

The Hard Problems Are Almost Everywhere For Random CNF-XOR Formulas

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Motivation

- Approximate Model Counting/Sampling: Given a CNF formula F ,
 - *Counting*: Estimate the number of solutions of F .
 - *Sampling*: Near-uniformly sample a solution of F .
 - Applications in probabilistic inference, automated reasoning, ...
- Hashing-based sampling and counting algorithms reduce these problems to finding solutions to CNF-XOR Formulas (Boolean formulas with both CNF and XOR clauses).

[Gomes *et al.* 2007], [Chakraborty *et al.*, 2013], [Ermon *et al.* 2013]

- **Goal**: Analyze the “behavior” of CNF-XOR formulas.
 - Which CNF-XOR Formulas are “hard” to solve (find a solution)?
 - **Empirical observation**: solving CNF-XOR formulas are hard in practice!

Prior Work: Which CNF Formulas are Hard?

A k -CNF clause is the “or” of k variables, each possibly negated.

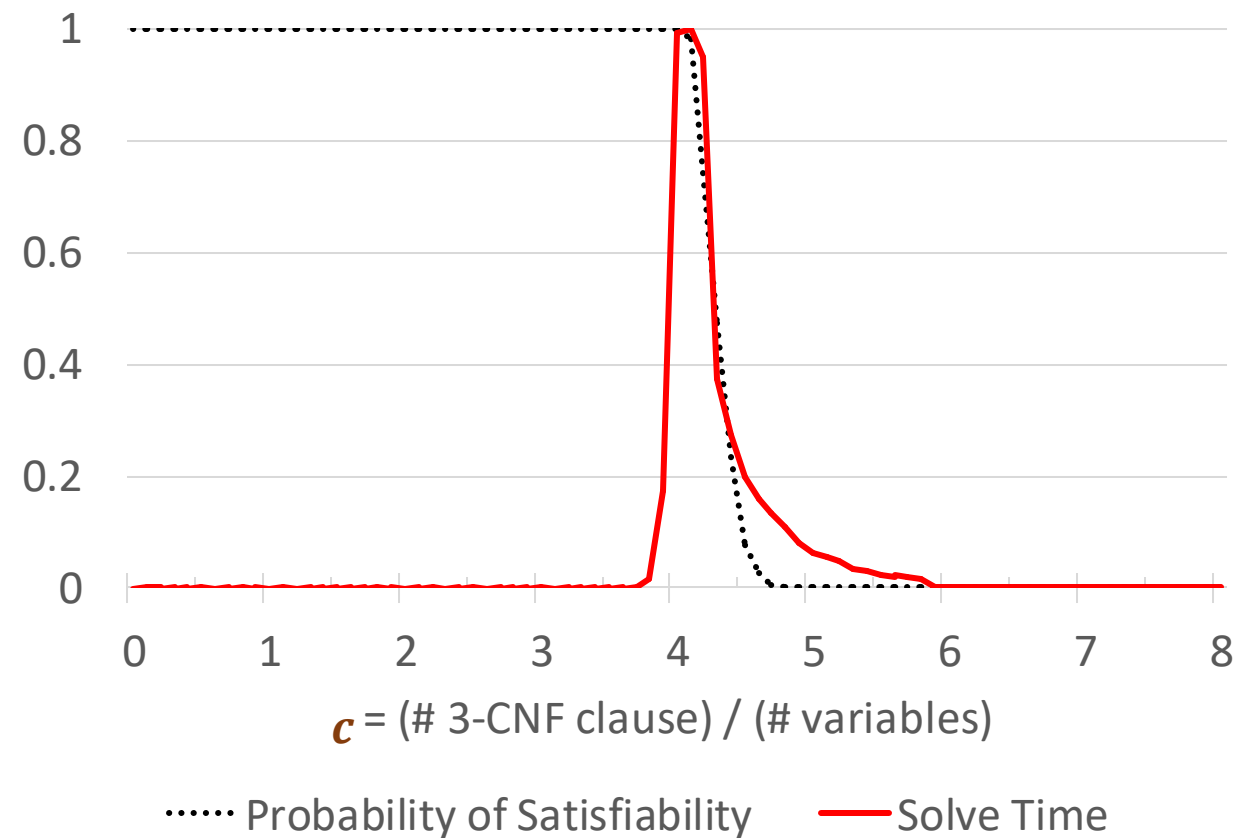
Ex: $X_1 \vee \neg X_5 \vee X_6$

$\text{CNF}_k(n, c)$ is a random formula with:
 n variables.

cn independent, uniformly
selected k -CNF clauses.

