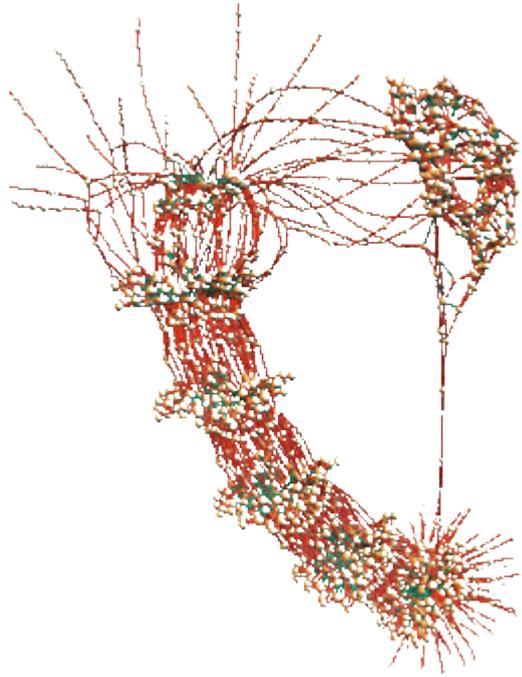


Substructure in SAT



Ryan Williams



Stanford

Two Decades of Significant Progress in SAT Solving

Two major applications:

- Checking programs/circuits for bugs
 - Finding exploits in software
- (“does there exist an input which will yield the following undesired behavior?”)



Many designs can be checked completely by

- reducing the “bug finding” problem to a huge CNF-SAT instance
(e.g., 1 million variables and 5 million clauses)
- checking UNSAT with a solver

A Huge Theory-Practice Gap

The performance of modern solvers seems to defy the theoretical claim that SAT is hard!

Practice: SAT instances that arise from a wide variety of domains are *easy*, more often than not!

- **The unreasonable effectiveness of the Cook-Levin Theorem**

Theory: SAT should not be easy... but it's not impossible

- **Fastest worst-case 3SAT algorithm [Hertli'11]: $O(1.308^n)$ time**
- **Exponential Time Hypothesis [IPZ'01]**
 - 3SAT requires $\Omega((1 + \varepsilon)^n)$ time, for some $\varepsilon > 0$
- **Strong Exponential Time Hypothesis [CIP'09]**
 - For all $\varepsilon > 0$ there is a k such that k SAT needs $\Omega((2 - \varepsilon)^n)$ time

A Huge Theory-Practice Gap

The performance of modern solvers seems to defy the theoretical claim that SAT is hard!

How can we bridge the gap?

There is *tractable* substructure in real-world problems
But what structure? How do we quantify it?

Selman's World

Bart Selman:

'Our world may be "friendly enough" to make many typical reasoning tasks poly-time --- challenging the value of the conventional worst-case complexity view in CS.'

We can formalize what "friendly enough" means, and ask precise questions about "how friendly" tasks can be, while remaining in a "worst-case complexity" perspective

Outline

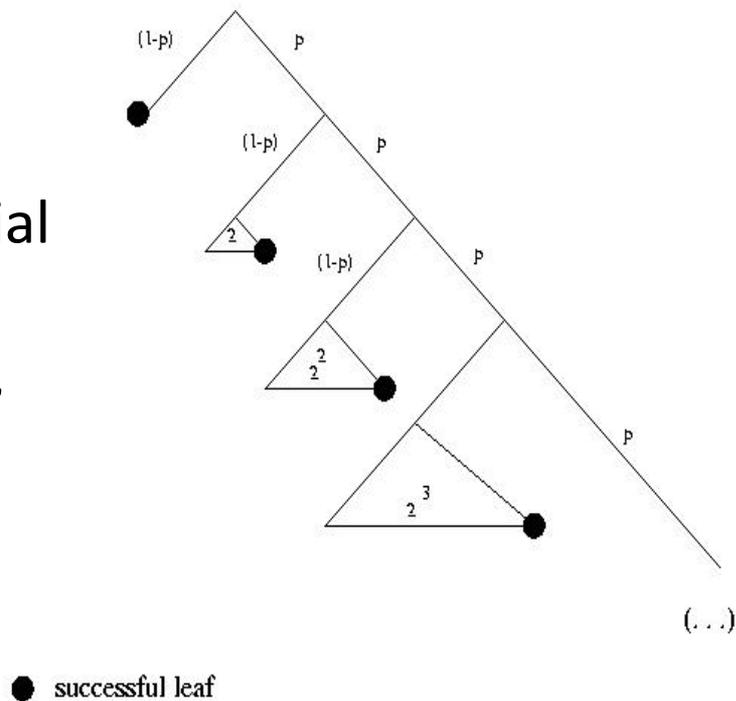
- The Origins of Backdoors
- Intuition and Properties of Backdoors
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- Backdoors in Practice
- Related Work
- Final Thoughts

