

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

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Simons Workshop on Rigorous Evidence for Information-Computation Tradeoffs

Joint work with Guy Bresler

# Random $k$ -SAT

## Problem (Random $k$ -SAT)

$\Phi \sim \Phi_k(n, m)$  is a  $k$ -CNF with  $m$  clauses, whose  $km$  literals are sampled i.i.d. from  $\text{unif}(\{x_1, \dots, x_n, \bar{x}_1, \dots, \bar{x}_n\})$ .

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**Q:** What prevents efficient algorithms from succeeding beyond ALG?

# Thresholds for Random $k$ -SAT

In double limit  $n \rightarrow \infty$  with  $\alpha = \alpha(k)$  fixed, then  $k \rightarrow \infty$ :

$$\text{OPT} = 2^k \log 2 - \frac{1}{2}(1 + \log 2) + o_k(1) \text{ [Ding, Sly, Sun '15]}$$

$$\text{ALG} \stackrel{?}{=} (1 - o_k(1))2^k \log k/k \text{ (FIX) [Coja-Oghlan '10]}$$

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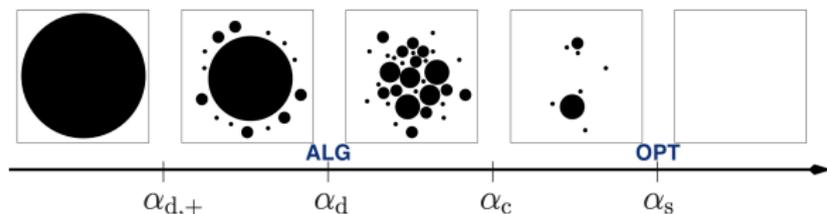
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Structural evidence: no better algorithm! [Achlioptas, Coja-Oghlan '08]

Solution space *shatters* beyond ALG.



[Krzakala, Montanari, Ricci-Tersenghi, Semerjian, Zdeborová '07]

# Main Result (informal)

Theorem (Bresler, H. '21)

*Due to shattering, low degree polynomial algorithms cannot solve random  $k$ -SAT above clause density  $4.911 \cdot 2^k \log k/k$ .*

# Low Degree Polynomials

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- $\text{sign}(\mathcal{A}(\Phi))$  is close (in Hamming distance) to an assignment satisfying most clauses of  $\Phi$ ,
- $|\mathcal{A}(\Phi)_i| \geq 1$  for most  $i$ ,
- $\mathbb{E} \|\mathcal{A}(\Phi)\|_2^2 = O(n)$ .

# Low Degree Polynomials

Can simulate:

- Spectral algorithms
- AMP
- Local algorithms on sparse graphs (including  $k$ -SAT factor graph)
- Belief / Survey Propagation guided decimation (bounded iterations)
- FIX

# Main Result

Let

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

Theorem (Bresler, H. '21)

*If  $\kappa > \kappa^*$  and  $m/n = \alpha = \kappa 2^k \log k/k$ , then no polynomial  $\mathcal{A} : \{0, 1\}^{mk \cdot 2^n} \rightarrow \mathbb{R}^n$  of degree  $D = o(n/\log n)$  solves random  $k$ -SAT with probability  $1 - \exp(-\Omega(D \log n))$ .*

# Algorithmic Lower Bounds on Random $k$ -SAT

$$\text{OPT} \sim 2^k \log 2$$

$$\text{ALG} \stackrel{?}{\sim} 2^k \log k/k$$

Clause Density Bound	Algorithm(s)	Reference
$O_k(2^k/k)$	DPLL algorithms	[Achlioptas, Beame, Molloy '04]
$(1 + o_k(1))2^{k-1} \log^2 k/k$	Balanced sequential local algorithms on NAE- $k$ -SAT	[Gamarnik, Sudan '17]
$(1 + o_k(1))2^k \log k/k$	Survey Propagation guided decimation	[Hetterich '16]
$O_k(2^k \log^2 k/k)$	Walksat	[Coja-Oghlan, Haqshenas, Hetterich '17]
$(1 + o_k(1))\kappa^* 2^k \log k/k$ $\kappa^* \approx 4.911$	Low degree polynomials	[Bresler, H. '21]

# OGP: A Topological Explanation of Hardness

Definition (Overlap Gap Property; [Gamarnik, Sudan '17] )

OGP holds if any two solutions are either close or far.

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OGP holds if any two solutions are either close or far.

- Maximum Independent Set [Gamarnik, Sudan '17] [Rahman, Virág '17]  
[Gamarnik, Jagannath, Wein '20] [Wein '20]
- NAE- $k$ -SAT [Gamarnik, Sudan '17]
- Maxcut on hypergraphs [Chen, Gamarnik, Panchenko, Rahman '17]
- Spin Glasses [Gamarnik, Jagannath '19] [Gamarnik, Jagannath, Wein '20]  
[Gamarnik, Jagannath, Wein '21]
- Number Partitioning [Gamarnik, Kızıldağ '21]
- Planted Problems [Gamarnik, Zadik '19], [Gamarnik, Jagannath, Sen '19],  
[Ben Arous, Wein, Zadik '20]

Survey: [Gamarnik '21]

# Ensemble OGP: interpolation

Used in other contexts by [Chen, Gamarnik, Panchenko, Rahman '17],  
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Interpolation path:  $\Phi^{(0)} \quad \Phi^{(1)} \quad \Phi^{(2)} \quad \dots \quad \Phi^{(km)}$   
 $\Phi^{(0)} \sim \Phi_k(n, m)$ ,  $\Phi^{(t)}$  resamples  $t$ th literal of  $\Phi^{(t-1)}$ .

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Set  $x^{(t)} = \mathcal{A}(\Phi^{(t)})$ .