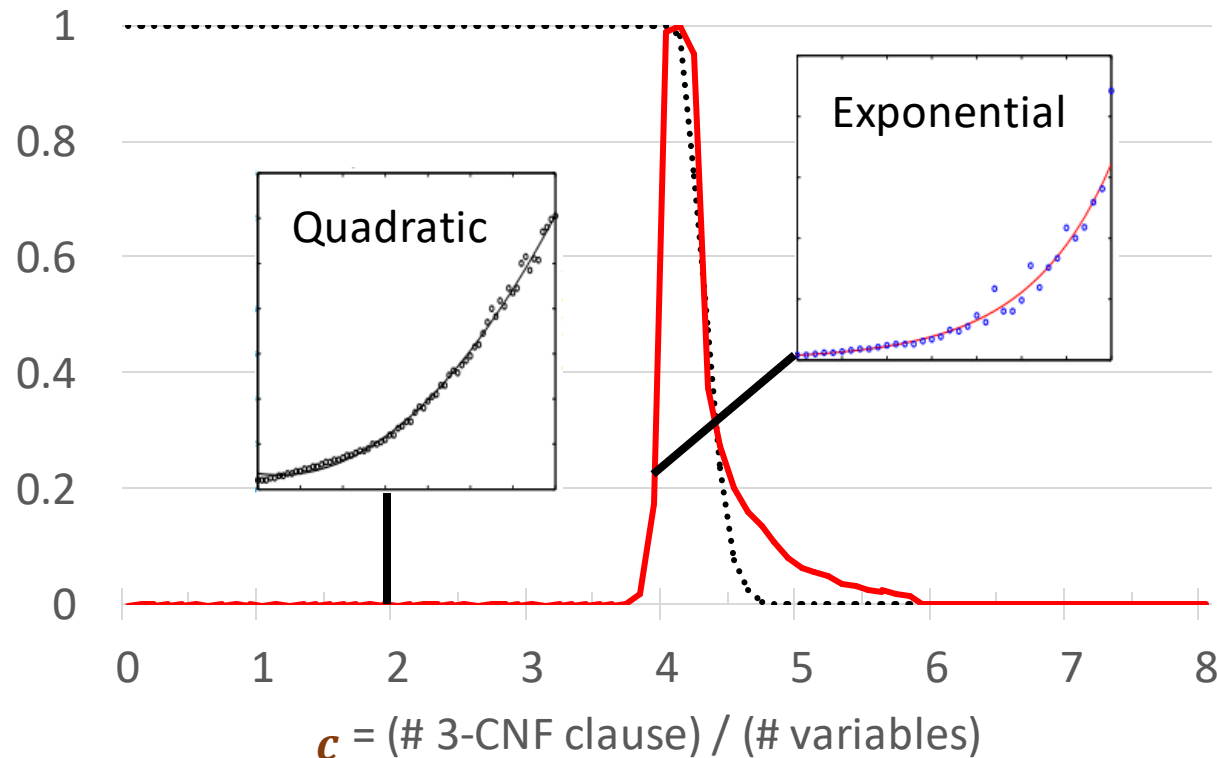
For which parameters is $\text{CNF}_k(n, c)$
hard to solve?

