

# Extremal cuts of sparse random graph

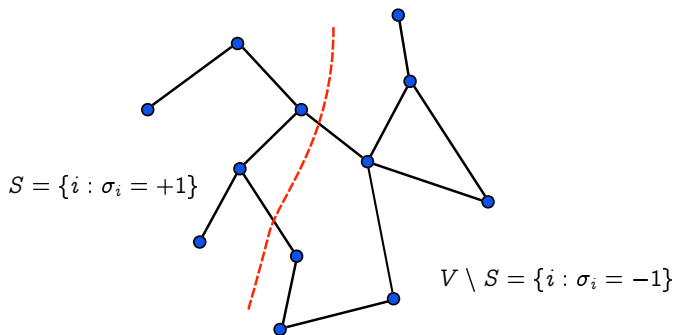
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## Cuts in graph $G = (V, E)$



$$\text{cut}_G(S) \equiv \left| \{(i, j) \in E : i \in S, j \in V \setminus S\} \right|$$

$$2\text{cut}_G(S) - |E| = \mathcal{H}_G(\sigma) \quad \text{for AF-Ising on } G$$

# Extremal cuts

## Minimum bisection

$$\text{mcut}(G) = \min \left\{ \text{cut}_G(S) : S \subseteq V, |S| = |V|/2 \right\}$$

## Maximum bisection

$$\text{MCUT}(G) = \max \left\{ \text{cut}_G(S) : S \subseteq V, |S| = |V|/2 \right\}$$

## Maximum Cut

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# Interesting for many reasons

- ▶ Clustering similarity matrices
- ▶ Graph layout (optimal  $V \mapsto \mathbb{N}$ )
- ▶ Isoperimetric number
- ▶ Community structure in social networks
- ▶ ...

## Complexity: hard to solve (worst case)

mcut }  
MCUT } are NP-hard  
MaxCut }

(hard to approximate to within  $1 + o(1)$  factor:  
SDP [Goemans,Williamson '95]; hardness [Trevisan,Sorkin,Sudan,Williamson '00]  
mcut [Feige,Krauthgamer '00]  
(... very rich theory)

Typical complexity (average graph)?

# Random graph models

Erdős-Renyi Random graph  $G = (V, E) \sim \mathcal{G}(n, p)$

- ▶  $|V| = n$  vertices
- ▶ Edges independently present with probability  $p$   
(or uniform among graphs of  $m = \binom{n}{2}p$  edges)
- ▶ Average degree  $\gamma = np$

Random regular graph  $G = (V, E) \sim \mathcal{G}^{\text{reg}}(n, \gamma)$

- ▶  $|V| = n$  vertices
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## Extremal cuts of random graphs: long history

- ▶ Bollobas '88 ( $\mathcal{G}^{\text{reg}}$ , mcut, concentration bound)
- ▶ Alon '97 ( $\mathcal{G}^{\text{reg}}$ , mcut, algorithmic bound)
- ▶ Coppersmith, Gamarnik, Hajiaghayi, Sorkin '04 ( $\mathcal{G}$ , MaxCut transition  $\gamma = 0.5$ )
- ▶ Díaz, Serna, Wormald, '07 ( $\mathcal{G}^{\text{reg}}$ , MCUT, algorithmic bound)
- ▶ Daudé, Martínez, Rasendrasahina, Ravelomanana '12 (MaxCut, scaling-window)
- ▶ Gamarnik, Li '14 ( $\mathcal{G}$ , MaxCut, improved concentration)
- ▶ ...