# The Forbidden Structure in Ensemble OGP

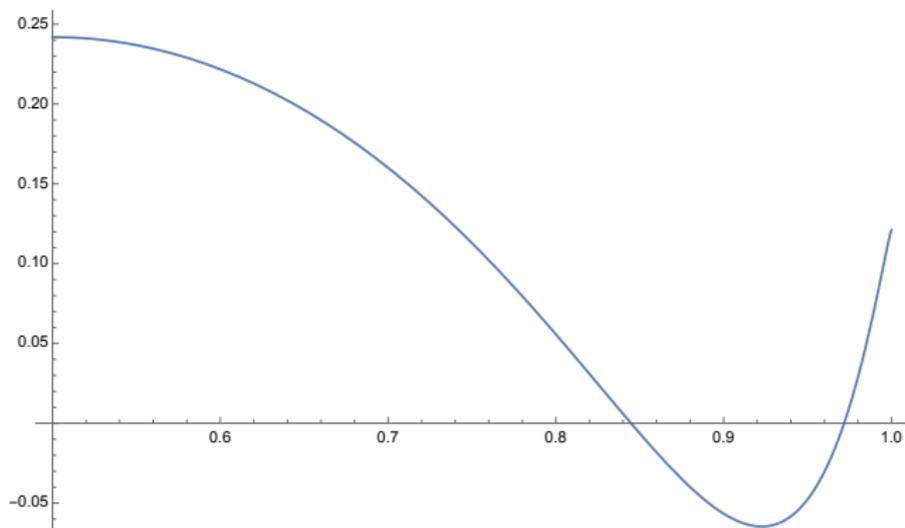
Two assignments  $y^{(1)}, y^{(2)} \in \{\text{T}, \text{F}\}^n$  such that:

- $y^{(i)}$  satisfies some  $\Phi^{(t_i)}$ ;
- $y^{(2)}$  is medium distance from  $y^{(1)}$ .

(w.h.p. does not exist if  $\alpha \geq \frac{1+\epsilon}{2}\text{OPT}$  by 1st moment argument)

## 2-overlap landscape for $\alpha = \frac{1+\varepsilon}{2}\text{OPT}$

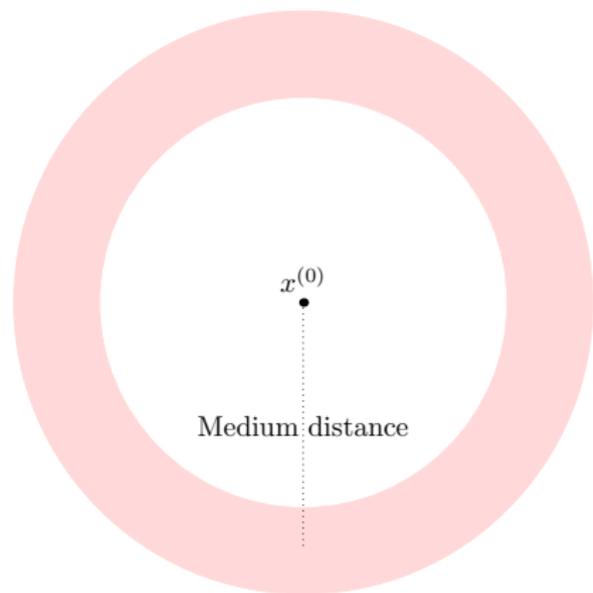
$\frac{1}{n} \log \mathbb{E} \#$  (pairs of satisfying assignments with overlap  $x$ )



Graph negative  $\Rightarrow$  OGP holds.

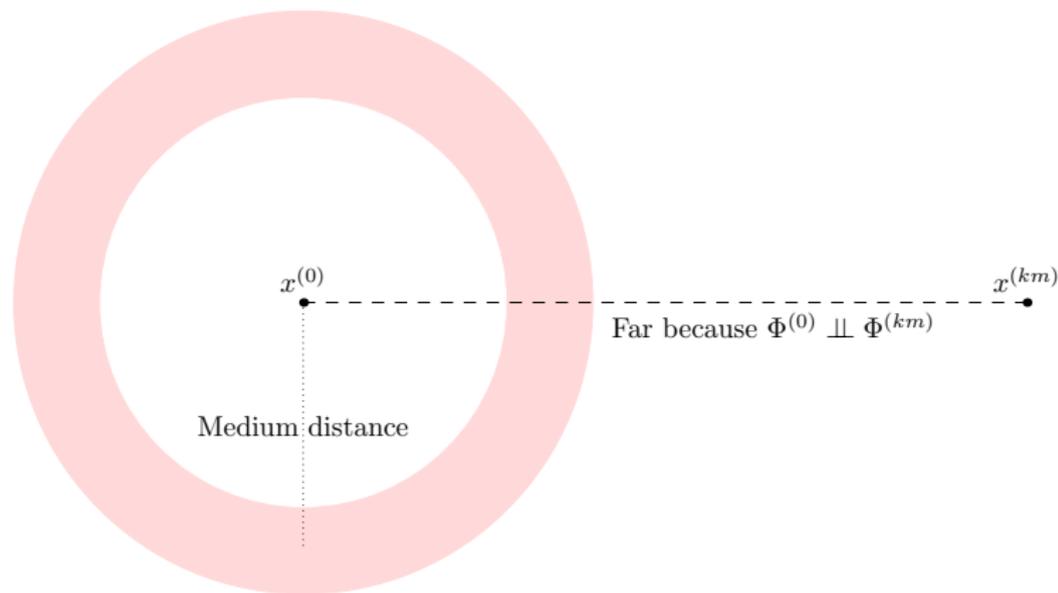
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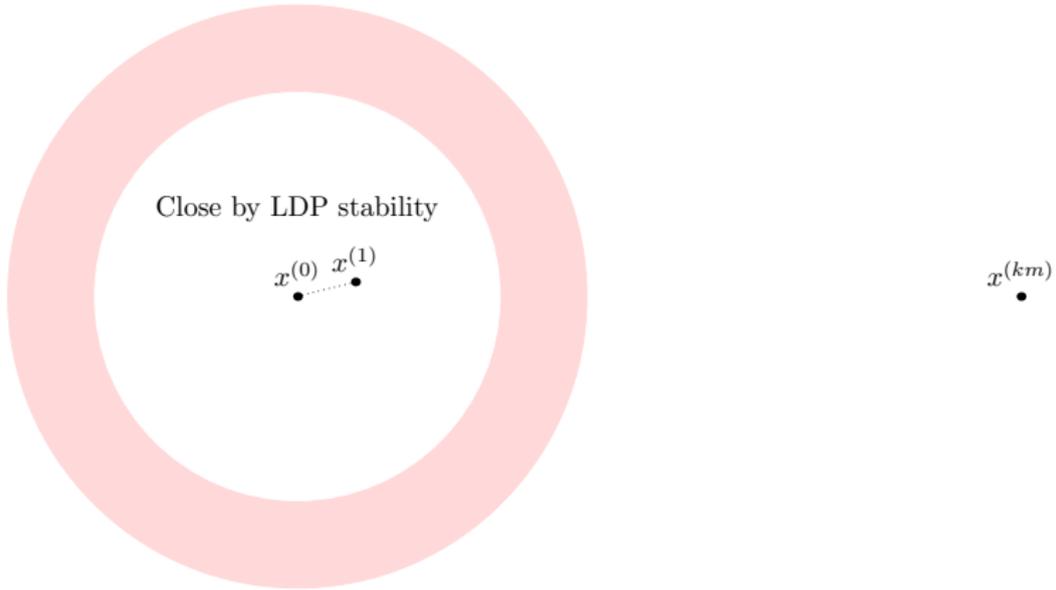