The Origin of Backdoors

Heavy-Tailed Running Time Distributions

Many diverse instances of combinatorial search problems, when solved by randomized backtracking algorithms, yield a runtime distribution that empirically looks “heavy-tailed”

[Gomes-Selman-Crato-Kautz'00]



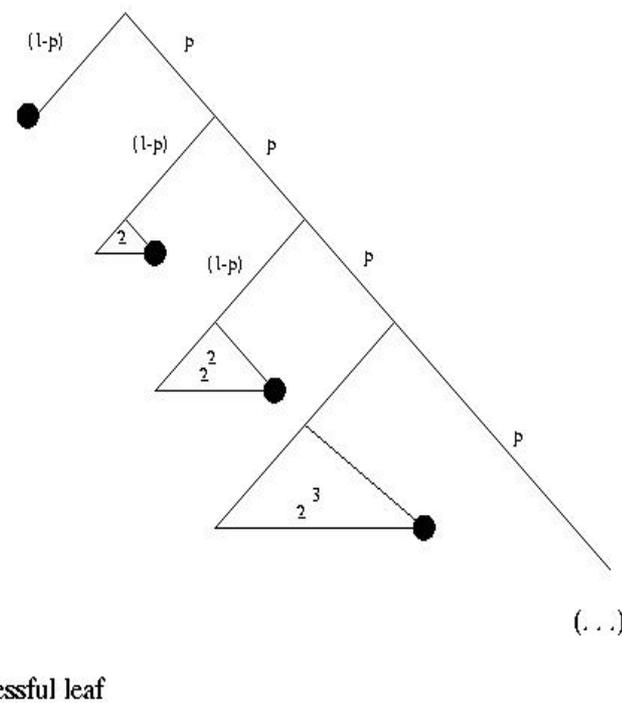
$\Pr[\text{Running time is at least } T] \sim 1/T^\alpha$
where α is a small positive constant

The Origin of Backdoors

Chen, Gomes, Selman '01

Consider a SAT instance F and branching solver S with the following properties:

1. There is one “special” variable x in F
2. Solver S chooses x with probability $1-p$
3. If S chooses the variables y_1, \dots, y_k, x , then S runs for 2^k steps



Then, $\Pr[(\text{Runtime of } S) \geq 2^k] = p^{k+1}$

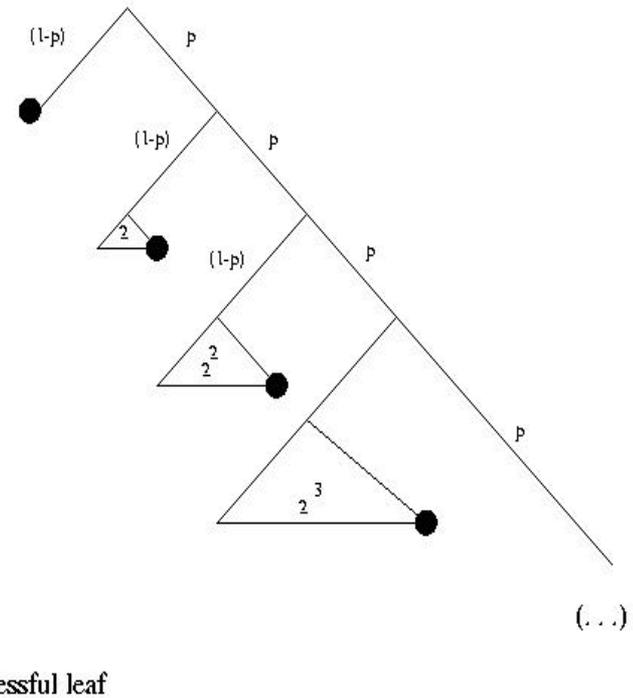
When $p \sim \frac{1}{2^\alpha}$, have heavy-tailed running time

