

FourierSAT: A Fourier Expansion-Based Algebraic Framework for Solving Hybrid Boolean Constraints



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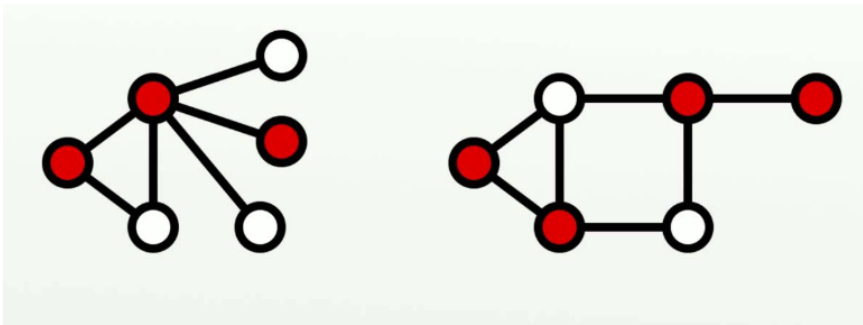
Boolean Constraints and Satisfiability (SAT)

- Boolean variables: $x, y, z, w \dots \in \{0, 1\}$
- Examples of Boolean Constraints:
 - CNF constraints: $x \vee y \vee z$
 - XOR constraints: $x \oplus y \oplus z$
 - Cardinality constraints: $x + y + z + w \geq 2$
 - Not-All-Equal constraints: $\neg(x \leftrightarrow y \leftrightarrow z \leftrightarrow w)$
- Boolean Satisfiability (SAT) Solving: finding a solution of a system consists of Boolean constraints.

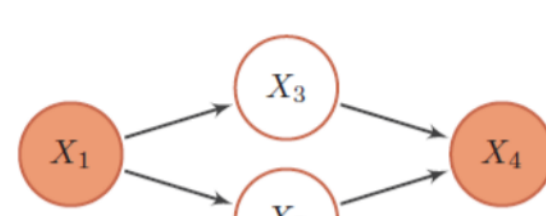
$$\begin{cases} x \vee y \vee z \\ x \oplus y \oplus z \\ x + y + z + w \geq 2 \\ NAE(x, y, z, w) \end{cases}$$

$x = y = z = 1, w = 0$ is a solution!