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Close by LDP stability

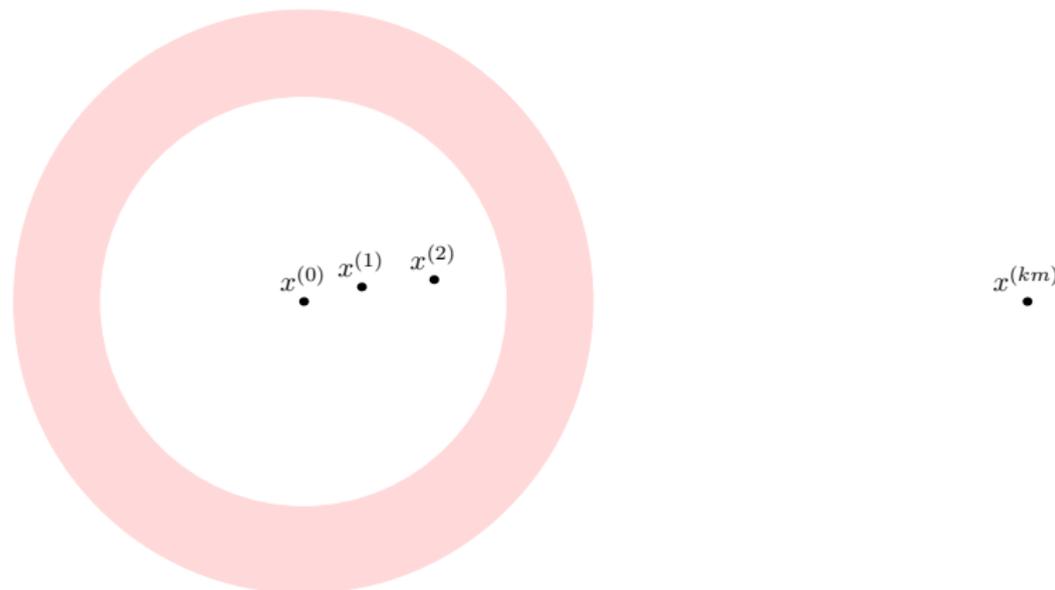
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$x^{(km)}$

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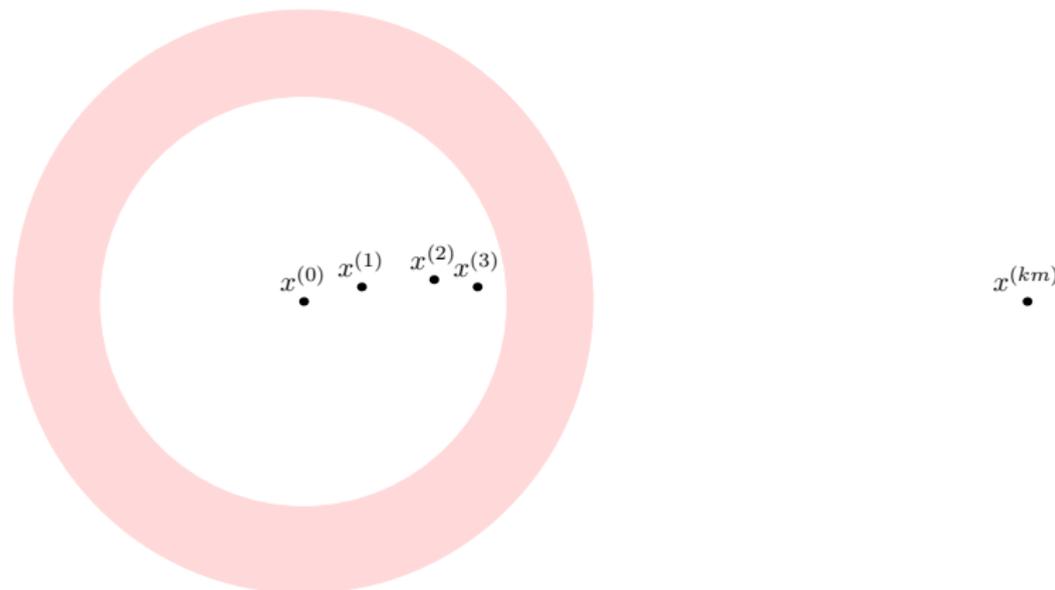


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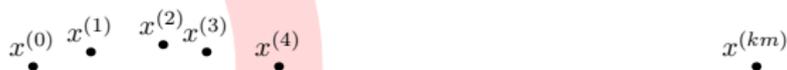


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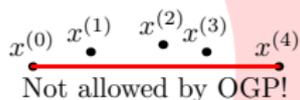
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Not allowed by OGP!