The Origin of Backdoors

Heavy-Tailed Running Time Distributions

There did not seem to be universal agreement about whether the runtime distributions are truly heavy tailed

But there *is* universal agreement that quick restarts of a SAT solver can be remarkably effective!



How to explain short runs?

Backdoors to Tractability [WGS'03]

Informally:

- A **backdoor** to a given SAT instance is a subset of variables such that, once assigned appropriately, the remaining instance lies within a tractable subset of SAT
- The **entire set** of variables is always a backdoor set...
The primary question is: when do *small* backdoors exist?

More formally:

- We define the notion of a “subsolver”
(handles the tractability of problem instance)
- distinguish two types: *backdoors* and *strong backdoors*

Subsolvers (Polytime Heuristics)

Def. A **subsolver** A is an algorithm that, given any formula F , satisfies:

(Trichotomy) $A(F) \in \{\text{SAT}, \text{UNSAT}, \text{DK}\}$ and never errs

(Efficiency) A on F runs in $\text{poly}(|F|)$ time

(Triviality) On the formula F with no clauses, $A(F) = \text{SAT}$
On every F with an empty clause, $A(F) = \text{UNSAT}$

(Self-reducibility) If $A(F) \neq \text{DK}$, then for every variable x of F ,
 $A(F[x=0]) \neq \text{DK}$ and $A(F[x=1]) \neq \text{DK}$

Canonical example:

$A(F) =$ if F is 2CNF/Horn/anti-Horn then output the answer
else output **DK**

The only non-trivial property is self-reducibility:

2CNF and Horn formulas are closed under variable substitution

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1. **Definition is general enough to encompass many polynomial time constraint propagation methods**

(including those for which there does not exist a clean syntactic characterization of the tractable subproblem)

2. **Notion makes perfect sense for other constraint problems:**

e.g., Mixed Integer Programming, Constraint Satisfaction

Backdoor Sets (w.r.t. Subsolvers)

Backdoors (applies to satisfiable instances):

Def. A subset S of variables of F is an ***A-backdoor for F*** if there is an assignment $a_S: S \rightarrow \{0,1\}$ such that $\mathbf{A}(F[a_S]) = \mathbf{SAT}$

Strong backdoors (applies to satisfiable and unsatisfiable instances):

Def. A subset S of variables of F is an ***A-strong backdoor for F*** if for every assignment $a_S: S \rightarrow \{0,1\}$ we have $\mathbf{A}(F[a_S]) \neq \mathbf{DK}$

Backdoors are an *algorithm-dependent* notion

A problem instance may have a “small” backdoor or “large” backdoor depending on which polynomial time heuristics are in the SAT solver

Observation: If $P=NP$ then there exists a subsolver \mathbf{A} such that every SAT formula has an ***A-backdoor*** of size ***zero***

