-XOR problem, we can apply trivial (and seemingly not useful) reduction to 2-XOR:

Max # satisfiable constraints in

$$x_{1}x_{3}x_{5}x_{7} = 1 \leq x_{2}x_{3}x_{6}x_{7} = -1 \\ x_{1}x_{4}x_{5}x_{7} = 1 \\ \dots$$

 $x_1,\ldots,x_n \ \in \{1,-1\}^n$ 

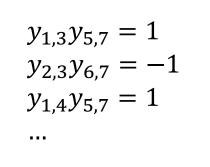
Max # satisfiable constraints in

 $y_{1,3}y_{5,7} = 1$   $y_{2,3}y_{6,7} = -1$   $y_{1,4}y_{5,7} = 1$ ...

$$y_{1,1}, \dots, y_{n,n} \in \{1, -1\}^{n^2}$$

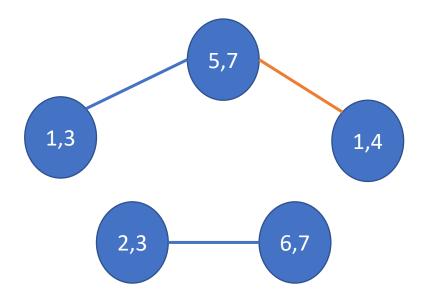
# **Reduction to random 2-XOR**

- Strong refutation of random 4-XOR with n variables, m constraints reduces to proving that the optimum is small in
- A random 2-XOR problem with  $n^2$  clauses, m constraints
- Equivalently, a random correlation clustering problem in a graph with  $n^2$  vertices, m random edges



Max # satisfiable constraints in

$$y_{1,1},\ldots,y_{n,n} \in \{1,-1\}^{n^2}$$



## **Reduction to random 2-XOR**

 Strong refutation of random 4-XOR with n variables, m constraints reduces to proving that

$$\max_{y_{1,1},\dots,y_{n,n} \in \{-1,1\}^{n^2}} y^T M y \le \varepsilon m$$

where

$$M_{i,j,h,k} = \begin{cases} 1 \text{ if } x_i x_j x_h x_k = 1 \text{ is a constraint} \\ -1 \text{ if } x_i x_j x_h x_k = -1 \text{ is a constraint} \\ 0 \text{ otherwise} \end{cases}$$

### **Reduction to random 2-XOR**

• Want to prove that

$$\max_{y_{1,1},\ldots,y_{n,n} \in \{-1,1\}^{n^2}} y^T M y \leq \varepsilon m$$

Proof:

$$\max_{\substack{y_{1,1},\dots,y_{n,n} \in \{-1,1\}^{n^2}}} y^T M y$$

$$\leq \max_{\substack{y_{1,1},\dots,y_{n,n} \in \{-1,1\}^{n^2}}} y^T M z$$

$$= \frac{z_{1,1},\dots,z_{n,n} \in \{-1,1\}^{n^2}}{||M||_{\infty \to 1}}$$

$$\leq \sqrt{mn^2} \text{ whp}$$

# Strong refutations of random 4-SAT

- Enough to provide strong refutation of random 4-XOR [Feige 2002] +...
- Can write random 4-XOR formula with n variables and m constraints as

• 
$$\max_{x_1...x_n \in \{-1,1\}^n} \frac{m}{2} + \frac{1}{2} \sum_{i,j,k,h} b_{i,j,k,h} x_i x_j x_k x_h$$

• Where m of the  $b_{i,j,k,h}$  are non-zero, and each is equally likely to be  $\pm 1$ 