$x^{(km)}$

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

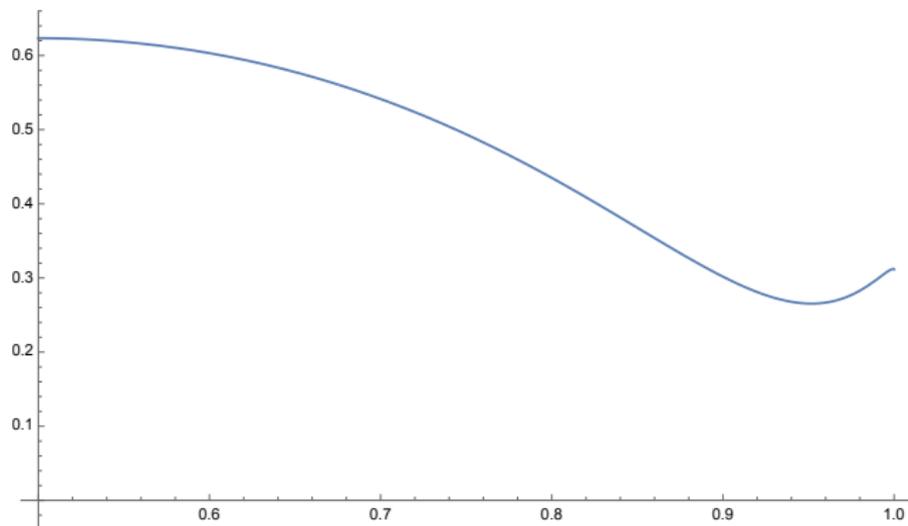
# How to improve past $\frac{1+\epsilon}{2}$ OPT?

Ensemble OGP gets stuck at  $\frac{1+\epsilon}{2}$  OPT, far above ALG.

Ensemble Multi-OGP: forbidden structure using several solutions.

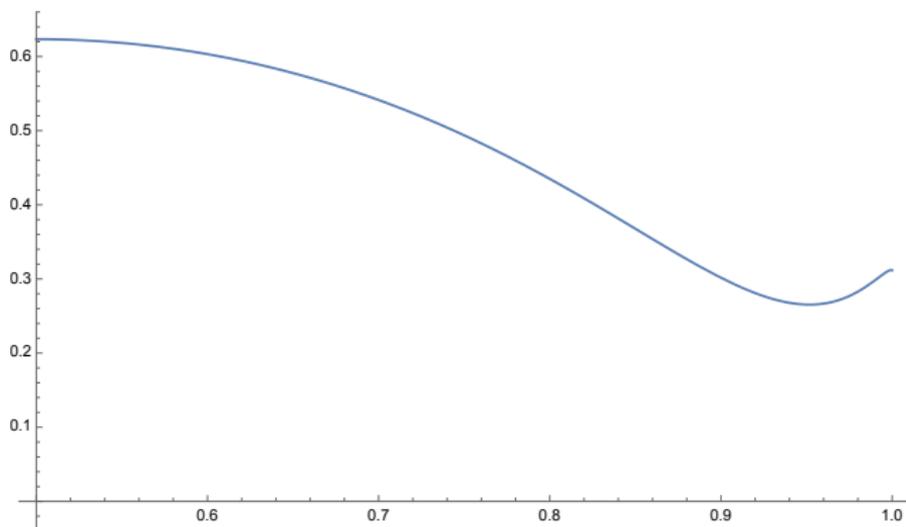
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Amplify the dip using forbidden structure with many medium overlaps!

# Ensemble Multi-OGP: interpolation

Used in other contexts by [Wein '20]

Related: [Rahman, Virág '17], [Gamarnik, Sudan '17], [Gamarnik, Kızıldağ '21]

Let  $\alpha \geq \kappa 2^k \log k/k$ . ( $\approx \kappa \text{ALG}$ )

Suppose LDP  $\mathcal{A}$  solves  $\Phi \sim \Phi_k(n, m)$  with sufficiently high probability, where  $m/n = \alpha$ .

Interpolation path:  $\Phi^{(0)} \quad \Phi^{(1)} \quad \Phi^{(2)} \quad \dots \quad \Phi^{(k^2 m)}$   
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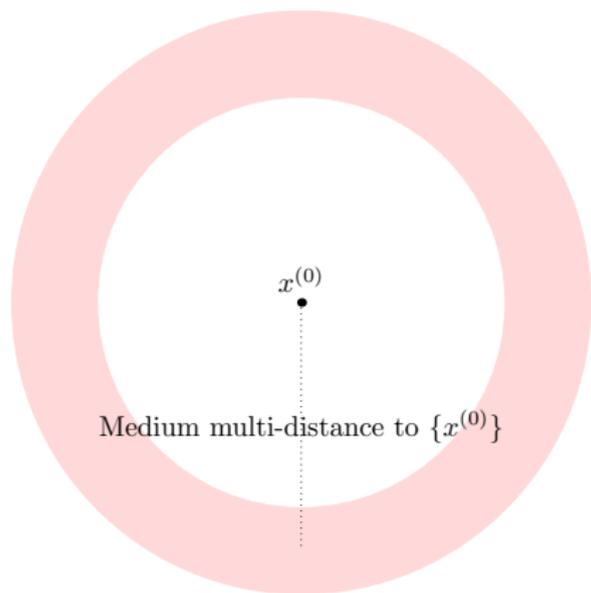
Forbidden structure:  $k$  assignments  $y^{(1)}, \dots, y^{(k)} \in \{\mathsf{T}, \mathsf{F}\}^n$  such that:

- $y^{(i)}$  satisfies some  $\Phi^{(t_i)}$ ;
- $y^{(i)}$  has medium multi-distance from  $\{y^{(1)}, \dots, y^{(i-1)}\}$  for  $i \geq 2$ .

(w.h.p. does not exist if  $\alpha \geq \kappa 2^k \log k/k$  by first moment argument)

