

# The Algorithmic Phase Transition of Random $k$ -SAT for Low Degree Polynomials

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Joint work with Guy Bresler



CanadAM 2023

# Random $k$ -SAT

$k$ -SAT formula: AND of  $m$  **clauses**, each an OR of  $k$  **literals**, e.g.

$$(x_1 \vee \bar{x}_3 \vee x_7) \wedge (\bar{x}_2 \vee x_3 \vee \bar{x}_5) \wedge (\bar{x}_1 \vee \bar{x}_2 \vee x_6)$$

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Parameters: fix large  $k$ ,  $\alpha = \alpha(k)$ . Then  $m, n \rightarrow \infty$  with  $m/n \rightarrow \alpha$

# Satisfiability and Algorithmic Thresholds

OPT: largest  $\alpha$  where solution **exists?** (w.h.p.)

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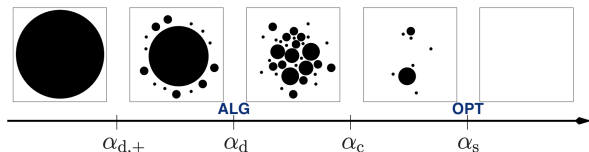
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Heuristic: **shattering** at  $\approx 2^k \log k/k$  obstructs algorithms (Achlioptas-Coja-Oghlan 08)



(Adapted from Krzakala-Montanari-Ricci-Tersenghi-Semerjian-Zdeborová 07)

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Clause Density	Algorithm(s)	Reference
$C2^k/k$	DPLL algorithms	[Achlioptas-Beame-Molloy 04]
$(1 + \varepsilon_k)2^k \log k/k$	Survey Propagation guided decimation	[Hetterich 16]
$(1 + \varepsilon_k)2^{k-1} \log^2 k/k$	Balanced sequential local algorithms (NAE- $k$ -SAT)	[Gamarnik-Sudan 17]
$C2^k \log^2 k/k$	Walksat	[Coja-Oghlan-Haqshenas-Hetterich 17]
$4.911 \cdot 2^k \log k/k$	Low degree polynomials	This work

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- i.e.  $\varepsilon_k n$  Hamming distance to assignment satisfying  $(1 - \varepsilon_k)m$  clauses



# Main Result

$$\kappa^* = \min_{\beta > 1} \frac{\beta}{1 - \beta e^{-(\beta-1)}} \approx 4.911$$

## Theorem (Bresler-H. 21)

*If  $\alpha > \kappa^* 2^k \log k/k$ , then no degree  $D = o(n/\log n)$  polynomial succeeds with probability  $1 - \exp(-\Omega(D \log n))$ .*

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For the same  $\alpha$ , no  $O(1)$ -local algorithm succeeds with probability  $\exp(-O(n^{1/3}))$ .

$O(1)$ -local algorithms also simulate FIX.

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- Max independent set (Gamarnik-Sudan 14, Rahman-Virág 17, Gamarnik-Jagannath-Wein 20, Wein 20)
- Random NAE- $k$ -SAT (Gamarnik-Sudan 17)
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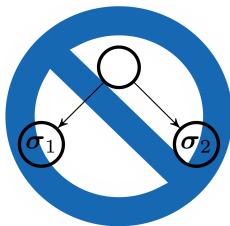
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**Overlap gap:** no solutions  $x, y$  have **medium** Hamming distance  $\in [\nu_1 n, \nu_2 n]$

- Intuition: solutions close together or far apart

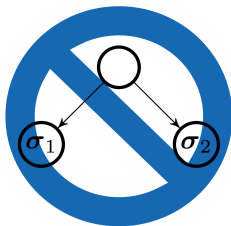
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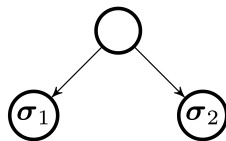
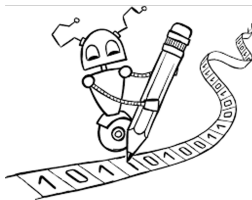


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Will show hardness for  $\alpha > \alpha_{\text{cl-ogp}} \equiv \frac{1}{2} \cdot 2^k \log 2 \approx \frac{1}{2} \text{OPT}$