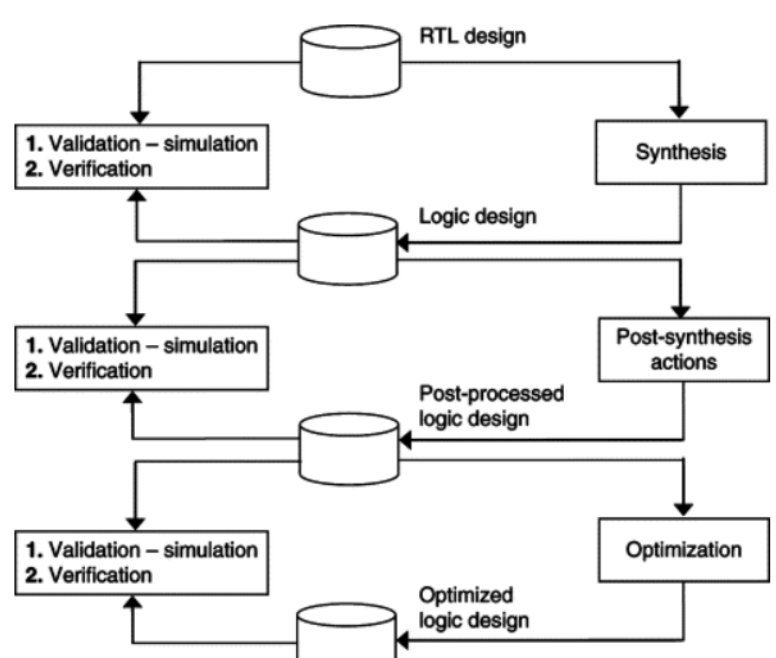
Applications of SAT



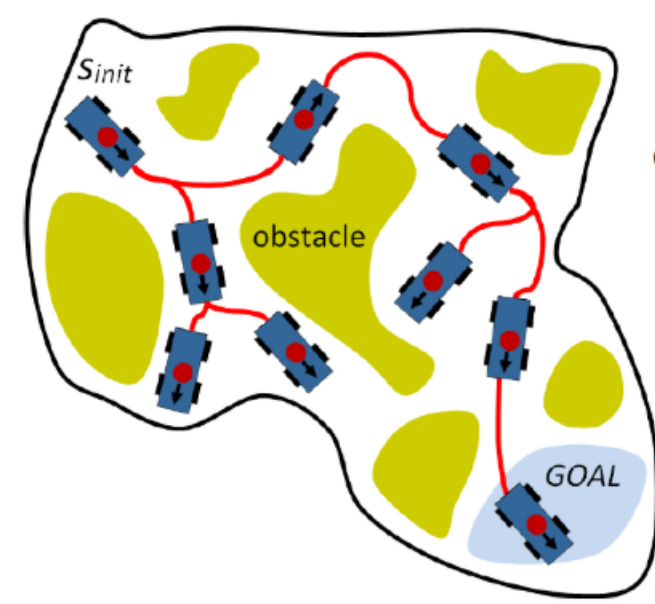
SAT can represent a lot of discrete optimization problem, e.g. minimum vertex covering and vertex coloring in graph theory.



SAT can be used to solve discrete integration problem such as computing the partition function, a fundamental problem in AI.



SAT solver is a critical tool in hardware design and software verification.