Median solve time of SAT Solver on $\text{CNF}_3(200, c)$



Prior Work: Runtime Scaling of random k-CNF

Median solve time of SAT Solver on $\text{CNF}_3(200, c)$



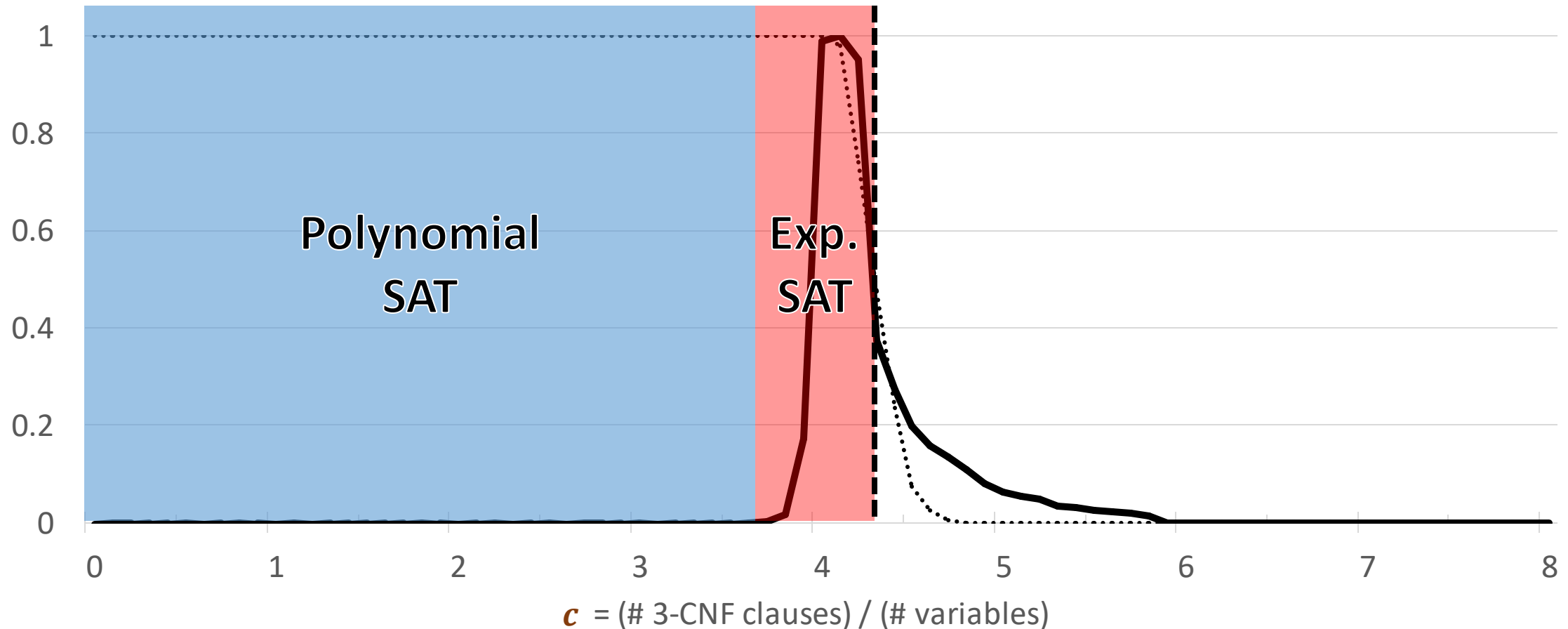
..... Probability of Satisfiability — Solve Time

Runtime-Scaling Experiment [Coarfa *et al.*, 2003]

1. Fix a SAT Solver (GRASP).
2. Fix a value for k and c (i.e., all parameters except n , the number of variables).
3. Incrementally increase n . At each n :
 $T(n)$ = median runtime of solving $\text{CNF}_k(n, c)$
4. Compute the best-fit line to $T(n)$

Prior Work: Runtime Scaling of random k-CNF

Average satisfiability and median solve time of SAT Solver (CDCL) on $\text{CNF}_3(200, c)$