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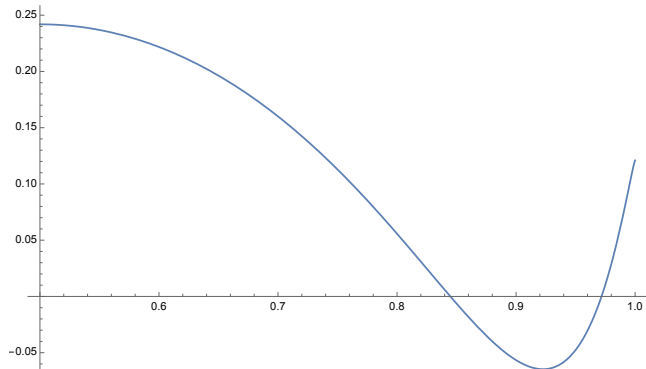
**Forbidden Structure:** two assignments  $y^{(1)}, y^{(2)} \in \{\text{T}, \text{F}\}^n$  such that

- Each  $y^{(i)}$  satisfies some  $\Phi^{(t_i)}$  (*not necessarily same  $t_i$* )
- $d_H(y^{(1)}, y^{(2)}) \in [\nu_1 n, \nu_2 n]$

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This structure doesn't occur w.h.p. by 1st moment calculation

$$\frac{1}{n} \log \mathbb{E} \# (\text{pairs of solutions } x, y \text{ with } d_H(x, y) \approx tn)$$



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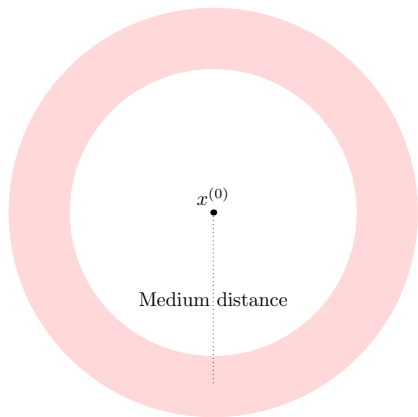
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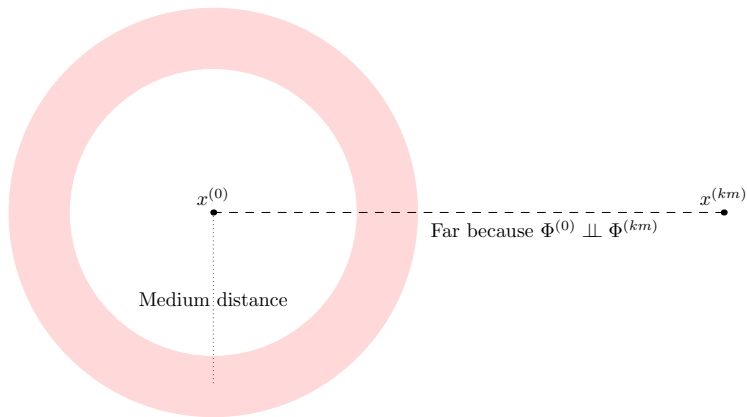
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Close by LDP stability

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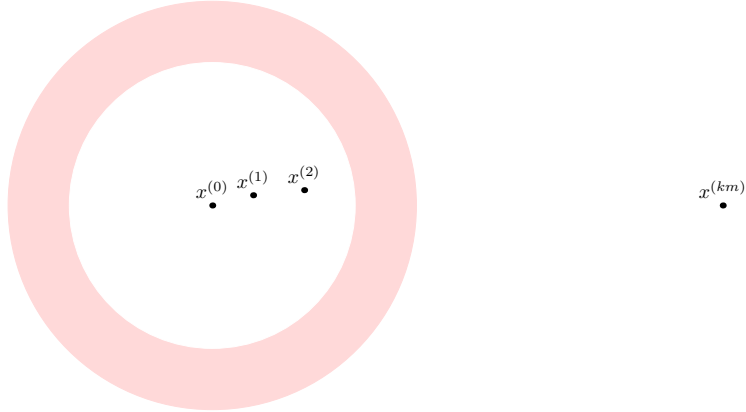


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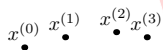


$x^{(0)}$   $x^{(1)}$   $x^{(2)}$

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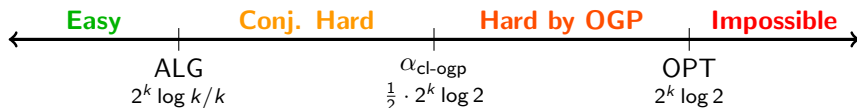
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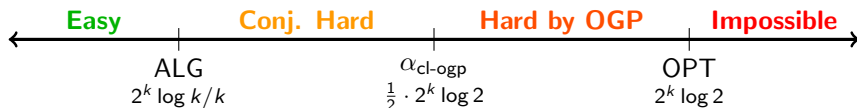
Contradiction  $\Rightarrow \mathcal{A}$  cannot succeed.