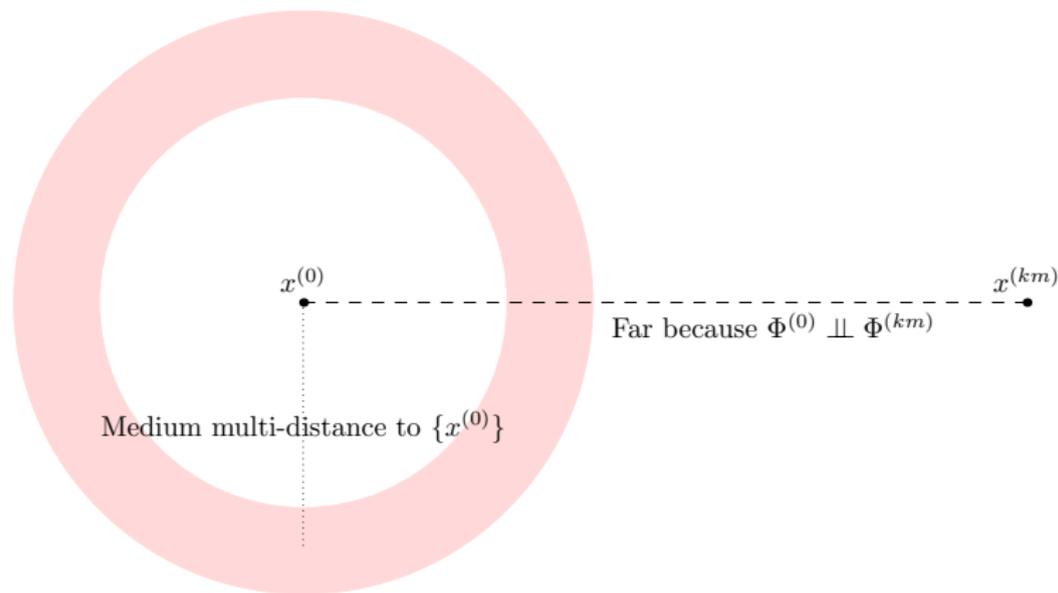
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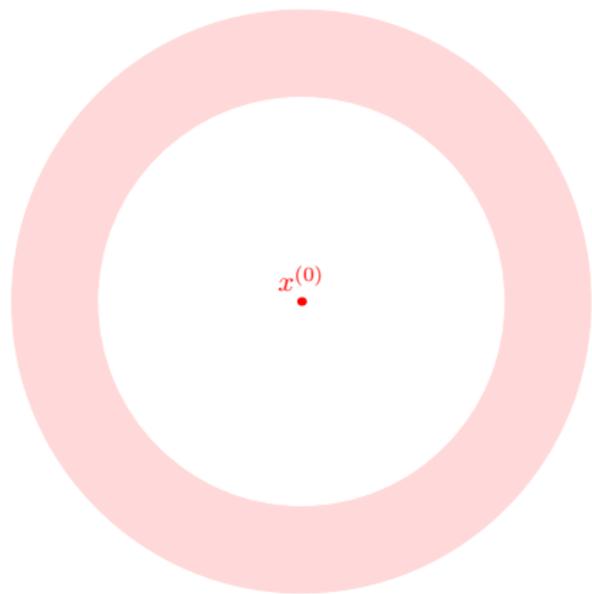
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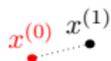
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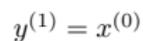
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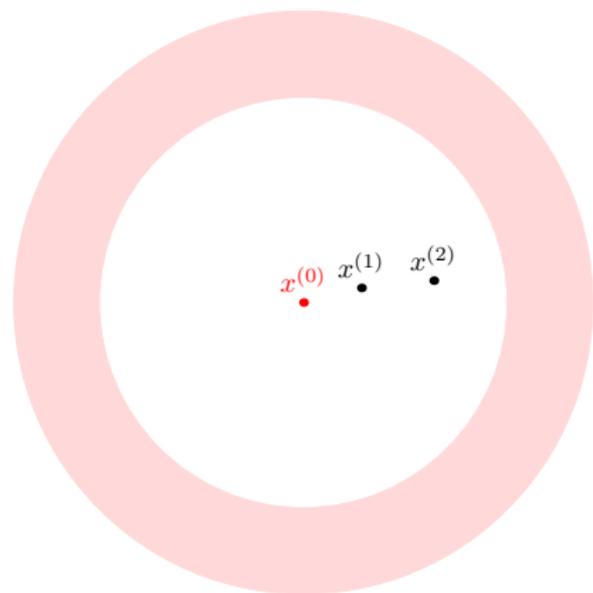


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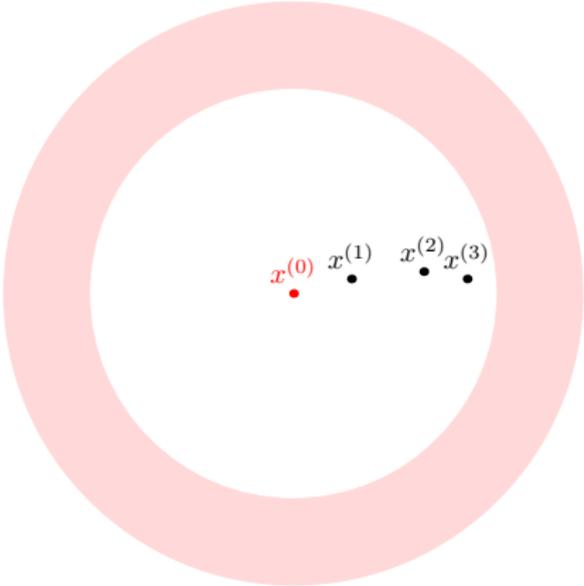


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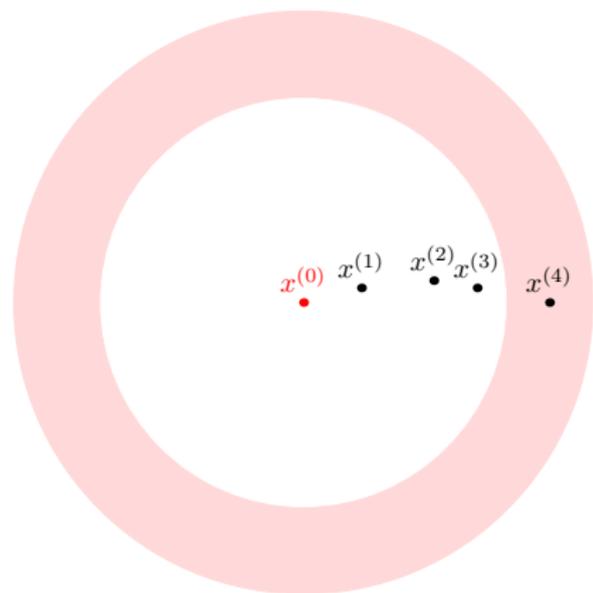
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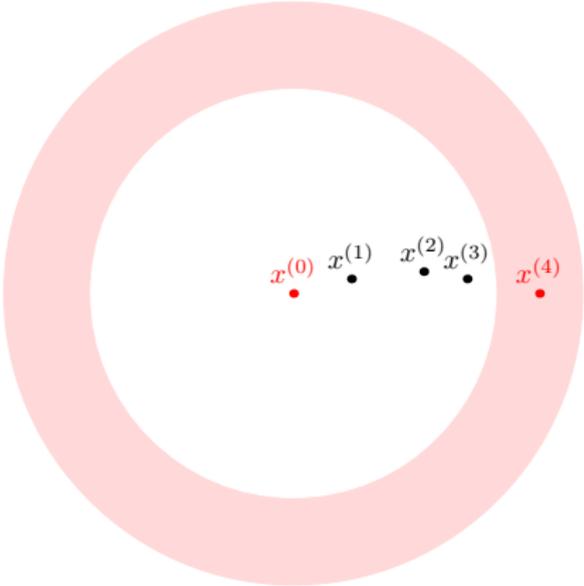