 Strong refutation of random 3-XOR with n variables, m constraints means proving that

$$\max_{x_1,\dots,x_n\in\{-1,1\}^n}\sum T_{i,j,k}x_ix_jx_k\leq\varepsilon m$$

where

$$T_{i,j,k} = \begin{cases} 1 \text{ if } x_i x_j x_k = 1 \text{ is a constraint} \\ -1 \text{ if } x_i x_j x_k = -1 \text{ is a constraint} \\ 0 \text{ otherwise} \end{cases}$$

$$\begin{split} & \max_{x_{1},...,x_{n} \in \{-1,1\}^{n}} \sum T_{i,j,k} x_{i} x_{j} x_{k} \\ & \leq \max_{x_{1},...,x_{n} \in \{-1,1\}^{n}} \sqrt{\sum_{i} x_{i}^{2}} \sqrt{\sum_{i} \left(\sum_{j,k} T_{i,j,k} x_{j} x_{k}\right)^{2}} \\ & = \sqrt{n} \cdot \max_{x_{1},...,x_{n} \in \{-1,1\}^{n}} \sqrt{\sum_{i,a,b,c,d} T_{i,a,b} T_{i,c,d} x_{a} x_{b} x_{c} x_{d}} \end{split}$$

Enough to prove  

$$\max_{\substack{x_1, \dots, x_n \in \{-1,1\}^n \\ i, a, b, c, d}} \sum_{\substack{T_{i,a,b} T_{i,c,d} x_a x_b x_c x_d}} x_b x_c x_d \leq \frac{\varepsilon^2 m}{n}$$

$$\max_{x_{1},...,x_{n}\in\{-1,1\}^{n}}\sum_{i,a,b,c,d}T_{i,a,b}T_{i,c,d}x_{a}x_{b}x_{c}x_{d}$$

$$=\max_{x_{1},...,x_{n}\in\{-1,1\}^{n}}\sum_{a,b,c,d}x_{a}x_{c}\left(\sum_{i}T_{i,a,b}T_{i,c,d}\right)x_{b}x_{d}$$

$$\leq\max_{y_{1,1},...,y_{n,n}\in\{-1,1\}^{n^{2}}}y^{T}My$$

where  $M_{a,c,b,d} = \sum_{i} T_{i,a,b} T_{i,c,d}$ 

$$\max_{y_{1,1},...,y_{n,n} \in \{-1,1\}^{n^2}} y^T M y$$

where  $M_{a,c,b,d} = \sum_{i} T_{i,a,b} T_{i,c,d}$ 

*M* is an  $n^2 \times n^2$  matrix where we expect to see  $\approx \frac{m^2}{n}$  non-zero entries

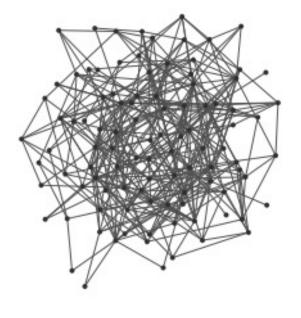
With trace methods, possible to prove spectral bounds sufficient for our goal when m is  $n^{1.5} \operatorname{poly} \log n$ 

[Allen, O'Donnell, Witmer 2015]



- Sample G so that each edge has probability  $\frac{a}{n}$  and edges are polylogn-wise independent
- Can we certify whp that max cut  $\leq \frac{1}{2} + \frac{c}{\sqrt{d}}$ ?
- Is it even true whp?

Yes, implicit in [Bordenave, Lelarge, Massoulié 2015] + [Fan, Montanari 2017]

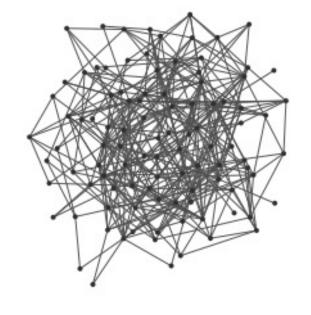


## Non-backtracking operator

- Given undirected graph G = (V, E)
- Non-backtracking operator *B* is a  $2|E| \times 2|E|$ Boolean 0/1 matrix such that

$$B_{(u,v),(v,z)} = 1$$
 iff

 $(u, v) \in E,$   $(v, z) \in E,$  $u \neq z$ 

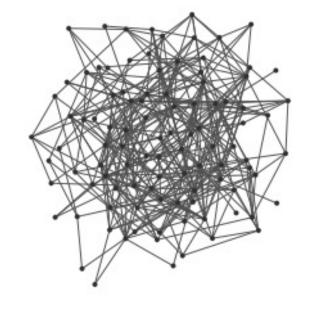


# Non-backtracking operator

• Sample 
$$G \sim \mathcal{G}_{n,\frac{d}{n}}$$

- Whp:
  - Largest real e-value of B is  $(1 + o(1)) \cdot d$
  - All others are  $\leq (1 + o(1)) \cdot \sqrt{d}$  in magnitude