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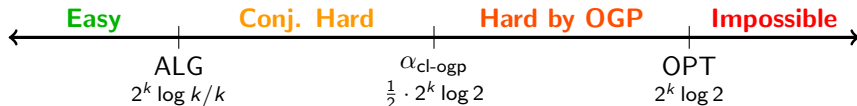
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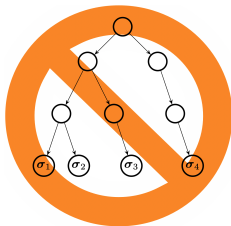
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Multi-OGP determined ALG for maximum independent set (Rahman-Virág 17, Wein 20) and mean-field spin glasses (H.-Sellke 21 & 23)

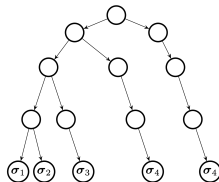
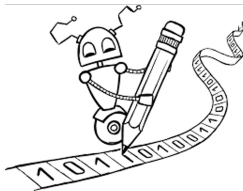


# “Ladder” Multi-OGP (Wein 20)

- 1 **Forbidden structure:** no constellation of solutions of prescribed geometry



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Doesn't occur w.h.p. by 1st moment argument

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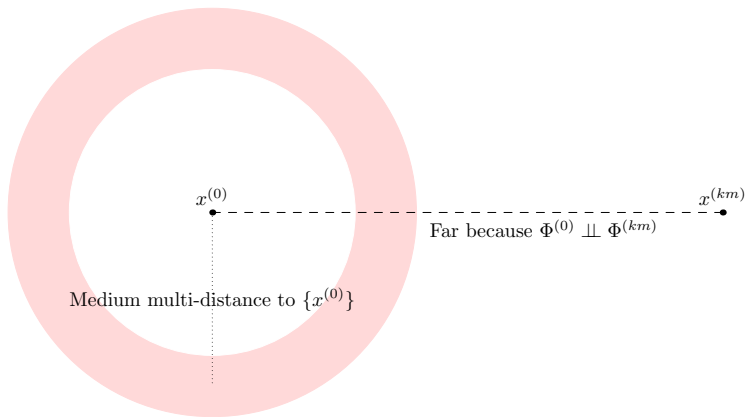
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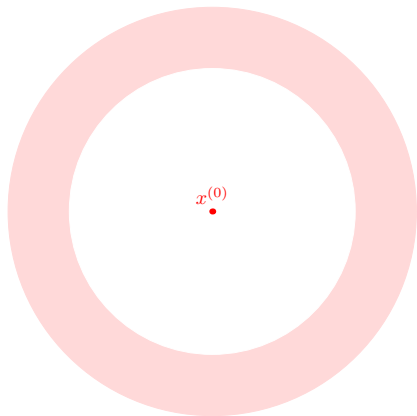
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$x^{(km)}$