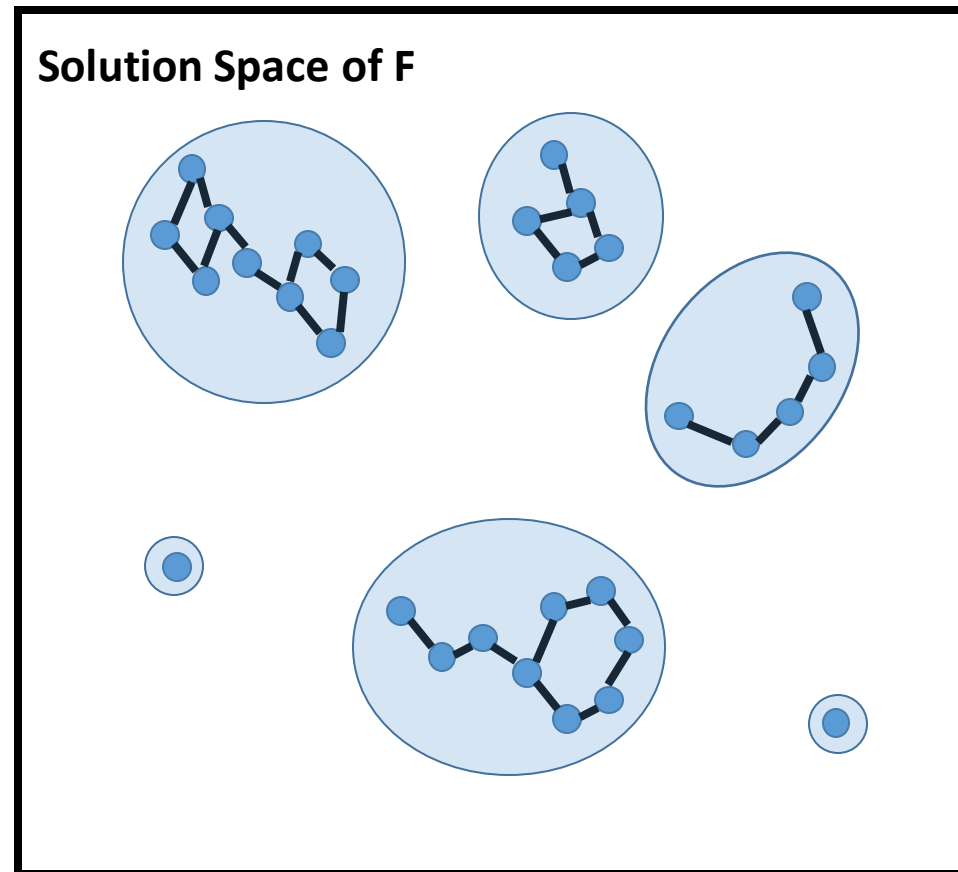
[Coarfa *et al.*, 2003]

..... Probability of Satisfiability — Solve Time