Typical result (sparse case):

If  $G \sim \mathcal{G}(n, \gamma/n)$  then, with high probability

$$\frac{n\gamma}{4} + C_1 n \sqrt{\gamma} \leq \text{MCUT}(G) \leq \frac{n\gamma}{4} + C_2 n \sqrt{\gamma}$$

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# A classical argument

(Bollobas '88)

$|V| = n, |E| = n\gamma/2.$       Fix  $S \subseteq V, |S| = n/2$

- ▶ Each edge is cut with probability  $1/2$

$$\mathbb{E}[\text{cut}(S)] = \frac{|E|}{2} = \frac{n\gamma}{4}.$$

(also factor 0.5 approximation of MaxCut)

- ▶ Azuma-Hoeffding argument

$$\mathbb{P}\left\{|\text{cut}(S) - \mathbb{E}\text{cut}(S)| \geq \Delta\right\} \leq 2 \exp\left(-\frac{\Delta^2}{4n\gamma}\right)$$

(simpler: Binomial( $\frac{n^2}{4}, \frac{\gamma}{n}$ ) in  $\mathcal{G}(n, \gamma/n)$ )

- ▶ Union bound

$$\mathbb{P}\left\{\max_{S, |S|=n/2} \left|\text{cut}(S) - \frac{n\gamma}{4}\right| \geq \delta n \sqrt{\gamma}\right\} \leq 2^{n+1} e^{-n\delta^2/4}.$$

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So, w.h.p.

$$\frac{n\gamma}{4} - C_1 n\sqrt{\gamma} \leq \text{mcut}(G) \leq \text{MCUT}(G) \leq \frac{n\gamma}{4} + C_2 n\sqrt{\gamma}$$

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# Extremal cuts and Ising spins

**Ising Hamiltonian:**  $\mathcal{H}_G(\sigma) = \frac{1}{\sqrt{\gamma}} \sum_{(i,j) \in E} \sigma_i \sigma_j : \{-1, +1\}^n \mapsto \mathbb{R}.$

$$\begin{aligned} \text{MaxCut}(G) &= \max_{\sigma} \sum_{(i,j) \in E} \left( \frac{1 - \sigma_i \sigma_j}{2} \right) \\ &= \frac{1}{2} |E| - \frac{1}{2} \min_{\sigma} \sum_{(i,j) \in E} \sigma_i \sigma_j = \frac{n\gamma}{4} - \sqrt{\frac{\gamma}{4}} \min_{\sigma} \{\mathcal{H}_G(\sigma)\} \end{aligned}$$

$$(S = \{i : \sigma_i = +1\}, \quad |E| = \frac{n\gamma}{2}).$$

$$|S| = \frac{n}{2} \iff \Omega_n = \left\{ \sigma \in \{-1, +1\}^n : \sum_{i=1}^n \sigma_i = 0 \right\}$$

$$\text{MCUT}(G) = \frac{n\gamma}{4} - \sqrt{\frac{\gamma}{4}} \min_{\sigma \in \Omega_n} \{\mathcal{H}_G(\sigma)\},$$

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# Statistical physics: Ising measures

Ising measures (on balanced configurations):

$$\mathcal{H}_G(\sigma) = \frac{1}{\sqrt{\gamma}} \sum_{(i,j) \in E} \sigma_i \sigma_j \quad (\sigma \in \Omega_n),$$
$$\mu_{n,\beta}(\sigma) \equiv \frac{1}{Z_n(\beta)} e^{\beta \mathcal{H}_G(\sigma)}, \quad Z_n(\beta) \equiv \sum_{\tilde{\sigma} \in \Omega_n} e^{\beta \mathcal{H}_G(\tilde{\sigma})}.$$

(ferromagnetic iff  $\beta \geq 0$ ).

Ground state energy:

$$\lim_{\beta \rightarrow +\infty} \frac{1}{\beta} \log Z_n(\beta) = \max_{\sigma \in \Omega_n} \{ \mathcal{H}_G(\sigma) \} \implies \text{mcut}(G),$$
$$\lim_{\beta \rightarrow -\infty} \frac{1}{\beta} \log Z_n(\beta) = \min_{\sigma \in \Omega_n} \{ \mathcal{H}_G(\sigma) \} \implies \text{MCUT}(G).$$

# Insights from statistical physics

## Conjectures

- ▶ Fu, Anderson, '86
- ▶ Mézard, Parisi, '01
- ▶ [Zdéborova, Boettcher '10] For  $G_n \sim \mathcal{G}^{\text{reg}}(n, \gamma)$ , w.h.p.

$$\text{MCUT}(G_n) = \text{MaxCut}(G_n) + o(n) = |E_n| - \text{mcut}(G_n) + o(n).$$

## Theorem (Bayati, Gamarnik, Tetali, '09)

For  $G_n \sim \mathcal{G}(n, \gamma/n)$ , or  $G_n \in \mathcal{G}^{\text{reg}}(n, \gamma)$  w.h.p.