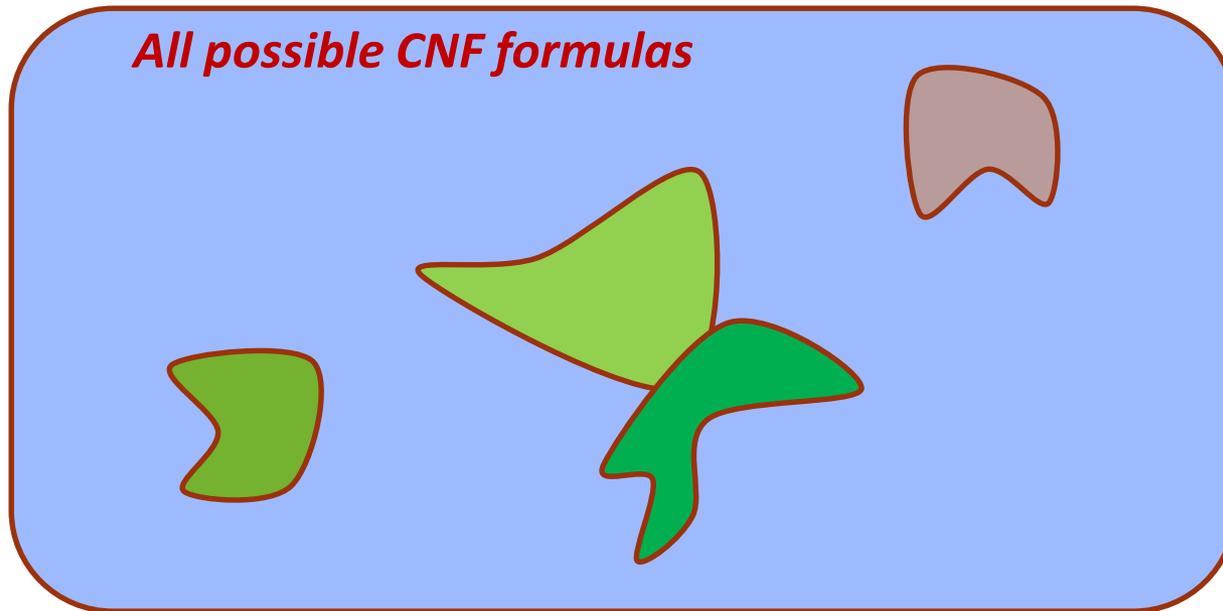
Outline

- The Origins of Backdoors
- **Intuition and Properties of Backdoors**
- Backdoors in Theory
- Backdoors in Practice
- Related Work
- Final Thoughts