Solution-Space Geometry

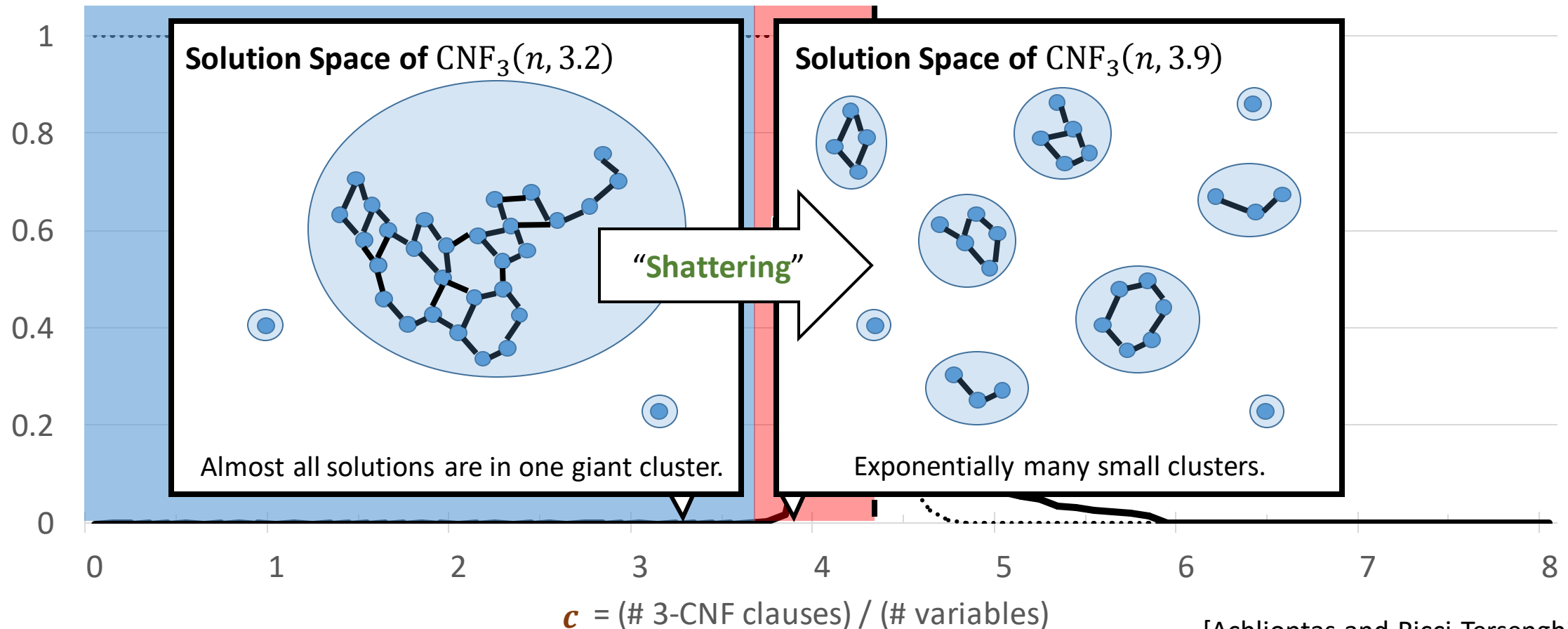
[Achlioptas and Ricci-Tersenghi, 2011]

