BACKDOORS TO SATISFIABILITY

NEW DEVELOPMENTS IN EXACT ALGORITHMS AND LOWER BOUNDS

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M. S. RAMANUJAN UNIVERSITY OF BERGEN, NORWAY

Outline

Motivation

· 2 perspectives on backdoors

Parameterized algorithms for SAT via backdoors

Satisfiability

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 Best known algorithm for 3-SAT — 1.308ⁿ (Hertli, FOCS 2011)

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 Even for 300 variables, worst case bounds exceed age of the universe. That's all well and good in practice, but how does it work in theory?



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The instances arising in practice must have some structure!

 'Complete' SAT solvers are variants of the DPLL algorithm.

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DPLL= Davis-Putnam-Logemann-Loveland

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Horn formulas : formulas with at most one positive literal in every clause, solved just by unit propagation.

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$$\downarrow$$
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Select a variable (based on some heuristic) and explore both assignments.



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Modern Sat solvers

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· 'Learning' clauses.

· 'watching' literals for fast unit propagations.





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- Fixing an assignment to this set propagates to the rest of the variables.
- Lots of real world instances seem to have a small set on which the remaining variables are dependent.
- Can we capture the structure of an instance through this small set of variables ?

Backdoor sets

Introduced by Williams, Gomes, Selman (IJCAI 2003) and Crama, Ekin, Hammer (D. A. M. 1997)

Informally, a set of variables whose instantiation results in a significantly simplified formula.

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For every tractable base class for SAT, we have a sub-solver that solves instances in this class and rejects the rest. Subsolver1: Solve all 2cnf formulas and reject the rest.

Base class = sub-solver

• Weak Backdoor to C

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Some assignment leads to a satisfiable instance in the base class C

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• Strong Backdoor to C



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Strong Backdoor to C



Some assignment leads to a satisfiable instance in the base class C



All assignments lead to an instance in the base class C.



If the instance is satisfiable then every strong backdoor is also a weak backdoor!



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wbd ≤ sbd

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- (b) (Crama, Ekin, Hammer) backdoor sets provide an excellent framework to extend tractability results for SAT.

eg. SAT is in P for 2-cnf formulas—> SAT is in P for formulas with a strong backdoor of size 10 to 2-cnf.

Islands of Tractability


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Think of the base classes as `Islands of tractability'.

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Objective: If an instance is close to an island of tractability, then we can solve it efficiently.

Research Agenda



Instances with a backdoor of size log² n to an island

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Instances with a backdoor of size c to an island

 One approach: Find a weak/strong backdoor to a base class, explore all assignments to the backdoor variables.

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How do we detect that an instance is `close' to an island of tractability?

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 For which islands can we do this detection efficiently (in polynomial time)?

 For any reasonable island of tractability, detecting if an instance is `close' to this island is NP-complete.

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 For which islands can we do this detection efficiently (for a relaxed notion of efficient)?

Finding Backdoors

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 How to develop and analyze `efficient' algorithms to detect small backdoors?

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Fixed-Parameter Algorithms!

• Parameterized problems - (x,k); k is the parameter.

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FPT

• Running time $f(k) |x|^c$ implies that for k bounded by $f^{-1}(poly(n))$, we have a poly time algorithm.

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• W-hierarchy: $FPT \subseteq W[1] \subseteq W[2] \subseteq ... \subseteq XP$

Rest of this talk

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 Recent advances in FPT algorithms for computing backdoors to some base classes (q-Horn, tw-SAT, composite classes)

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 Recent advances in FPT algorithms for computing backdoors to some base classes (q-Horn, tw-SAT, composite classes)

 Discuss some interesting connections between the 2 perspectives.

Schaefer Classes

Horn (at most one positive lit in each clause)

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Classical sub-solvers

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Nishimura, Ragde, Szeider SAT 2004

Classical sub-solvers

Class Backdoor	Schaefer [Nishimura, Ragde, Szeider 2004]	Unit Prop + Pure Lit. Elim [Szeider 2005]
Weak	W[2]-hard (FPT for 3-cnf)	W[2]-hard (FPT for 3-cnf)
Strong	FPT	W[2]-hard

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- Allows fairly straightforward encodings from <u>Hitting Set/Set Cover</u>, both W[2]-hard parameterized by the size of the solution (the hitting set or the set cover).
- Can change if restricted to 3-cnf formulas.