Intuition for Backdoors

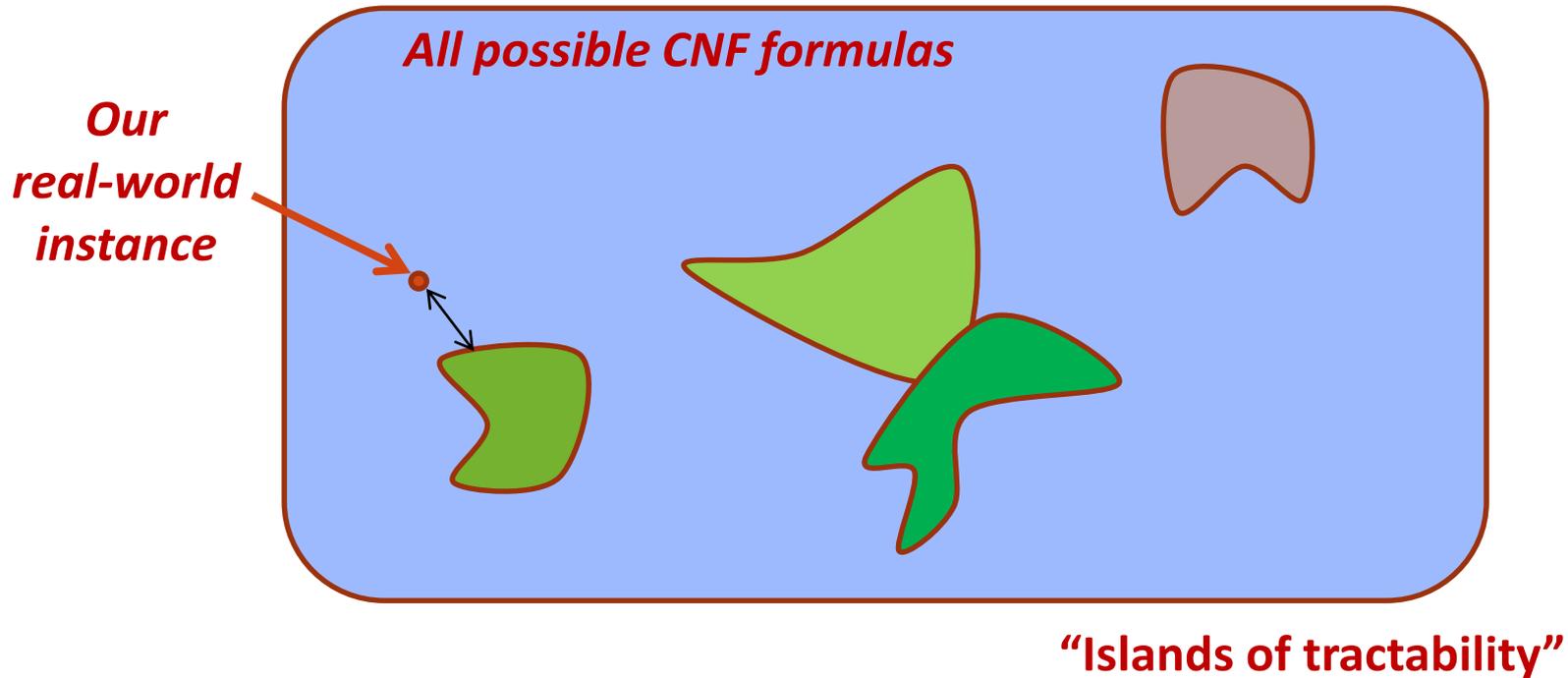
All possible CNF formulas

Intuition for Backdoors



“Islands of tractability”

Intuition for Backdoors



A “small” backdoor set means that the problem instance is “close” to one of these “islands of tractability”
After setting a small number of variables, we arrive at some island

The Importance of Self-Reducibility

Lemma (“Backdoors are monotone”) If S is an A-backdoor for F , then for all $T \supseteq S$, the set T is also an A-backdoor for F

Proof: Suppose F is SAT and S is an A-backdoor for F

Then there is $a_S: S \rightarrow \{0, 1\}$ such that $A(F[a_S]) = \text{SAT}$

That is, a_S can be extended to a SAT assignment a on all variables

Let $a_T: T \rightarrow \{0, 1\}$ be the restriction of a to the set T

(i.e., for all variables x in T , $a_T(x) = a(x)$)

By self-reducibility, if $A(F[a_S]) = \text{SAT}$ then $A(F[a_T]) = \text{SAT}$ as well

QED

This property seems critical to the utility of backdoors in SAT solvers. One only has to assign the backdoor variables *at some point* in the branching, rather than having to necessarily choose them first

The existence of small backdoor sets is not tautological!

*Just because a problem instance is solved efficiently in practice, that does not necessarily imply the instance **must** have a small backdoor (w.r.t. the subsolver being used)*

For example: it could be that even the smallest backdoors are “large” but there are many of them, so finding a backdoor is like finding hay in a haystack

Proposition: A “small” backdoor of size B implies that there are at least $\binom{n-B}{k}$ backdoors of size $k + B$

Possessing small backdoors is a stronger condition than possessing many backdoors

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Almost all formulas *don't have* small (weak or strong) backdoors

Theorem: Let \mathbf{A} be a subsolver handling 2-SAT or Horn-SAT
Whp, for sufficiently large d (*below the k -SAT threshold*)
a random k -SAT instance with n variables and dn clauses has
minimum \mathbf{A} -backdoor size at least cn

Intuition: With high probability, a backdoor set of variables must
“hit” many clauses in order to simplify a *random* k -CNF instance
enough to become Horn or 2-SAT

So for “*almost all*” instances, there is no small backdoor set with
respect to these natural subsolvers.

