

Cylindrical Algebraic Decomposition for Nonlinear Arithmetic Problems

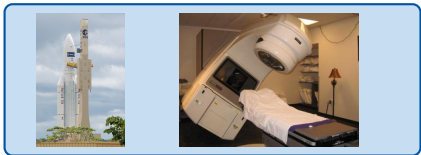
Gereon Kremer
PhD defense talk

12.03.2020

Grateful for the cooperation with and advise from:

Erika Ábrahám, Jens Brandt, Christopher Brown, Florian Corzilius, James H. Davenport, Matthew England, Vijay Ganesh, Rebecca Haehn, Marcel Hark, Einar Broch Johnsen, Sebastian Junges, Viktor Levandovskyy, Jacopo Mauro, Jasper Nalbach, Stefan Schupp, Tarik Viehmann, and many more.

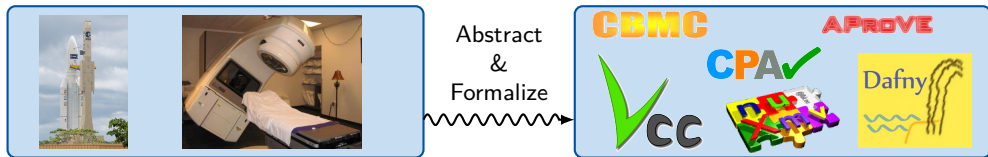
Formal verification (with satisfiability modulo theories)