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F is q-Horn if there is a weight function
w: lit(F)->{0,1/2,1} s.t

 $w(x)+w(\neg x)=1$ and for every clause C, $w(C) \le 1$.

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g-Horn generalizes Horn and 2-cnf. •

w(x)=1 and $w(\neg x)=0$ for all x $w(x)=w(\neg x)=1/2$ for all x

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SAT is in P for
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• What other notions of `distance' can we have?

What about distance through deletion instead of instantiation?

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Deleting x₂

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Deleting x_3

x₃ is a deletion backdoor into 2-cnf.

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x₃ =0

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Deleting X₃

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 $x_3 = 0 (x_1 \vee x_2) \land (\neg x_1 \vee x_5)$

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Deleting x_3

 $\begin{array}{c} x_3 = 1 \quad (x_1 \lor x_2) \land (x_2 \lor x_4) \\ x_3 \text{ is a deletion backdoor into 2-cnf.} \end{array}$



If F is in C, every subformula of F induced by a subset of clauses is in C.

If the base class C is 'clause induced' then S is also a strong backdoor to C.



wbd ≤ sbd ≤del. bd

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Deletion Backdoors

 The detection of strong backdoors being W-hard is not a dead end.

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 If we can detect deletion backdoors then we can still extend the tractable region for SAT.

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- In $O(6^{k} \text{ mn})$ time, we can either conclude no del backdoor of size k or compute a deletion backdoor of size at most 2 k^{2} .

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SAT parameterized by size of deletion backdoor to q-Horn can be solved in time $2^{O(k^2)}$ mn .



SAT is in P for instances with a del. backdoor of size O(√log n) to q-Horn.

SAT is in P for q-Horn

[R. and Saurabh, SODA 2014]

Deletion backdoor set detection to q-Horn can be solved in time $O(12^k\,\,\text{m})$.

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SAT is in P for instances with a del. backdoor of size O(log n) to q-Horn [R. and Saurabh 2014].

SAT is in P for instances with a del. backdoor of size O(Jlog n) to q-Horn [Gaspers, Ordyniak, R., Saurabh, Szeider 2014].

SAT is in P for q-Horn

A linear time algorithm for SAT instances
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- Corollary: Deletion backdoor detection for RHorn can be done in time O(4^k m).

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 `close' to being q-Horn.
- Corollary: Deletion backdoor detection for RHorn can be done in time O(4^k m).
- A further consequence of this algorithm: the first linear time FPT algorithm for Odd Cycle Transversal (open problem of Reed, Smith and Vetta, 2003).

Backdoors to Bounded Treewidth SAT

Modeling CNF-formulas as graphs



If the Incidence graph is a forest then SAT is in P (Fischer, Makowsky, Ravve 2008).



What about formulas with small backdoors to Acyclic SAT? Is SAT tractable on these formulas?



Gaspers and Szeider (ICALP 2012):

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weak backdoor detection to acyclic SAT is W[2]-hard.

weak backdoor detection to acyclic 3-SAT is FPT.

strong backdoor detection to acyclic SAT is FPT-approximable.

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In FPT time, either conclude there is no strong backdoor of size k or compute a strong backdoor of size 2^k

If the Incidence graph is tree-like then SAT is in P (Fischer, Makowsky, Ravve 2008).



Gaspers and Szeider (FOCS 2013):

Strong backdoor detection to tw SAT is FPT-approximable.

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SAT parameterized by size of sbd to tw SAT is FPT.

Running time : $2^{2^k} n^3$



SAT is in P for instances with sbd of size O(log log n) to tw SAT.

tw SAT is in P.

Fomin, Lokshtanov, Misra, R., Saurabh (SODA 2015)

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3-SAT parameterized by k=min{sbd,wbd} to tw 3-SAT can be solved in time $2^{O(k)}$ m.