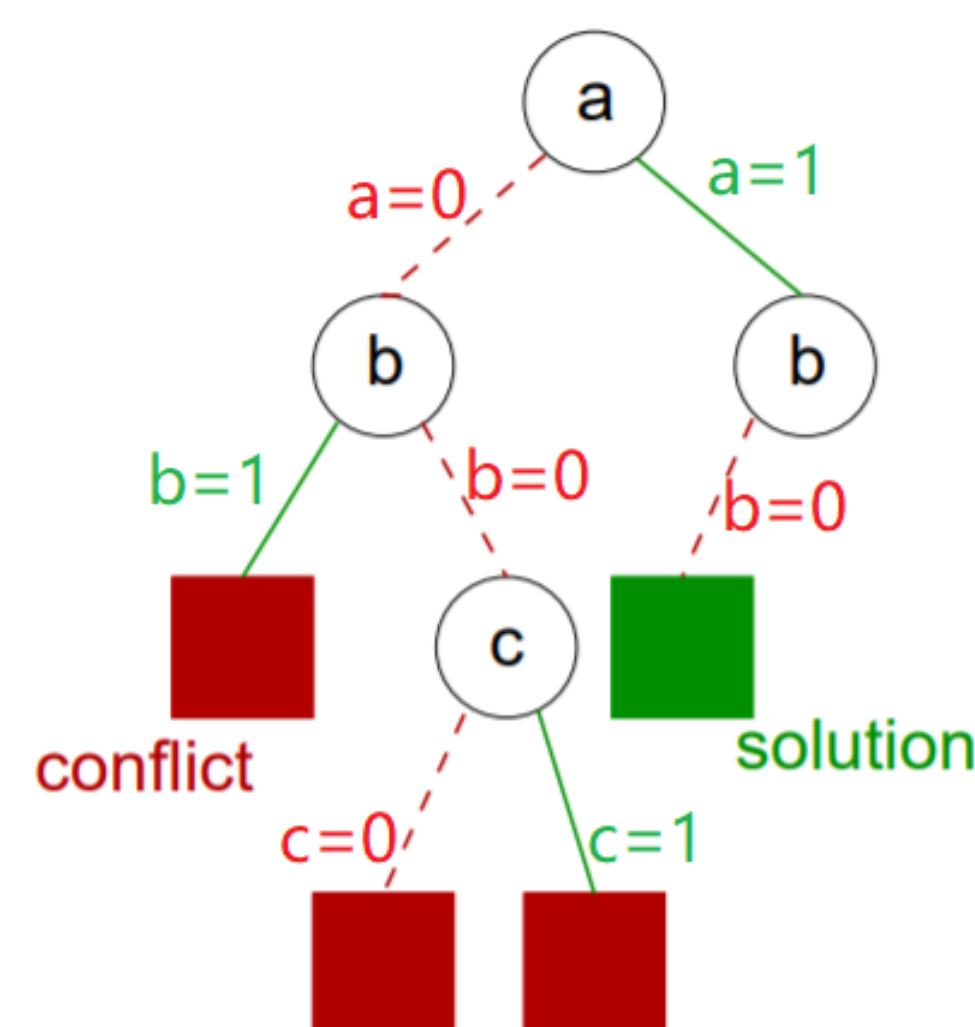


SAT can represent problems in motion planning problems from robotics

Previous Methods

- Searching with backtracking

$$\begin{aligned} \varphi = & (a \vee \neg b \vee d) \wedge (a \vee \neg b \vee e) \wedge \\ & (\neg b \vee \neg d \vee \neg e) \wedge \\ & (a \vee b \vee c \vee d) \wedge (a \vee b \vee c \vee \neg d) \wedge \\ & (a \vee b \vee \neg c \vee e) \wedge (a \vee b \vee \neg c \vee \neg e) \end{aligned}$$



- Local search
Flip one bit "greedily" with most satisfied constraints.

Motivation

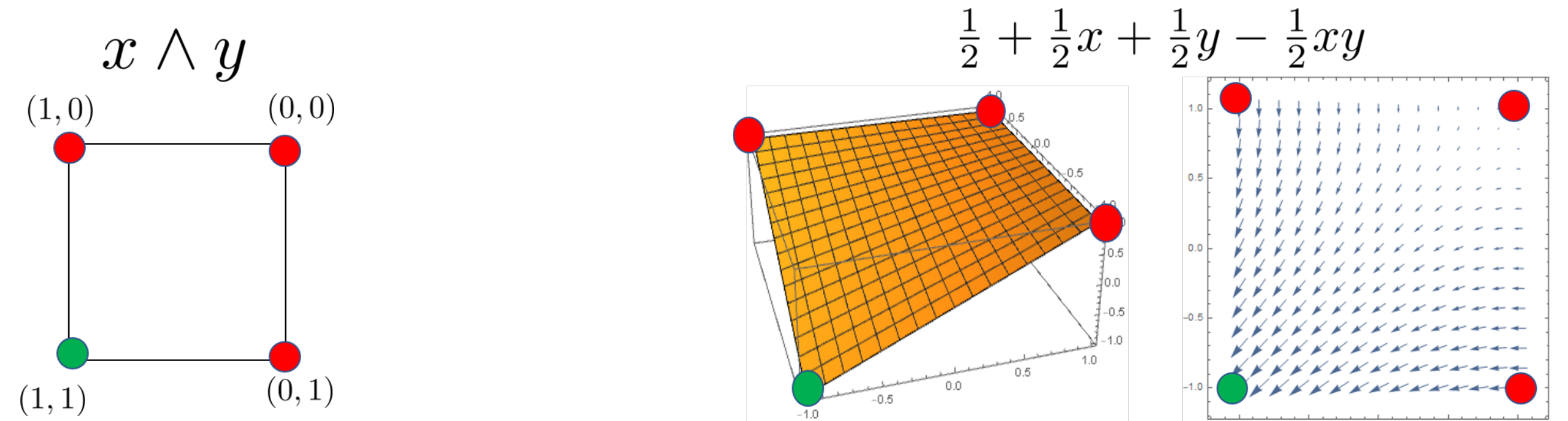
There is no general versatile SAT solver that can handle different types of constraints!

References

- Rong Ge, Furong Huang, Chi Jin, Yang Yuan Escaping From Saddle Points — Online Stochastic Gradient for Tensor Decomposition
- Simons Institution — Bridging Continuous and Discrete Optimization Aug. 16 — Dec. 15, 2017

From Discrete to Continuous

By Fourier transform, Boolean formulas can be expanded to multilinear polynomials. Gradient can be computed and continuous optimization methods can be applied.