[Bordenave, Lelarge, Massoulié 2015]



# Ihara-Bass formula

- If G = (V, E) is an undirected graph
- *A* is the adjacency matrix
- **D** is the diagonal matrix such that  $D_{v,v} = \text{degree}(v)$
- *B* is the non-backtracking operator

#### Then

$$\det(I - xB) = (1 - x^2)^{|E| - |V|} \det(I - xA + x^2(D - I))$$

#### Fan-Montanari

- If G = (V, E) is an undirected graph
- A is the adjacency matrix of G
- B is the non-backtracking operator of G
- $\lambda_{\min}$  is the smallest (most negative) real eigenvalue of *B*

Then

$$A \ge -|\lambda_{\min}| \cdot I - \frac{1}{|\lambda_{\min}|} \cdot (D - I)$$

- Sample  $G \sim \mathcal{G}_{n,\frac{d}{n}}$
- By combining [Bordenave, Lelarge, Massoulié 2015] + [Fan, Montanari 2017]:

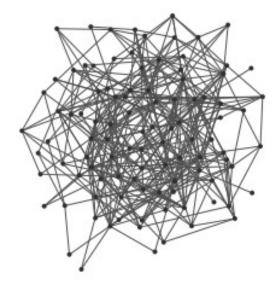
 $A \ge -(1+o(1)\sqrt{d} \cdot I + (1+o(1) \cdot D/\sqrt{d}))$ 

• Enough to imply:

 $\max \operatorname{cut} \le \frac{1}{2} + \frac{1 + o(1)}{\sqrt{d}}$ 

Goemans-Williamson relaxation can certify it

• [FM17] works for all graphs, [BLM15] works in random graphs with polylogn-wise independent edges and constant *d* 



# Our technical contributions

- Give a definition of non-backtracking operator B associated to an arbitrary symmetric matrix A (with arbitrary positive and negative entries)
- Prove a Ihara-Bass formula
- Prove a Fan-Montanari type result
- Prove a Bordenave-Leland-Massoulié type result for the matrices coming from the 3-XOR reduction

# Our Ihara-Bass type formula

• We give a definition of a non-backtracking operator B associated to an arbitrary symmetric  $n \times n$  matrix A with m non-zero entries (which can be arbitrary positive and negative numbers) such that

$$\det(I - xB + xL - xJ) = (1 - x^2)^{\frac{m}{2} - n} \cdot \det(I - xA + x^2(D - I))$$

- Where *D* is the analog of the matrix of degrees and *L*, *J* are matrices associated to *A* that are equal if *A* is Boolean
- A Fan-Montanari type result can be proved from the above formula

### Our Bordenave-Leland-Massoulié type bound

- Take a random 3-XOR formula with n variables and m constraints
- Reduce bounding the max 3-XOR problem to a quadratic optimization problem defined by a  $n^2 \times n^2$  matrix A with  $\frac{m^2}{n}$  non-zero entries
- The non-backtracking operator *B* of *A* satisfies whp  $||B - L + J|| \le O\left(\frac{m}{n^{1.5}}\right)$
- There is a certificate that in the 3-XOR, at most

$$\frac{m}{2} + c\sqrt{n^{1.5} \cdot m}$$
  
constraints can be simultaneously satisfied

### Conclusions

- We give an algorithm that, whp, finds strong refutations of random 3XOR and random 3SAT problems where the number of constraints/clauses is order of  $n^{1.5}$
- Breaks long-standing barrier
- Shows that one can analyze random matrices that have an expected constant number of non-zero entries per row, and such that the entries are nonindependent
- Generalize the theory of non-backtracking operators to arbitrary matrices (graphs with arbitrary positive and negative weights) in a way that recovers both spectral bounds and algorithmic applications of the boolean/unweighted case