This running time is optimal both w.r.t parameter and input-size.



Combining the perspectives

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• This algorithm: revisit this perspective.

· Apply UP and PLE

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DPLL' is an FPT algorithm for 3-SAT par by min{sbd,wbd} to tw 3-SAT.

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But intuition remains the same: Remove 'irrelevant' parts of the formula or at the very least replace them with a `small' equivalent formula.

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Optimal running time (parameter and i/p size)

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Optimal running time (parameter and i/p size)

 Again, techniques developed here have other applications: improving several kernelization and FPT algorithms to linear time.

Composite Base Classes

Consider the following formula.

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land (\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) .. \land (\neg x \lor b_n \lor c_n)$$

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land (\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \ldots \land (\neg x \lor b_n \lor c_n)$$

What is the size of a smallest strong backdoor set into Horn?

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land (\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \land (\neg x \lor b_n \lor c_n)$$

What is the size of a smallest strong backdoor set into Horn?

at least n

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land (\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \land (\neg x \lor b_n \lor c_n)$$

What is the size of a smallest strong backdoor set into Horn?

at least n

What is the size of a smallest strong backdoor set into 2-cnf?

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land (\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \land (\neg x \lor b_n \lor c_n)$$

What is the size of a smallest strong backdoor set into Horn?

at least n

What is the size of a smallest strong backdoor set into 2-cnf?

at least n-1

Consider the following formula.

 $(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) .. \land (\neg x \lor b_n \lor c_n)$

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$$
$$(\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) .. \land (\neg x \lor b_n \lor c_n)$$

Consider F[x=0] ($\neg a_1 \lor \neg a_2 \lor \lor \neg a_n$)

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$$
$$(\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \ldots \land (\neg x \lor b_n \lor c_n)$$

Consider F[x=0] ($\neg a_1 \lor \neg a_2 \lor \lor \neg a_n$)

Consider F[x=1] $(b_1 \vee c_1) \wedge (b_2 \vee c_2) \dots \wedge (b_n \vee c_n)$

Consider the following formula.

$$(x \lor \neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$$
$$(\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \ldots \land (\neg x \lor b_n \lor c_n)$$

Consider F[x=0] ($\neg a_1 \lor \neg a_2 \lor \lor \neg a_n$) Horn

Consider F[x=1] $(b_1 \vee c_1) \wedge (b_2 \vee c_2) \dots \wedge (b_n \vee c_n)$

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Consider F[x=0] ($\neg a_1 \lor \neg a_2 \lor \lor \neg a_n$) Horn

Consider F[x=1] $(b_1 \vee c_1) \wedge (b_2 \vee c_2) \dots \wedge (b_n \vee c_n)$ 2-cnf

$C_{1,C_{2}}$: Horn, 2-cnf



Let $C_{1,...}$, C_r be islands of tractability.

X is a heterogenous backdoor into C_1, \dots, C_r if for every assignment of X, the reduced formula is in some C_i .

 Heterogenous backdoors can be arbitrarily smaller than normal strong backdoors.

 Heterogenous backdoors can be arbitrarily smaller than normal strong backdoors.

 Class of instances with small heterogenous backdoors is a much larger class than instances with small strong backdoor.



Gaspers, Ordyniak, Misra, Szeider, Zivny (AAAI 2014)

Gaspers, Ordyniak, Misra, Szeider, Zivny (AAAI 2014)

1. If H =Horn/dual-Horn \cup 2CNF then detecting heterogenous backdoors to H is FPT

2. For every other combination of Schaefer classes, detecting heterogenous backdoors to H is W[2]-hard.

but FPT for 3-cnf formulas
Archipelagos of tractability

What is the size of a smallest heterogenous backdoor set into Horn u 2-cnf?