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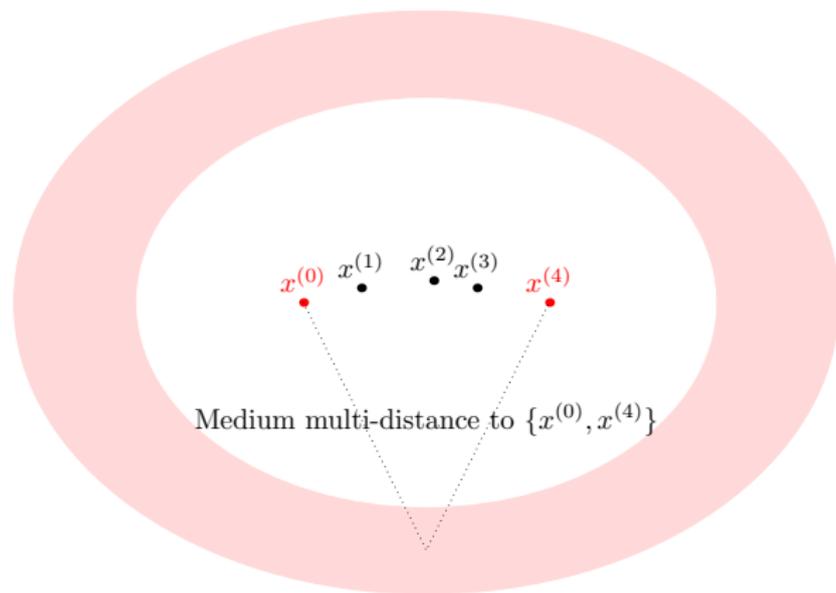
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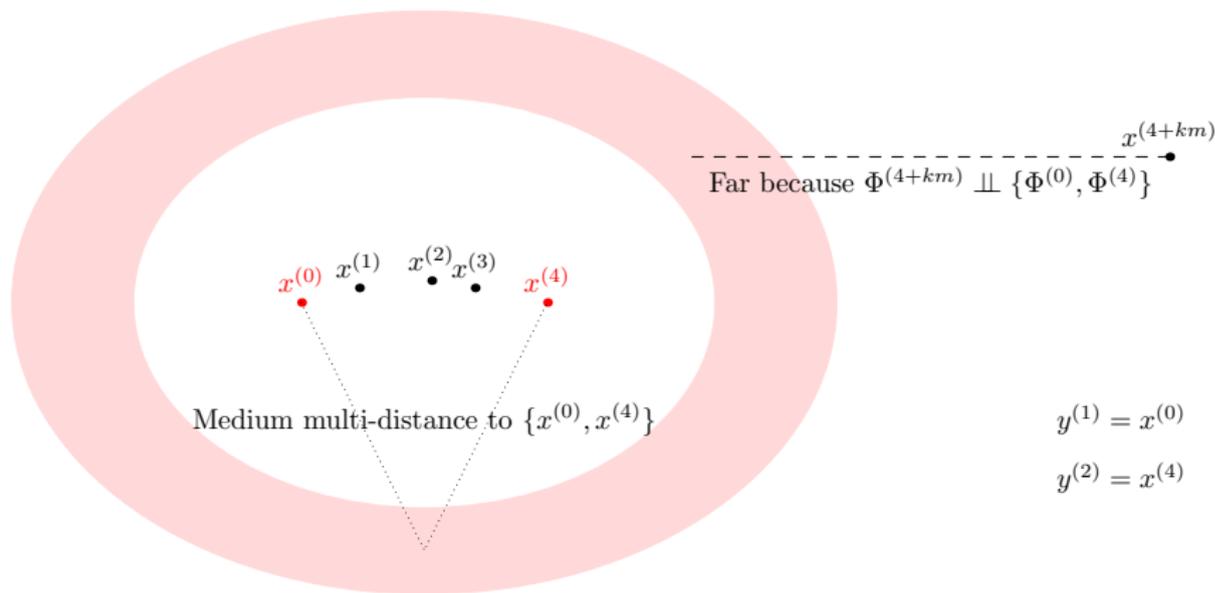