$y^{(1)} = x^{(0)}$

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Close by LDP stability

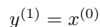
$x^{(0)}$   $x^{(1)}$



$x^{(km)}$

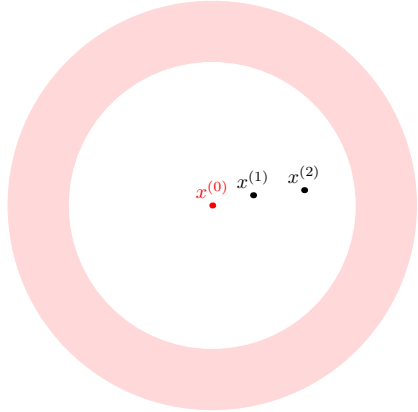


$y^{(1)} = x^{(0)}$



# Multi-OGP: Interpolation

Suppose LDP  $\mathcal{A}$  succeeds with high enough probability.  $x^{(t)} = \mathcal{A}(\Phi^{(t)})$ .  
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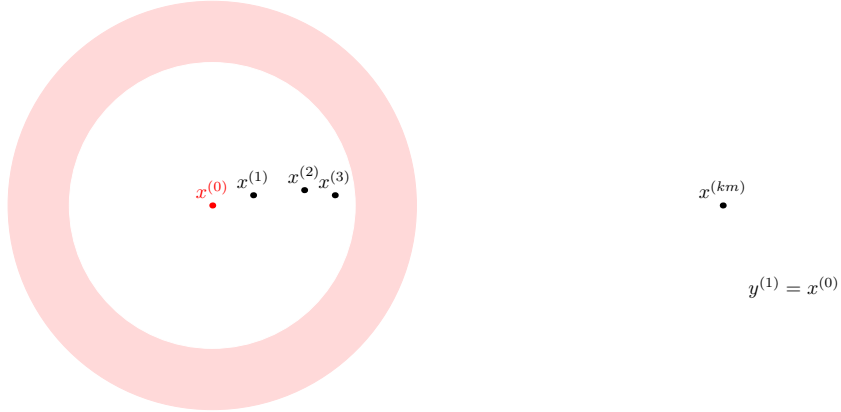
$x^{(0)}$   $x^{(1)}$   $x^{(2)}$

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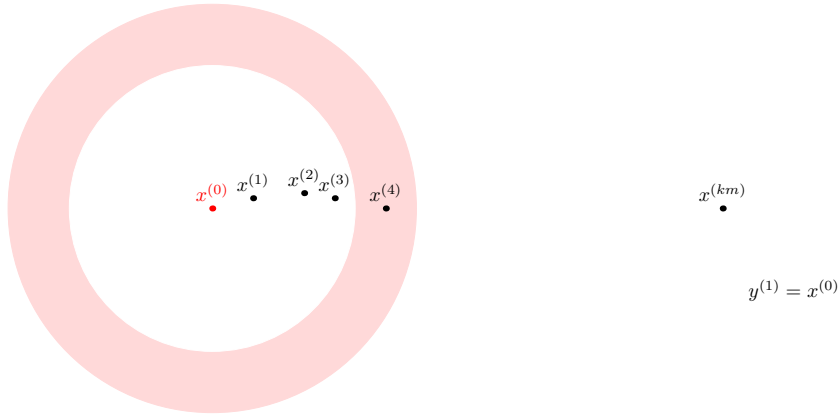
$x^{(0)}$   $x^{(1)}$   $x^{(2)}$   $x^{(3)}$

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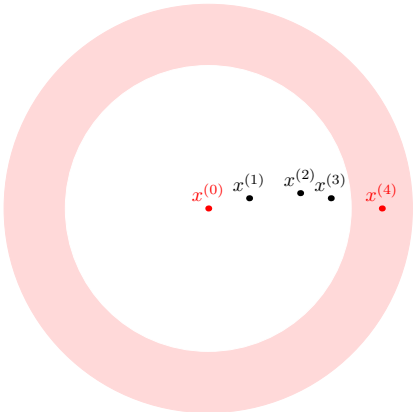
$x^{(0)}$   $x^{(1)}$   $x^{(2)}$   $x^{(3)}$   $x^{(4)}$

$x^{(km)}$

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$x^{(0)}$   $x^{(1)}$   $x^{(2)}$   $x^{(3)}$   $x^{(4)}$

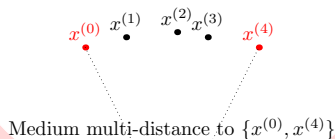
$x^{(km)}$

$$y^{(1)} = x^{(0)}$$

$$y^{(2)} = x^{(4)}$$

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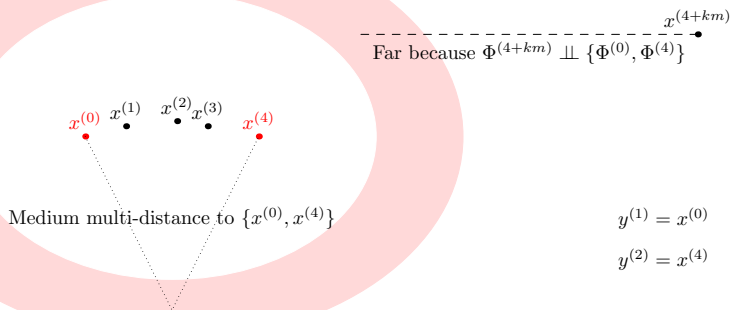
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$x^{(0)}$   $x^{(1)}$   $x^{(2)}$   $x^{(3)}$   $x^{(4)}$   $x^{(5)}$