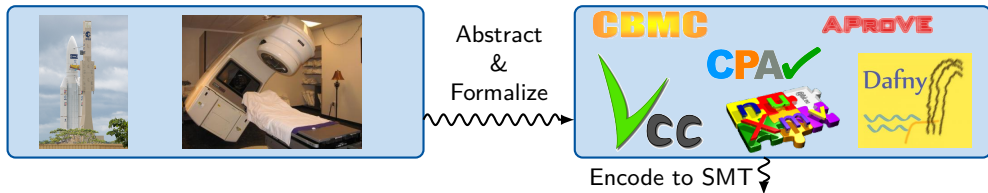
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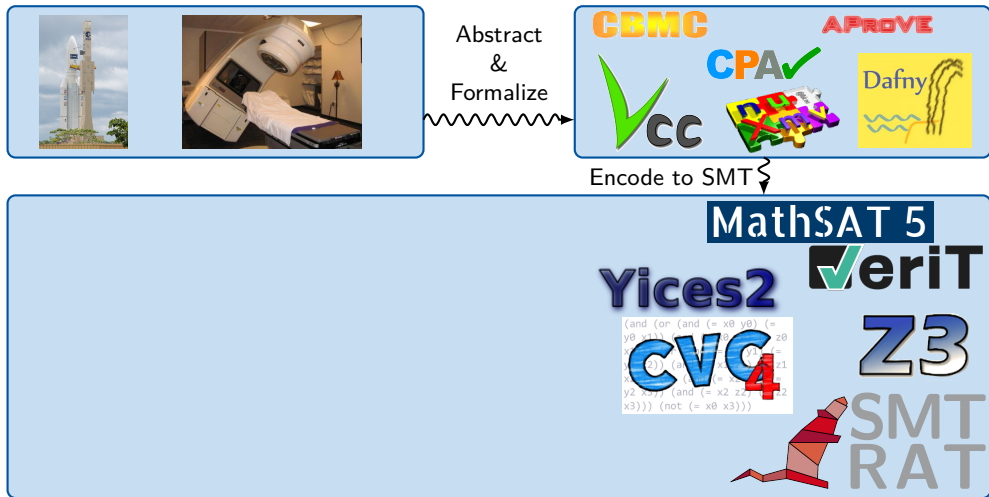
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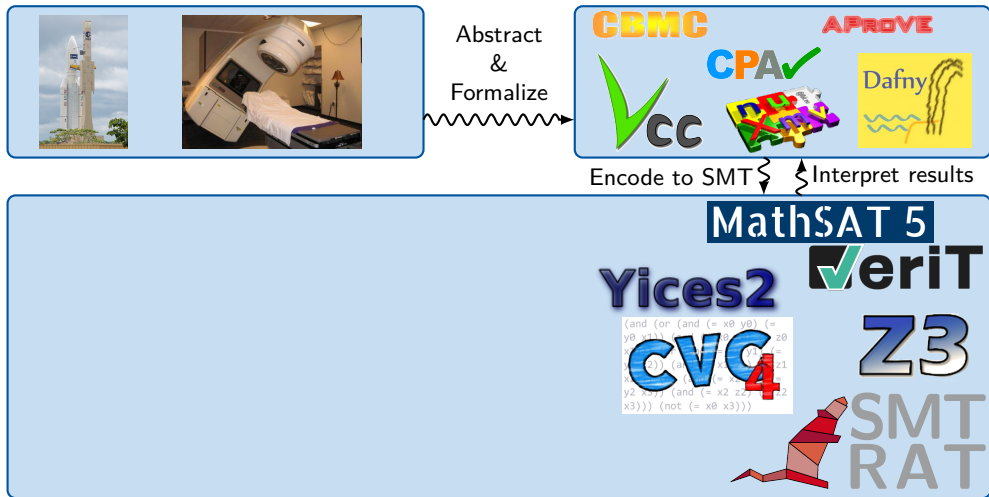
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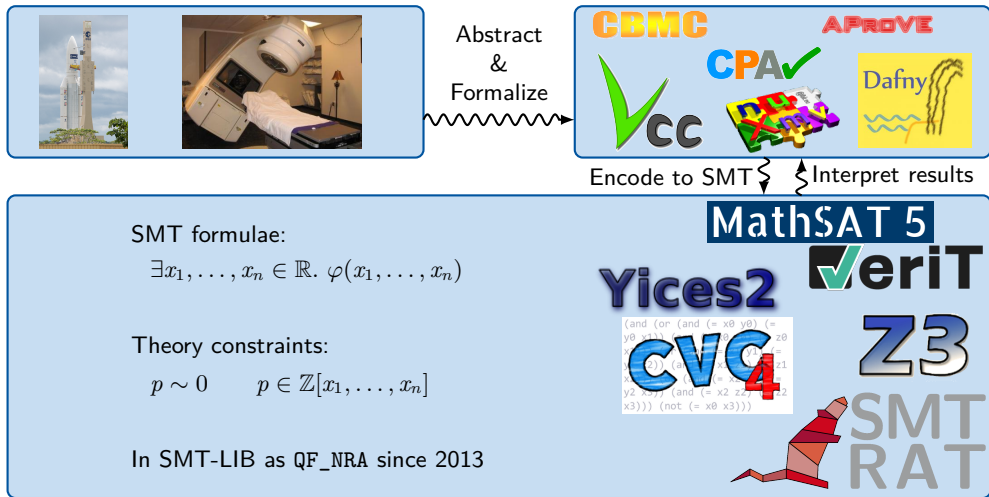