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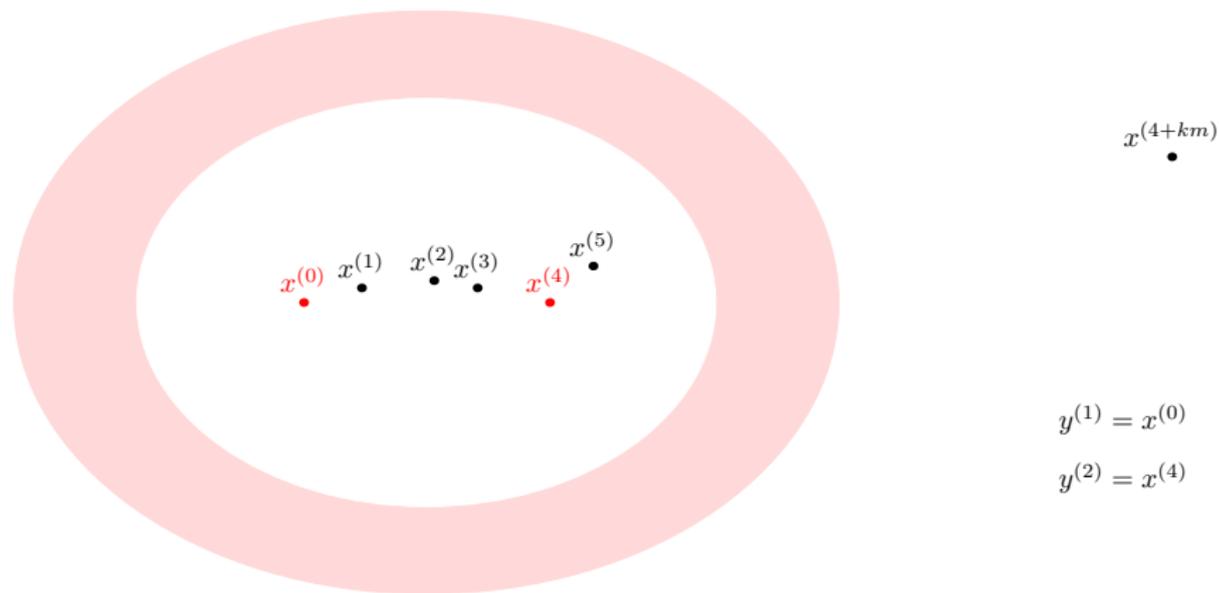
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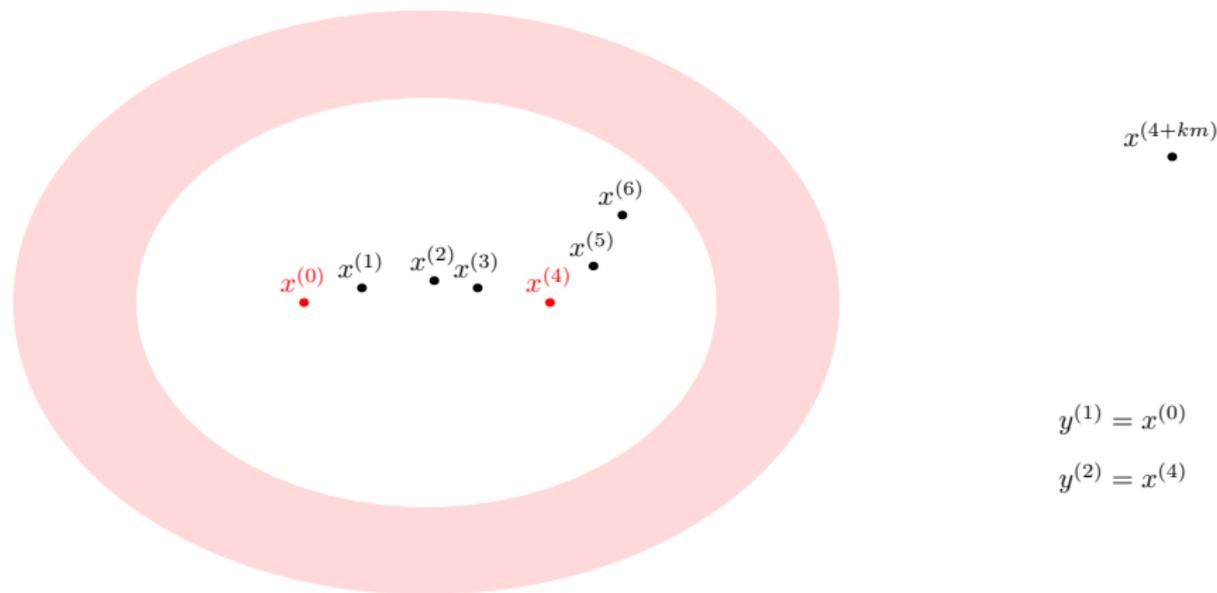
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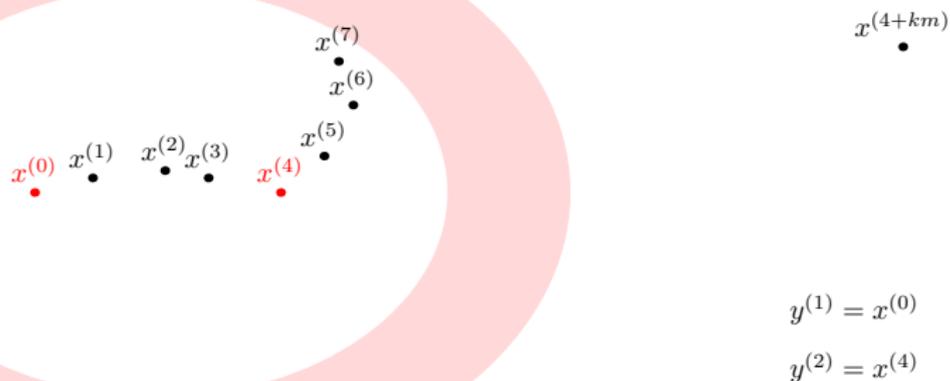
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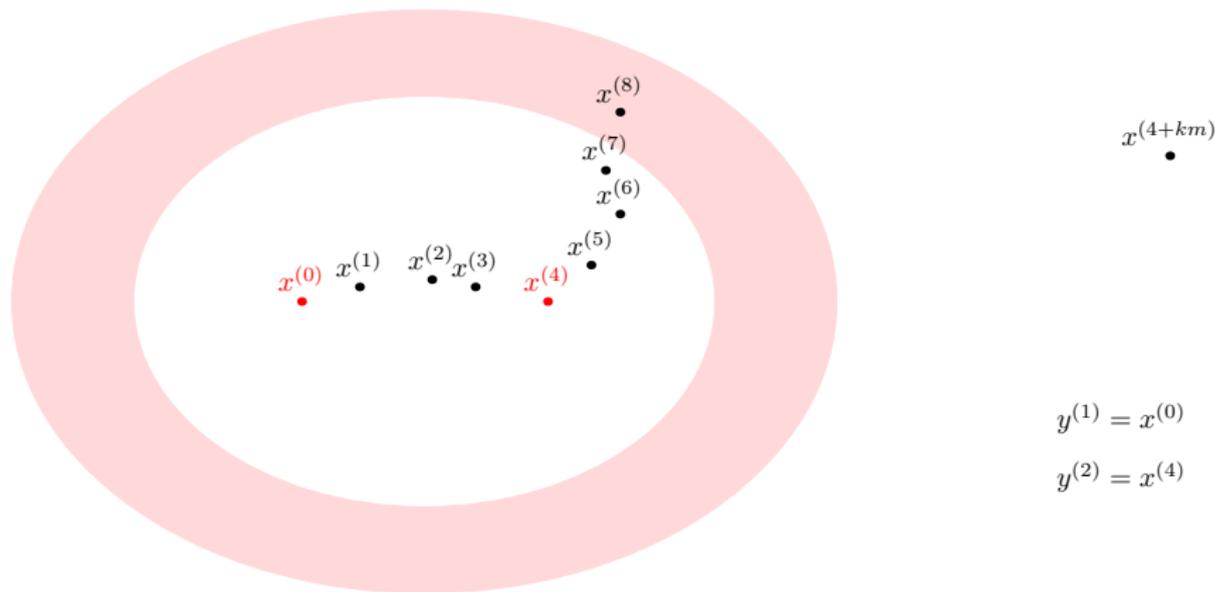
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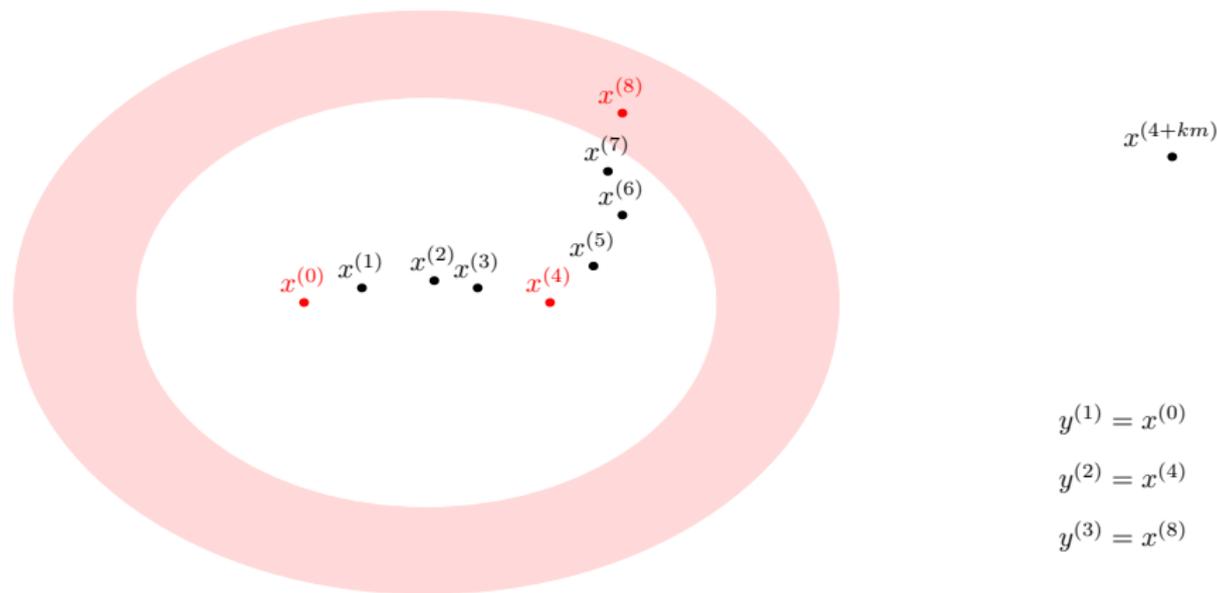
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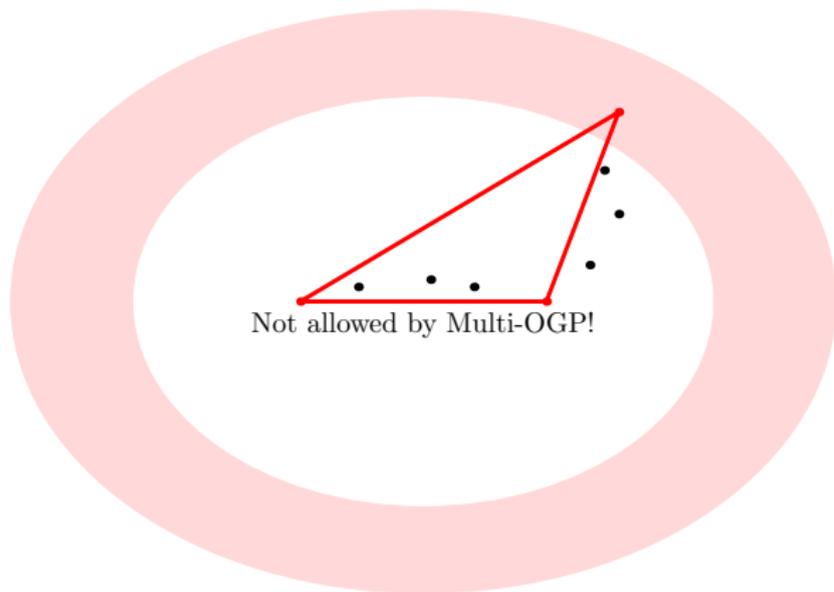
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$$y^{(1)} = x^{(0)}$$