(This could also explain why randomized backtracking performs
poorly on large enough random 3-SAT instances)

*The existence of small backdoors in a problem instance means
that it is “far from random”*

Every satisfiable k-CNF formula has a backdoor of “nontrivial” size

Theorem [Implicit in PPZ’99, “Satisfiability Coding Lemma”]

Let A be a subsolver that does unit propagation

(whenever it finds a clause of size 1, it sets the variable)

Every satisfiable k-CNF formula contains a backdoor set (wrt A) of size at most $n(1-1/(2k))$

Furthermore, such a backdoor can be found whp, by simply trying random variable assignments and unit propagation.

Intuition: A $1/(2k)$ -fraction of the variables will be assigned by unit propagation, in expectation

The rest is set to correct values with probability $\geq 2^{-n(1-1/(2k))}$

Corollary k-SAT is solvable in $O(2^{n(1-1/(2k))})$ time

Generic Strategies for Solving SAT with Small Backdoors

Three simple strategies for solving instances with small backdoor sets, that work for all subsolvers

- **A deterministic algorithm**
- **A randomized algorithm**
 - *Provably better worst-case performance over the deterministic one*
- **A heuristic randomized algorithm**
 - *Assumes existence of a good heuristic for choosing variables to branch on*
 - *We believe this is close to what happens in practice*

Easy SAT algorithm for small backdoors

For increasing $k=1,2,\dots$

Try all k -sets S of variables and all possible Boolean assignments to S .

If the subsolver outputs SAT on some S ,
output “SAT”

If there is an S for which the subsolver outputs
UNSAT on all assignments to S , output “UNSAT”

*When there is a backdoor of size k ,
takes about $O\left(\binom{n}{k} 2^k\right)$ calls to the sub-solver*

Randomized algorithm

Idea: Try to backtrack on a superset of t variables that *contain* the backdoor set of size k

Assume a backdoor of size k .

A randomly chosen t -set of variables contains the backdoor, with probability at least $\binom{n-k}{t-k} / \binom{n}{t} \geq \binom{t}{k} / \binom{n}{k}$

Pick such a set and try all 2^t assignments with the subsolver.

Repeat for $2 \binom{n}{k} / \binom{t}{k}$ times; takes about $2^t \binom{n}{k} / \binom{t}{k}$ calls.

When $2^k \binom{t}{k} > 2^t$ then this strategy is faster

For example, if $t = 2k$ then $2^k \binom{2k}{k} > 7^k > 2^{2k}$

**OPEN: What is the *optimal* randomized strategy?
(Count only the number of calls to the subsolver)**

Heuristic Randomized Algorithm

Assume we have:

DFS, a generic **depth first search** randomized backtrack search solver with:

- *(polytime) subsolver* **A**
- **Heuristic** **H** that (randomly) chooses variables to branch on, in polynomial time
 - **H** has probability $1/h$ of choosing a backdoor variable ($h > 1$ is a fixed constant)

Call this ensemble **(DFS, H, A)**

Heuristic Randomized Algorithm

Theorem [WGS'03]

If the minimum A-backdoor for F has size $O(\log n)/(\log h)$, then (DFS, H, A) has a restart strategy that solves F in polynomial time.

If there is a small backdoor,

then (DFS, H, A) has a restart strategy that runs in polynomial time.

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Backdoors in Practice

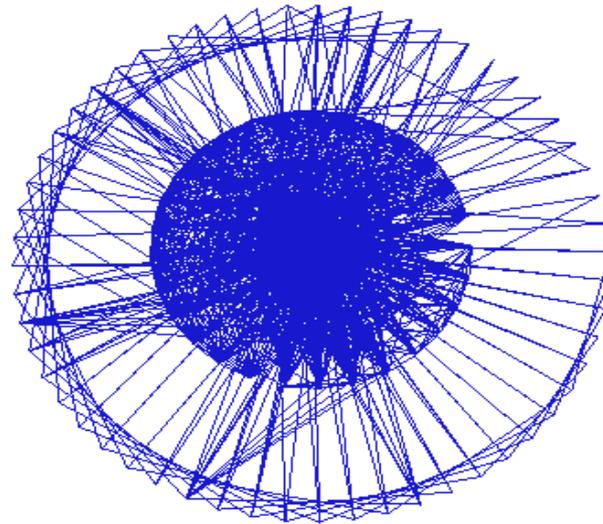
instance	# vars	# clauses	backdoor	fract.
logistics.d	6783	437431	12	0.0018
3bitadd_32	8704	32316	53	0.0061
pipe_01	7736	26087	23	0.0030
qg_30_1	1235	8523	14	0.0113
qg_35_1	1597	10658	15	0.0094

Subsolver used here: the SATz heuristics

A great deal of follow-up work since the initial experiments!