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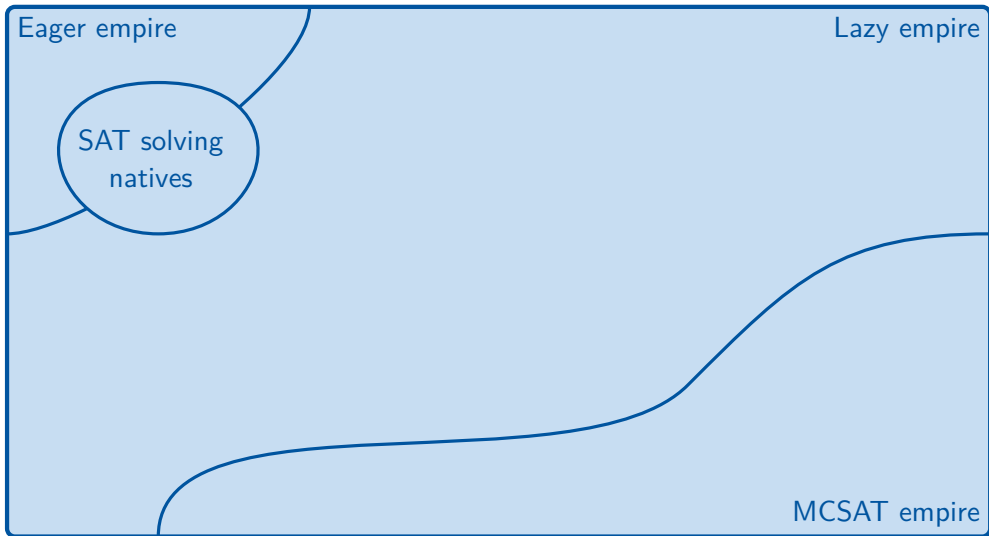
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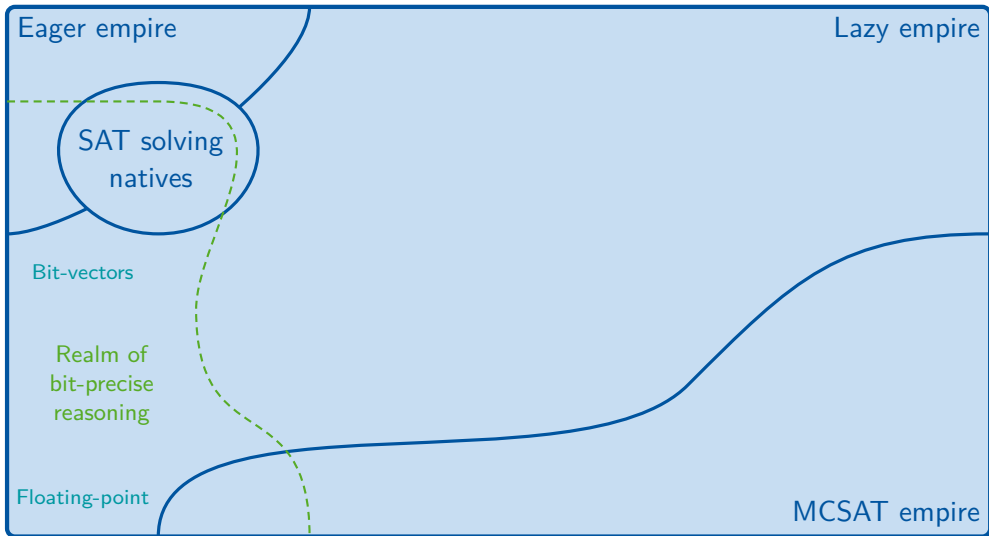
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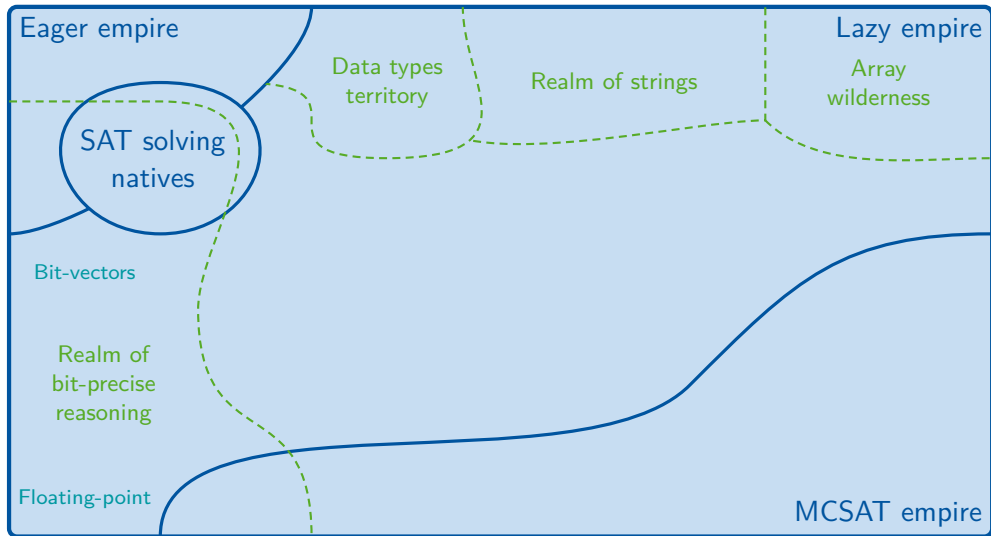
World Map of SMT