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

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

# The Challenge: Picking the Right Moats

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Only previous ensemble multi-OGP: [Wein '20], max independent set.  
Forbidden structure: large ind. sets  $S_1, \dots, S_L$ , where for all  $i \geq 2$ ,

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Our contributions:

- Identify correct forbidden structure for random  $k$ -SAT;
- Navigate more intricate first moment computation;
- We believe methods can generalize to more problems.

# Overlap Profiles

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

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*Overlap entropy:*

$$H(\pi) = - \sum_{S,T \text{ partition } \{1, \dots, \ell\}} \pi_{S,T} \log \pi_{S,T}$$

# Defining the Forbidden Structure

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Forbidden structure:  $y^{(1)}, \dots, y^{(k)} \in \{\mathbf{T}, \mathbf{F}\}^n$  such that:

- $y^{(i)}$  satisfies some  $\Phi^{(t_i)}$ ;
- For all  $i \geq 2$ , (this is the medium multi-distance condition)

$$H(\pi(y^{(1)}, \dots, y^{(i)})) - H(\pi(y^{(1)}, \dots, y^{(i-1)})) \approx \beta^* \frac{\log k}{k}$$

# Summary

We generalize ensemble multi-OGP, previously tailored to max independent set, to random  $k$ -SAT.

We believe this methodology generalizes.

LDPs don't solve random  $k$ -SAT at clause density  $4.911 \cdot 2^k \log k/k$ .

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**Takeaway:** many-way OGP can show algorithmic hardness at or just beyond ALG.