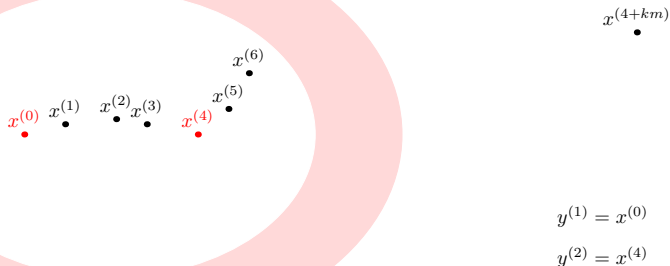
$x^{(4+km)}$

$$y^{(1)} = x^{(0)}$$

$$y^{(2)} = x^{(4)}$$

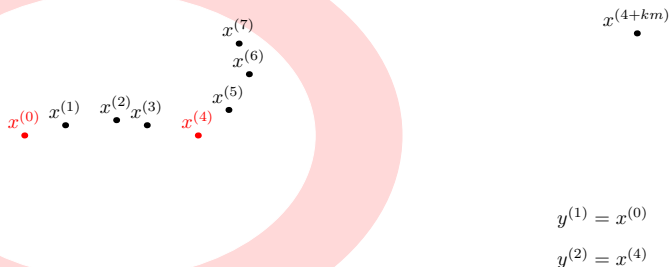
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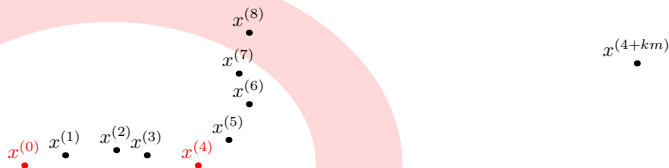
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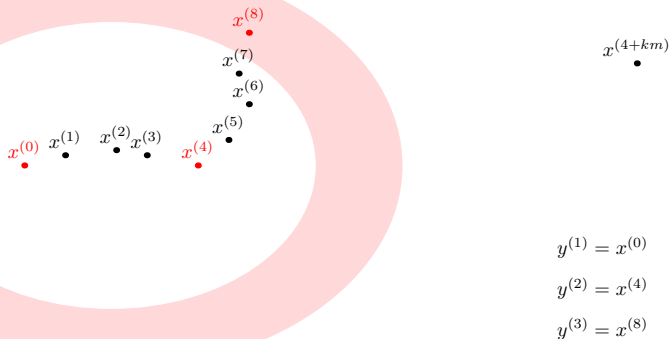


$$y^{(1)} = x^{(0)}$$

$$y^{(2)} = x^{(4)}$$

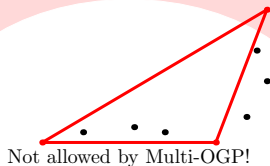
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$$y^{(1)} = x^{(0)}$$

$$y^{(2)} = x^{(4)}$$

$$y^{(3)} = x^{(8)}$$

Contradiction  $\Rightarrow \mathcal{A}$  cannot succeed.

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For us,

$$\mathbb{E}[\#\text{forbidden structure}] = (\text{entropic term})\mathbb{P}[y^{(1)}, \dots, y^{(k)} \text{ all satisfying assignments}].$$

Main challenge:  $\mathbb{P}$  term depends on  $y^{(1)}, \dots, y^{(k)}$  in complicated way.

**How to choose condition so  $\mathbb{P}$  term beats entropic term?**

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*Overlap profile:* for  $y^{(1)}, \dots, y^{(L)} \in \{\mathbf{T}, \mathbf{F}\}^n$ ,  $\pi = \pi(y^{(1)}, \dots, y^{(L)}) \in \mathbb{R}^{2^L - 1}$

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Multi-distance condition:

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is medium

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**Thank you!**