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \text{MaxCut}(G_n) = \limsup_{n \rightarrow \infty} \frac{1}{n} \text{MaxCut}(G_n)$$

Sub-additivity: no limit value.

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## A new result

Theorem (Dembo, Montanari, Sen '15)

For  $G_n \sim \mathcal{G}^{\text{reg}}(n, \gamma)$ ,  $G_n \sim \mathcal{G}(n, \gamma/n)$  w.h.p.

$$\begin{aligned}\frac{1}{n} \text{mcut}(G_n) &= \frac{\gamma}{4} - P_* \sqrt{\frac{\gamma}{4}} - o_\gamma(\sqrt{\gamma}) + o_n(1), \\ \frac{1}{n} \text{MCUT}(G_n) &= \frac{\gamma}{4} + P_* \sqrt{\frac{\gamma}{4}} + o_\gamma(\sqrt{\gamma}) + o_n(1), \\ \frac{1}{n} \text{MaxCut}(G_n) &= \frac{\gamma}{4} + P_* \sqrt{\frac{\gamma}{4}} + o_\gamma(\sqrt{\gamma}) + o_n(1),\end{aligned}$$

where  $P_* = \dots$  (wait a minute).

## Remarks

- ▶ The  $\sqrt{\gamma}$  term has a well defined coefficient.
- ▶ The coefficient has an ‘explicit’ formula.
- ▶ Same for 3 problems and 2 graph models.
- ▶ Partial confirmation of [Zdéborova,Boettcher '10] conjecture.
- ▶  $P_* \approx 0.7632$  (numerically; so  $P_*/\sqrt{2} \approx 0.5397$ ).

## What is $P_*$ ?

GOE random matrix:

$$J \in \mathbb{R}^{n \times n}, \quad J = J^T, \quad (J_{ij})_{i < j} \sim N(0, n^{-1}), \quad J_{ii} \sim N(0, 2n^{-1}).$$

Sherrington-Kirkpatrick spin-glass model

$$\mathcal{H}_{\text{SK}}(\sigma) \equiv \frac{1}{2} \sigma^T J \sigma.$$

Finally

$$P_* \equiv \lim_{n \rightarrow \infty} \frac{1}{n} \mathbb{E} \left[ \max_{\sigma \in \{-1, +1\}^n} \{\mathcal{H}_{\text{SK}}(\sigma)\} \right].$$

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- ▶ The limit exists (consequence of [Guerra, Toninelli '02])
- ▶ Given by '*Parisi's formula*' ([Talagrand '06])
- ▶ Clarifies why 'standard' combinatorial methods were unsuccessful.
- ▶ Set of near-extremal cuts predicted to have  $\infty$ -RSB structure.

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# Proof strategy

1. Proof for mcut, MCUT with  $G \sim \mathcal{G}(n, \gamma/n)$ :

Concentration, interpolation, 'smooth max'

2. Extend to  $G' \sim \mathcal{G}^{\text{reg}}(n, \gamma)$ : Couple  $G'$  and  $G \sim \mathcal{G}(n, \gamma'/n)$

(Tricky: need  $\gamma - \gamma' \gg \sqrt{\gamma}$  so  $|E_G \Delta E_{G'}| \gg n\sqrt{\gamma}$ )

3. Prove that  $\text{MaxCut}(G) - \text{MCUT}(G) = o(n\sqrt{\gamma})$

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mcut for  $\mathcal{G}(n, \gamma/n)$ :

Recall (our claim)

$$\text{mcut}(G) = \frac{n\gamma}{4} - \sqrt{\frac{\gamma}{4}} \max_{\sigma \in \Omega_n} \{\mathcal{H}_G(\sigma)\},$$

