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From Backdoor Key to Backdoor Completability: Improving a Known Measure of Hardness for the Satisfiable CSP

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• Understanding instance hardness?

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- The notion of backdoor

Intuition

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Definition (Williams, Gomes, Selman 2003)

Let *I* be a satisfiable CSP instance, a backdoor is a subset *B* of variables such that there exists an assignment of the variables in *B* that makes it easy to find a solution for *I*.

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Definition (Ruan, Kautz, Horvitz 2004)

Let *B* be a backdoor for a satisfiable instance *I*. Let $v \in B$ and let *S* a partial solution for $B \setminus \{v\}$.

- v is dependent: exactly one value a in the domain of v such that S ∪ {a} can be extended to a solution for I.
- key fraction of *B*: $\frac{\#\text{dependent variables in }B}{|B|}$

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Hypothesis

higher backdoor key fraction \Leftrightarrow harder instance

Limitation of BKF

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- Case when there is only one solution: \implies BFK always 1.
- Case when flipping the truth assignment of any variable in the backdoor and still extend the backdoor to a solution: ⇒ BFK always 0.

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- Large search space + small solution space
 - \Rightarrow A lot of choices made, few of them good
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Combining both

- Large search space + small solution space
 - \implies A lot of choices made, few of them good
 - \implies The solver takes a long time to find a solution.
- Solution space close to the search space
 - \implies Almost all the choices made are good
 - \implies solver finds a solution quickly.

Completability Ratio

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Completability Ratio

Definitions

Let *B* be a backdoor for *I*. Completability ratio of *B* is $\frac{\#completable}{\#partial}$, with:

- *#partial* the number of partial solutions for *B*.
- #*completable* the number of partial solutions for *B* that can be extended to a solution for *I*.

Backdoor Completability

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Backdoor Completability

Definition

Backdoor completability of a satisfiable instance *I*: average completability ratio of all minimal backdoors of *I*.

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lower backdoor completability \Leftrightarrow harder instance

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What we want

A class C of instances, a solver A.

- C is tractable for A → backdoor completability of all instances in C is high.
- C is not tractable for A → backdoor completability of some instances in C is low.

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Instance classes

- C₁ ⊂ C₂ ⊂ ... where a satisfiable instance is in C_p if the treewidth of its primal constraint graph is bounded by p.
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- The union of all C_p is equal to the whole satisfiable CSP.

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Solvers

- A_1, A_2, \ldots where A_q is based on (q, 1)-consistency.
- The transition between tractability and hardness is sharp: for any class *C_p* and solver *A_q*:
 - $p \leq q \rightarrow A_q$ finds C_p trivial.
 - $p>q
 ightarrow A_q$ fails to solve some instances in \mathcal{C}_p .

so the distinction between low and high completability values should be clear.

Theorem

For any two integers *p*, *q* exactly one of the following is true:

- 1. $p \leq q$ and for every $I \in C_p$, the backdoor completability of I is 1.
- p > q and there is an infinite number of instances in C_p with a backdoor completability exponentially low in the number of variables.

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Ргооf

- 1. From the construction of the solvers.
- 2. For each number *n*, we build an instance in C_p with more than *n* variables and a backdoor completability with regard to A_q asymptotically lower than $\frac{1}{2^n}$.

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- Mistral solver with default configuration
- Standard method to find backdoors

Quasigroup With Holes

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Quasigroup With Holes



Quasigroup With Holes



Pearson CC Pearson Correlation Coefficient RMSE Root Mean Square Error MAE Mean Absolute Error

Random CSP



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Future Research

- Predict hardness? Very expensive to compute : Could it be approximated?
- Generate hard instances?
- Design branching heuristics?

Thank you!