Draw an edge between two solutions if they differ on no more than δn variables.



Solution-Space Geometry of Random k-CNF

Average satisfiability and median solve time of SAT Solver (CDCL) on $\text{CNF}_3(200, c)$



[Achlioptas and Ricci-Tersenghi, 2011]

..... Probability of Satisfiability — Solve Time

Random p-XOR Formulas [Goerdt, 1996]

An **XOR clause** is the “sum mod 2” of variables, set equal to 0 or 1.

- **Ex:** $X_1 + X_4 + X_5 + X_6 = 0 \pmod{2}$

XOR^p(n, x) is a random formula with:

- **n** variables.
- **xn** independent XOR clauses, where each clause is randomly sampled by including each variable independently with probability **p**.

(Exactly as in hashing-based sampling and counting algorithms)

Which XOR Formulas are hard to solve?

XOR Formulas are all easy (polynomial) with Gaussian Elimination.

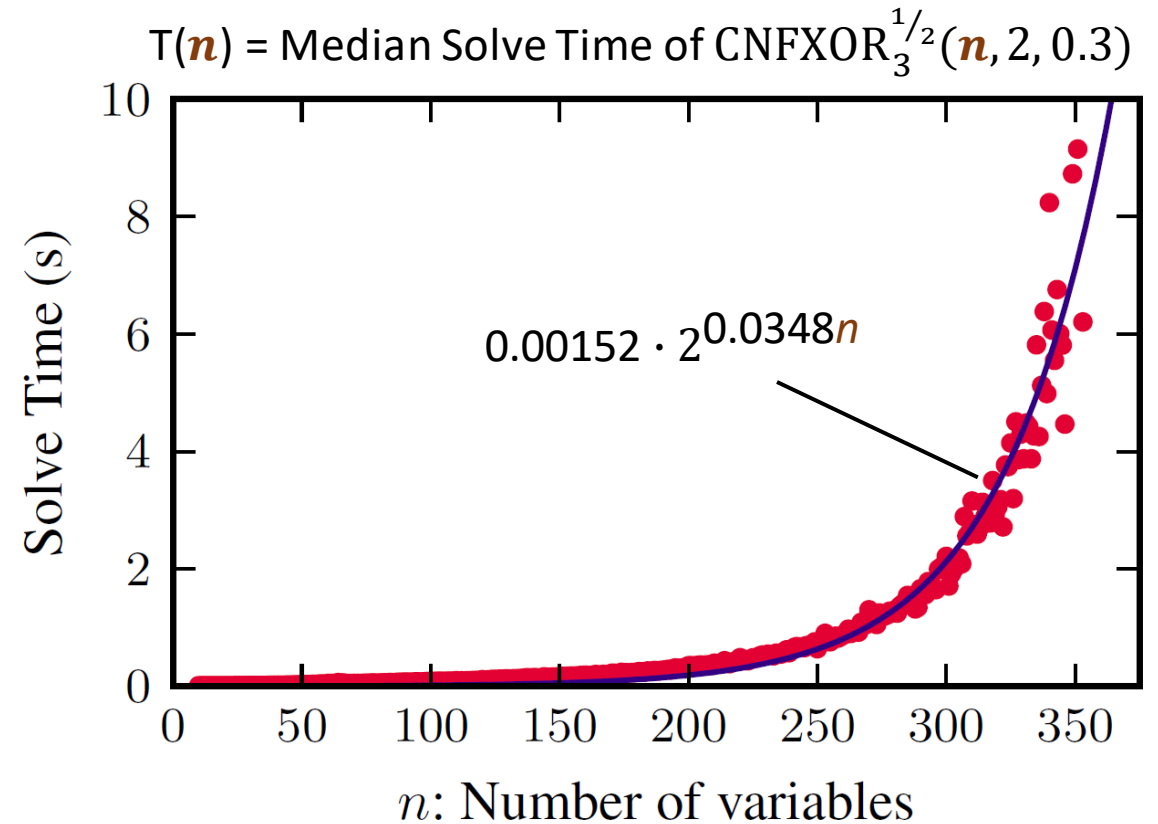
Combining CNF and XOR Together

- **Definition:** A **CNF-XOR Formula** is a formula where each clause is either a CNF clause or an XOR clause.
- **Definition:** $\text{CNFXOR}_k^p(n, c, x) := \text{CNF}_k(n, c) \wedge \text{XOR}^p(n, x)$
(A random formula with n variables, cn k -CNF clauses, and xn p -XOR clauses)
- **Goal:** Analyze the “behavior” of CNF-XOR formulas.
 - [Dudek *et al.*, 2016] There is a phase-transition in the satisfiability of $\text{CNFXOR}_k^{1/2}(n, c, x)$.
 - In this work, we analyze the runtime scaling behavior of SAT Solvers on $\text{CNFXOR}_k^p(n, c, x)$.