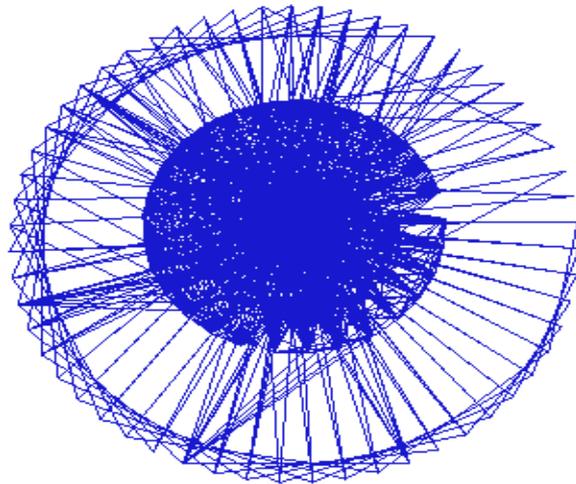
[Survey by Gomes, Kautz, Sabharwal, and Selman '07]

A Dynamic View (Bart's Movies)



***Variable-Variable Graph of an UNSAT Instance
as a SAT solver is being run on it
(random selection of variables to branch on)***

A Dynamic View (Bart's Movies)



*Graph when SAT solver backtracks
directly on strong backdoor of UNSAT instance
(variables chosen by heuristics of solver)*

Backdoors can help explain why QBF is still hard in practice

Recall QBF = Quantified Boolean Formulas

e.g. $(\exists x)(\forall y)(\exists z)((x \text{ AND NOT}(y)) \text{ OR } z)$

With QBF, the order of the quantified variables is critical: one can't just pick any old variables to branch on

If the presence of small backdoor sets are helping SAT solvers work well, this makes sense:

In SAT, you can branch on any desired variable, so small “bottleneck” variables can be eliminated early in search

(Note: Samer and Szeider have a notion of backdoor sets for QBF)

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Related Work

1. Operations Research [Crama, Ekin, Hammer '97]
Control sets: Small sets of variables for a formula that, once those variables are deleted/set to the right value, the resulting formula has some “nice property”
2. Parameterized algs [Guo, Hueffner, Niedermeier '04]
Distance from triviality: Suppose one can make k “edits” to a problem instance so that it’s then easy to solve.
(Presumably such edits preserve the solvability.)
Can we solve the instance in $\mathbf{O(f(k)n^c)}$?

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Final Thoughts

A *backdoor set* of variables tries to isolate the “difficult part” of a problem instance

Since instances in practice are often easy, this part is often small.

Many real-world instances have small backdoors w.r.t. modern SAT solver heuristics, and these solvers do exploit it

A significant question still remains...

Why are the backdoors there?

Are there deeper reasons why these small bottlenecks exist in practice, but not in random instances?

[Hemaspaandra-Williams '12]

Does the *compressibility* of practical SAT instances relate to the sizes of backdoors?

The CNFs encountered in practice are the outputs of highly regular reductions -- and the reductions are given inputs which also highly regular.

Do “compressible solutions” always arise from “compressible instances”?

Does structure imply suboptimality?

Small backdoors for hardware/software verification are typically seen as a very *positive aspect*

But their presence can also indicate *inefficiencies* in the designs of these systems.

(Indeed, SAT solvers can also be used to quickly find security exploits as well!) [Brumley, Engler]

Theory would predict that we must be missing a wide range of efficient and more secure software designs, if everything we verify in practice has such extreme structure.

[W '10,'11,'13] Improved algorithms solving circuit SAT

➔ Circuit complexity lower bounds!

Thank you!