Let C be the set of constraints and $|C| = m$,

$$Obj(x) = \sum_{c \in C} \text{FourierExpansion}(c)$$

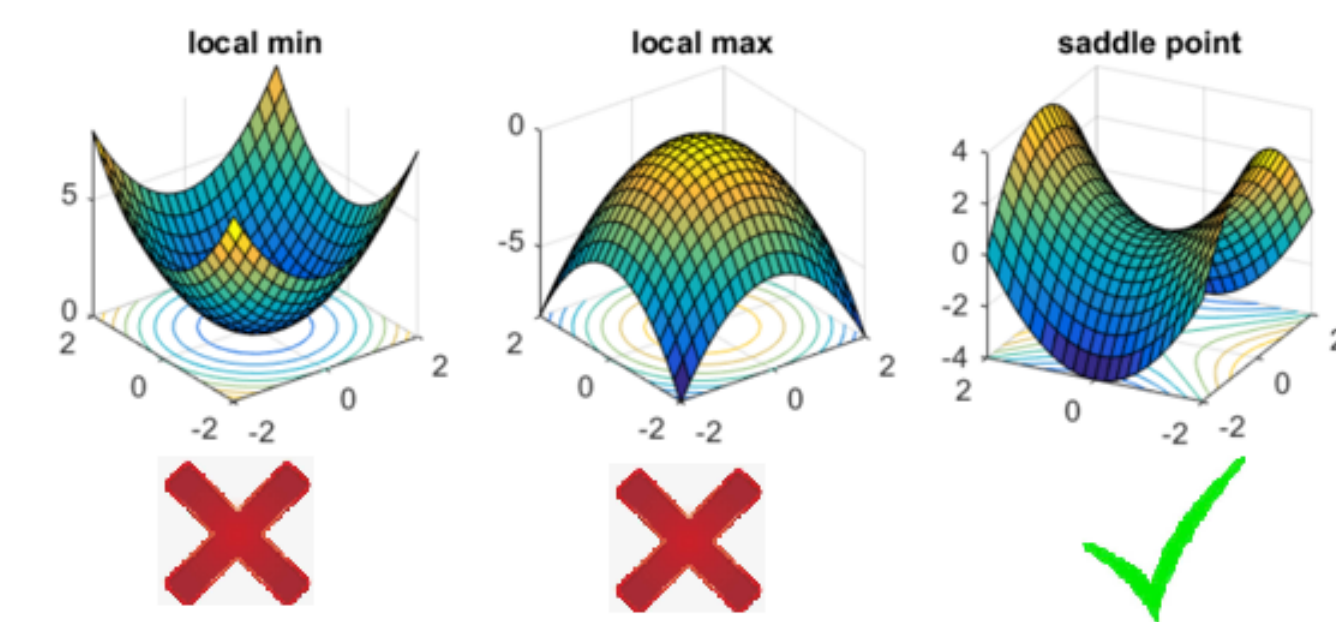
Reduction from SAT to Minimization

Theorem 1

$$c \text{ is satisfiable} \Leftrightarrow \min_{x \in [-1, 1]^n} \text{FourierExpansion}(x) = -1$$

$$C \text{ is satisfiable} \Leftrightarrow \min_{x \in [-1, 1]^n} Obj(x) = -|C|$$

The landscape of the curve contains only saddles instead of local min and local max.