Our Work: Runtime Scaling of CNF-XOR Formulas

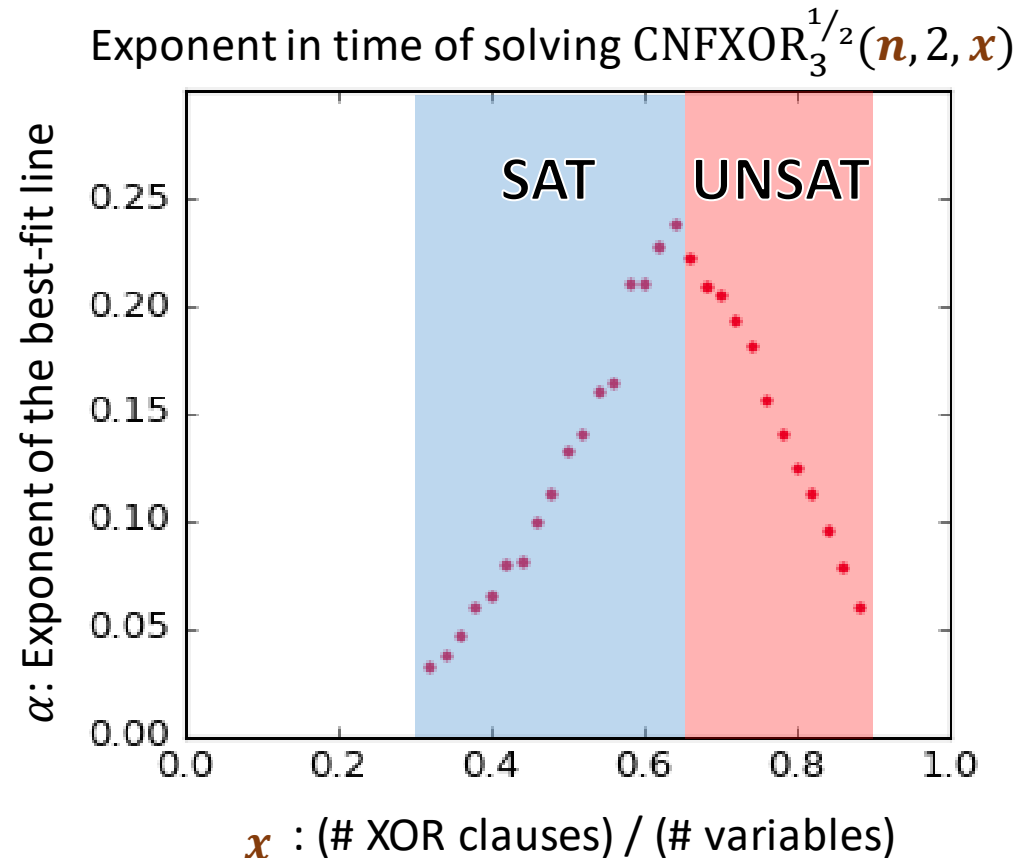
Runtime-Scaling Experiment

1. Fix a SAT Solver (CryptoMiniSAT).
2. Fix a value for k , p , c , and x (i.e., all parameters except n , the number of variables).
3. Incrementally increase n . At each n :
 $T(n)$ = median runtime of solving $\text{CNFXOR}_k^p(n, c, x)$
4. Compute the best-fit line to $T(n)$



How does the Runtime Scaling change?

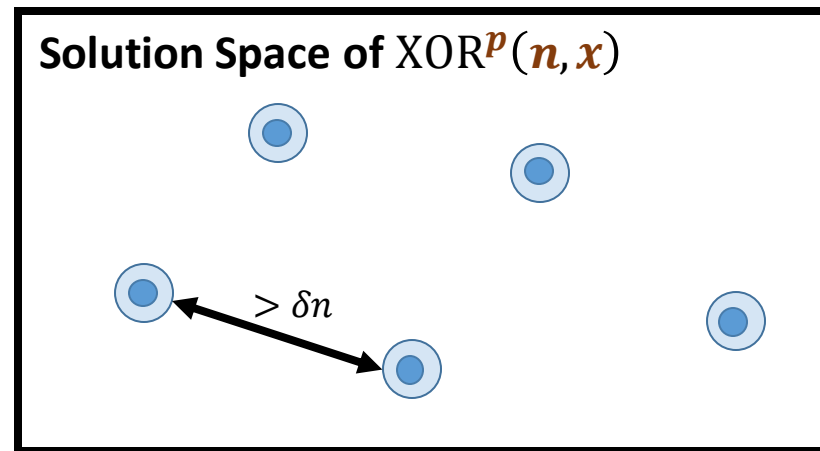
Q: How does the runtime scaling change when x changes?



A: The exponent peaks near the phase-transition location.

Theory: Shattering the CNF-XOR Solution Space

- What does the solution-space geometry of random CNF-XOR formulas look like?
- What does the solution-space geometry of random XOR formulas look like?
- **Theorem:** For all $p \in (0, 1/2]$, $x \in (0, 1)$, the solution space of $\text{XOR}^p(n, x)$ is **shattered**.



- **Corollary:** The solution space of random CNF-XOR formulas is *always* **shattered**.

Conclusion: CNF-XOR formulas are hard almost everywhere.

Prior Work: SAT Solvers scale exponentially on random k -CNF formulas when the solution-space shatters.

Our Contribution: “When do SAT solvers scale exponentially on CNF-XOR formulas?”

- CryptoMiniSAT scales exponentially for many parameters; no polynomial region.
- The CNF-XOR solution-space always shatters.
- Explain empirical observations on solving CNF-XOR formulas.

Future Work: Why are unsatisfiable CNF-XOR formulas hard?

Thanks!

Citations

- [Achlioptas and Ricci-Tersenghi, 2011] D. Achlioptas and F. Ricci-Tersenghi. On the Solution-Space Geometry of Random Constraint Satisfaction Problems. In *Random Structures & Algorithms*, 2011.
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