What is the size of a smallest heterogenous backdoor set at least 2n into Horn u 2-cnf? Archipelagos of tractability

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \land (q_n \lor r_n)$

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n)$ Consider F[x=1] $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \lor \land (b_n \lor c_n)$

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n)$ Consider F[x=1] $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \lor \land (b_n \lor c_n)$

Archipelagos of tractability $(x \vee \neg a_1 \vee \neg a_2 \vee \neg a_n) \wedge (\neg x \vee \neg p_1 \vee \neg p_2 \vee \neg p_n)$ $(\neg x \lor b_1 \lor c_1) \land (\neg x \lor b_2 \lor c_2) \land (\neg x \lor b_n \lor c_n)$ $(x \lor q_1 \lor r_1) \land (x \lor q_2 \lor r_2) \land (x \lor q_n \lor r_n)$ $(q_1 \vee r_1) \wedge (q_2 \vee r_2) \dots \wedge (q_n \vee r_n)$ Consider F[x=0] (¬p₁ ∨¬p_{2...} ∨¬p_n) ∧ Consider F[x=1] $(b_1 \vee c_1) \land (b_2 \vee c_2) \land (b_n \vee c_n)$

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n)$ Consider F[x=1] $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \lor \land (b_n \lor c_n)$

Consider F[x=0] $\begin{array}{c} (\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land \\ (q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n) \end{array}^{2-cnf} \\ (\neg p_1 \lor \neg p_2 \lor \neg p_n) \land \end{array}$

Consider F[x=1]

 $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \ldots \land (b_n \lor c_n)$

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Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \land \land (q_n \lor r_n)$ Horn $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \land \land (b_n \lor c_n)$

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n)$ Consider F[x=1] $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \lor \land (b_n \lor c_n)$

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n)$ Consider F[x=1] $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \lor \land (b_n \lor c_n)$ 2-cnf

Consider F[x=0] $(\neg a_1 \lor \neg a_2 \lor \lor \neg a_n) \land$ $(q_1 \lor r_1) \land (q_2 \lor r_2) \lor \land (q_n \lor r_n)$ Consider F[x=1] $(\neg p_1 \lor \neg p_2 \lor \lor \neg p_n) \land$ $(b_1 \lor c_1) \land (b_2 \lor c_2) \lor \land (b_n \lor c_n)$

Archipelagos of tractability

 $C_{1,C_{2}}$: Horn, 2-cnf



Run appropriate sub-solver C_i on each part variable-disjoint from the rest Run appropriate sub-solver C_i on each part variable-disjoint from the rest

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X is a split backdoor into $C_{1,...}, C_r$ if for every assignment of X, every connected component of the reduced formula is in some C_i .

A minimal set of clauses which is variable-disjoint from the remaining clauses.



 Split backdoors can be arbitrarily smaller than heterogenous backdoors.

Split backdoors can be arbitrarily smaller than heterogenous backdoors.

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 Class of instances with small split backdoors is a much larger class than class of instances with small heterogenous backdoor.



Ganian, R., Szeider (2014):

If H is a finite set of finite constraint languages, then detecting split-backdoors of the given CSP to H is FPT.

Builds on a combination of traditional FPT tools and new graph separation tools like important separators, sequences and CSP based pattern replacements.

We have seen how backdoors and fixed parameter tractability provide a framework to extend tractability results for SAT based on the `distance' of instances to islands of tractability.

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Stronger definitions of sub-solvers and base classes allowing us to prove tractability for larger classes of instances.

We have seen how backdoors and fixed parameter tractability provide a framework to extend tractability results for SAT based on the `distance' of instances to islands of tractability.

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- Stronger definitions of sub-solvers and base classes allowing us to prove tractability for larger classes of instances.
- Several other variants of backdoors have been proposed, eg. backdoor trees (Samer and Szeider AAAI 2008), learning sensitive backdoors (Dilkina, Gomes, Sabharwal SAT 2009).



So far backdoor sets and variants have provided the best and theoretically most robust explanation for the performances of SAT solvers.

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So far backdoor sets and variants have provided the best and theoretically most robust explanation for the performances of SAT solvers.

What other structural properties of instances are correlated to the computation time and can be effectively formalized in theory?



 So far, `small' backdoors treated as certificates for closeness.

Better measures
than size?

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i.e. backdoors ofpotentiallyunbounded size butwith some structure.



Analysis of existing SAT algorithms in terms of FPT parameteriz ed by backdoors.

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