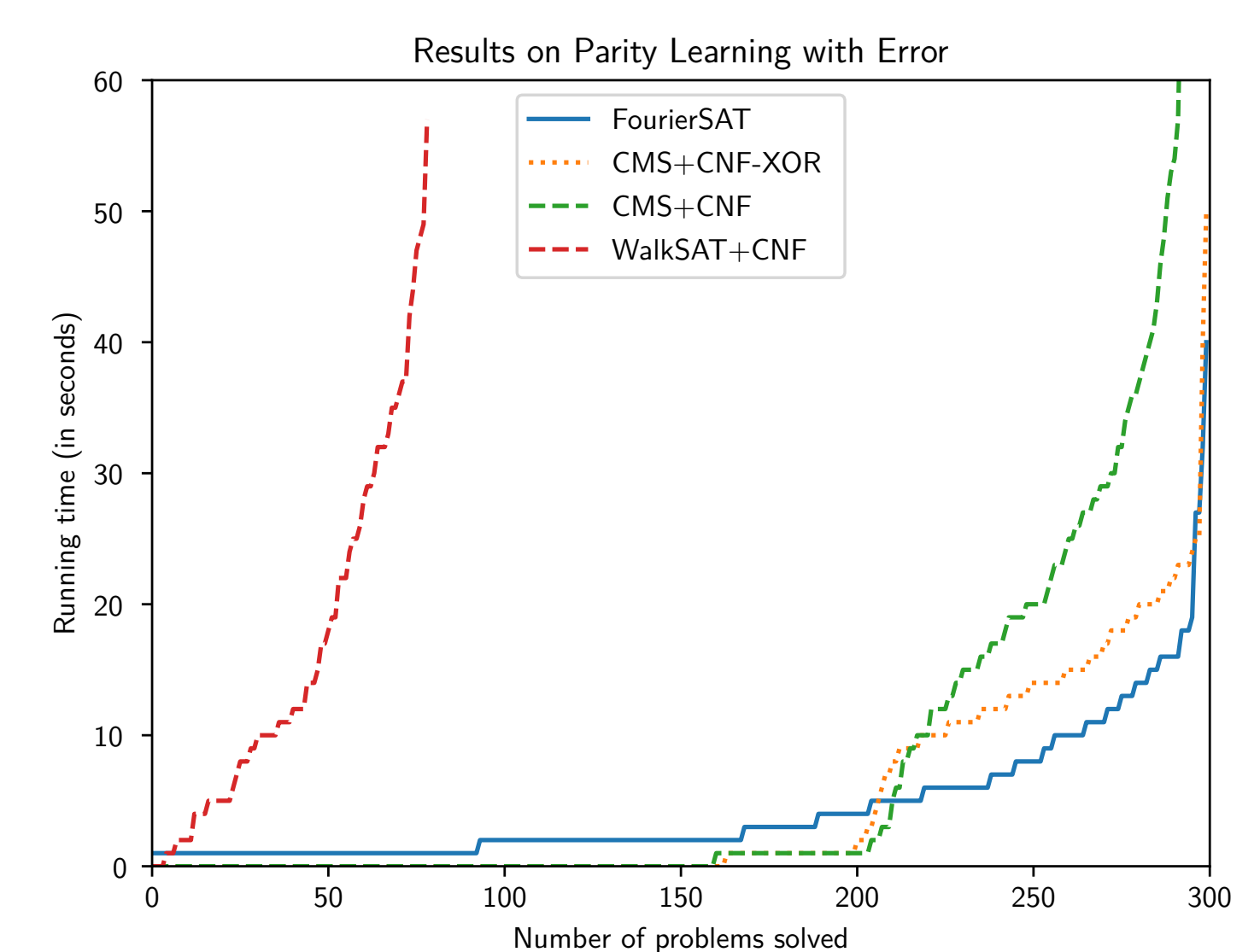
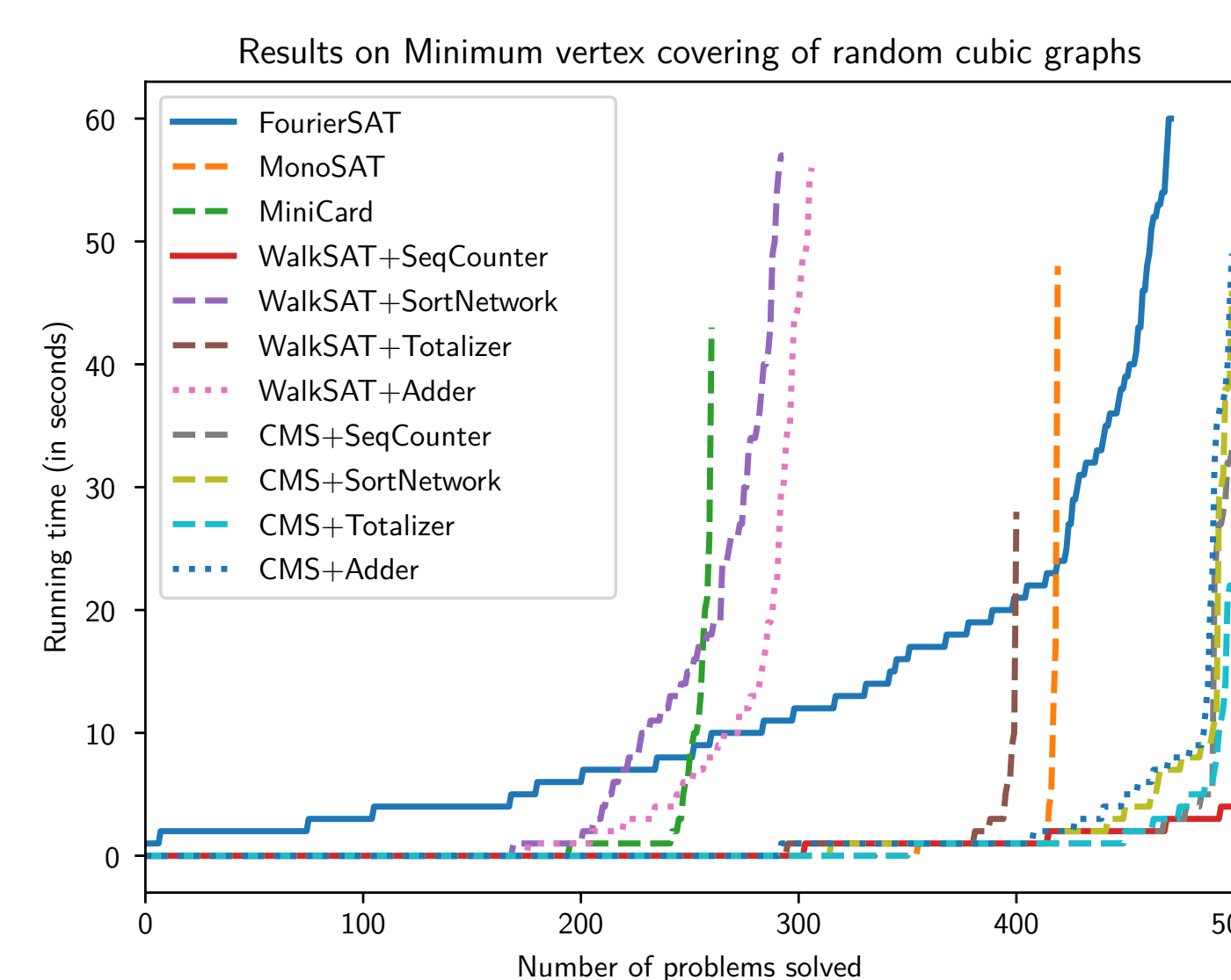
Connecting Continuous and Discrete: Expectation

Theorem 2 Let $a \in [-1, 1]^n$ be a continuous assignment, there exists a randomized rounding R such that $R(a) \in \{-1, 1\}^n$ and $\mathbb{E}[\#_{SAT}(R(a))] = \frac{m - Obj(a)}{2}$

Convergence Rate

Theorem 3 For a formula with n variables and m constraints, Gradient Descent converges to a ϵ -first order critical point in $O(\frac{nm^2}{\epsilon^2})$ iterations.

Experimental Results



Conclusion:

- FourierSAT is a versatile SAT solver with interests in both theory and practice.
- SAT solving by algebraic methods is a promising direction.
- More work on "Bridging Continuous and Discrete Optimization" is desirable.

