Exponential lower bounds for the running time of DPLL algorithms for SAT on satisfiable formulas

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SAT



Known results: exponential lower bounds for specific proof systems

Motivation

- most classical algorithms originating from Davis, Putnam, Logemann, Loveland, 1960-62,
- almost all deterministic SAT solvers use DPLL
- satisfiable formulas are much easier for solvers
- known lower bounds for resolution imply bounds for DPLL on unsatisfiable formulas only.

GOAL: exponential lower bound for satisfiable formulas

DPLL – general scheme



Specific algorithm:

- Heuristic A: chose variable x,
- Heuristic B: choose brunch to be examined first,
- Simplification rules.

Dead ends: trivial formulas.

Examples

A: choose the most frequent variable,

choose a variable from the shortest clause, ...

B: choose the most frequent sign, ...

Simplifications:

unit clause elimination: $x \wedge G \implies G[x \leftarrow 1]$ pure literal rule: \overline{x} does not appear in $F \Rightarrow$ $F \implies F[x \leftarrow 1]$

Known facts

- exponential lower bounds for resolution refutations of <u>unsatisfiable</u> formulas translate to DPLL [Tseytin, 1968],...,[Puldak, Impagliazzo, 2000]
- exponential lower bounds on satisfiable formulas for <u>specific</u> DPLL algorithms:

[Nikolenko, 2002]: Greedy+Unit Clause+Randomization [Achilioptas, Beame, Molloy, 2003]: Greedy+Unit Clause, Ordered DLL (conditional bounds)

[Achilioptas, Beame, Molloy, 2004]: Ordered DLL: exponential time with <u>constant</u> probability

<u>Ultimate goal</u>: Every DPLL algorithm takes exponential time with prob.1 – $2^{-\Omega(n)}$ on satisfiable $F_1, F_2, ..., F_n, ...$

Generalized myopic algorithms

A, B: read $n^{1-\varepsilon}$ clauses,

read other clauses without negations, query the number of occurrence of a literal Simplifications: unit clauses, pure literals.

Drunk algorithms

- A: any !
- **B**: random 50 : 50.

Simplifications: unit clauses, pure literals, subsumption.

<u>Theorem</u>: \exists sequence of (polynomial-size) satisfiables formulas F_n (*resp.*, G_n) such that \forall polynomial-time randomize generalized myopic (*resp.*, *drunk*) algorithm errs with probability $1 - 2^{-\Omega(n)}$.

<u>Proof strategy:</u> show that w.h.p. a DPLL algorithm obtain a hard <u>unsatisfiable</u> formula:



Construction for drunk algorithms Take any hard unsatisfiable F $F[x_i \leftarrow 0]$ remains hard $(F \lor x_1) \land (F \lor x_2) \land \dots \land (F \lor x_n) = G$ first move is wrong with probability $\frac{1}{2}$. Take n renamed copies $G^{(1)} \wedge G^{(2)} \wedge \dots \wedge G^{(n)}$ \Rightarrow wrong move with probability $1 - \frac{1}{2^n}$.

Construction for generalized myopic algorithms

- 1) construct $\mathbf{n} \times \mathbf{n}$ 0/1- matrix **A** that
 - is non-degenerate,
 - has 3 nonzero entries per row,
 - has certain expansion properties.

(solution: take a larger matrix at random, select n linearly independent rows)

2) take 0/1 vector **b** at random

3) convert **Ax=b** into 3-CNF (with unique satisfying assignment)

 $x + y + z = 1 \Leftrightarrow (x \lor y \lor z) \land (\overline{x} \lor \overline{y} \lor z) \land (\overline{x} \lor y \lor \overline{z}) \land (x \lor \overline{y} \lor \overline{z})$

Myopic: idea of the proof



If b and b' are similar solution may differ much.

Open question:

Generalize the model !

- myopic simplifications rule,
- oracle access to the input.