By concentration, suffices to show

$$\frac{1}{n} \mathbb{E} \left[ \max_{\sigma \in \Omega_n} \{\mathcal{H}_G(\sigma)\} \right] = P_* + o_\gamma(1) + o_n(1).$$

We know

$$\frac{1}{n} \mathbb{E} \left[ \max_{\sigma \in \{-1, +1\}^n} \{\mathcal{H}_{\text{SK}}(\sigma)\} \right] = P_* + o_n(1).$$

**Idea: Interpolation** ( $\sim$  smart path,  $\sim$  Lindeberg method, ...)

## Interpolation: Two steps

**Step 1:** For  $t \in [0, 1]$ ,  $G(t) = (V, E(t)) \sim \mathcal{G}(n, \gamma(1-t)/n)$

$$\mathcal{H}_t(\sigma; \gamma) = \frac{1}{\sqrt{\gamma}} \sum_{(i,j) \in E(t)} \sigma_i \sigma_j + \frac{\sqrt{t}}{2} \sum_{i,j=1}^n J_{ij} \sigma_i \sigma_j$$

$$\mathcal{H}_0 = \mathcal{H}_G \quad \mathcal{H}_1 = \mathcal{H}_{\text{SK}} \quad (\sigma \in \Omega_n).$$

$G(t)$  multi-graph of i.i.d. Poisson( $\gamma(1-t)/n$ ) multi-edges.

**Step 2:** From  $\max_{\sigma \in \Omega_n} \{\mathcal{H}_{\text{SK}}(\sigma)\}$  to  $\max_{\sigma \in \{-1, +1\}^n} \{\mathcal{H}_{\text{SK}}(\sigma)\}$ .

## Step 1: 'smooth max'

Free energy density (balanced configurations):

$$\phi_n(\beta; t, \gamma) \equiv \frac{1}{n} \mathbb{E} \left[ \log \left\{ \sum_{\sigma \in \Omega_n} e^{\beta \mathcal{H}_t(\sigma; \gamma)} \right\} \right]$$

$$\left| \frac{1}{\beta} \phi_n(\beta; t, \gamma) - \frac{1}{n} \mathbb{E} \left[ \max_{\sigma \in \Omega_n} \{ \mathcal{H}_t(\sigma; \gamma) \} \right] \right| \leq \frac{\log 2}{\beta}.$$

$\implies$  just find  $\beta(\gamma) \rightarrow \infty$  with

$$\sup_{n, t} \frac{1}{\beta} \left| \frac{\partial \phi_n}{\partial t}(\beta; t, \gamma) \right| \rightarrow 0.$$

### Lemma

$$\left| \frac{\partial \phi_n}{\partial t}(\beta; t, \gamma) \right| \leq C \left[ \frac{\beta^3}{\gamma^{1/2}} + \frac{\beta^4}{\gamma} \right].$$

**Proof:** Poisson & Gaussian calculus.

## Step 1: 'smooth max'

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### Lemma

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**Proof:** Poisson & Gaussian calculus.

## Extensions [Subhabrata Sen, in progress]

- ▶ Max  $q$ -cut in sparse random graphs  
 $\implies$  Potts spin-glass.
- ▶ Max-SAT for  $m = (1 + \delta)n$  random linear Eqn. over  $\{0, 1\}$   
(in UNSAT phase,  $p$  variables per Eqn.)  
 $\implies$   $p$ -spin model.



# Conclusion

- ▶ Extremal cuts in random graphs  
     $\implies$  balanced Ising spins  $\sigma_i \in \{-1, +1\}$
- ▶ Same for many other models, when  $\gamma \rightarrow \infty$ .
- ▶ Quite helpful, but much remains to understand!

[Zdéborova, Boettcher '10] For  $G_n \sim \mathcal{G}^{\text{reg}}(n, \gamma)$ , fixed  $\gamma \geq 3$ , w.h.p.

$$\text{MCUT}(G_n) = \text{MaxCut}(G_n) + o(n) = |E_n| - \text{mcut}(G_n) + o(n).$$

Thanks!