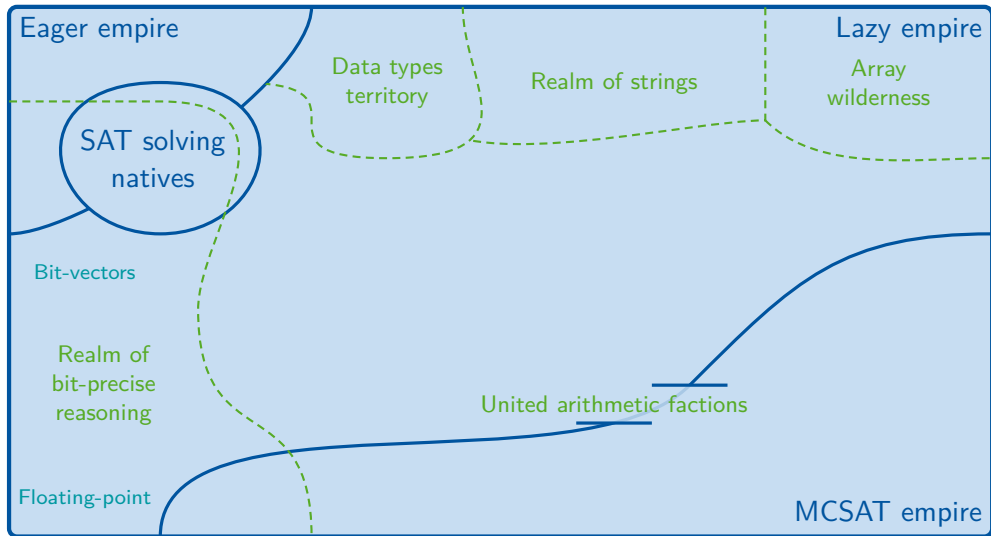
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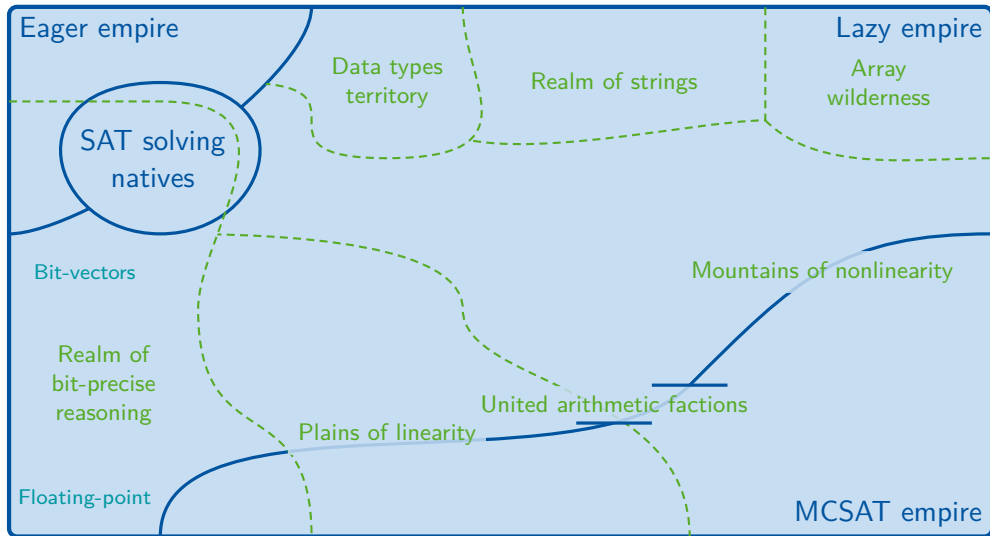
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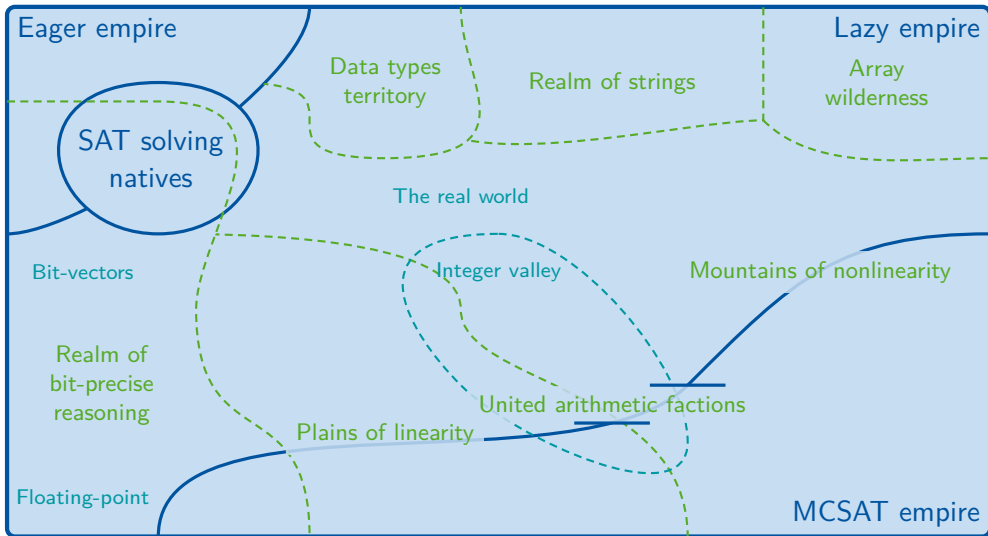
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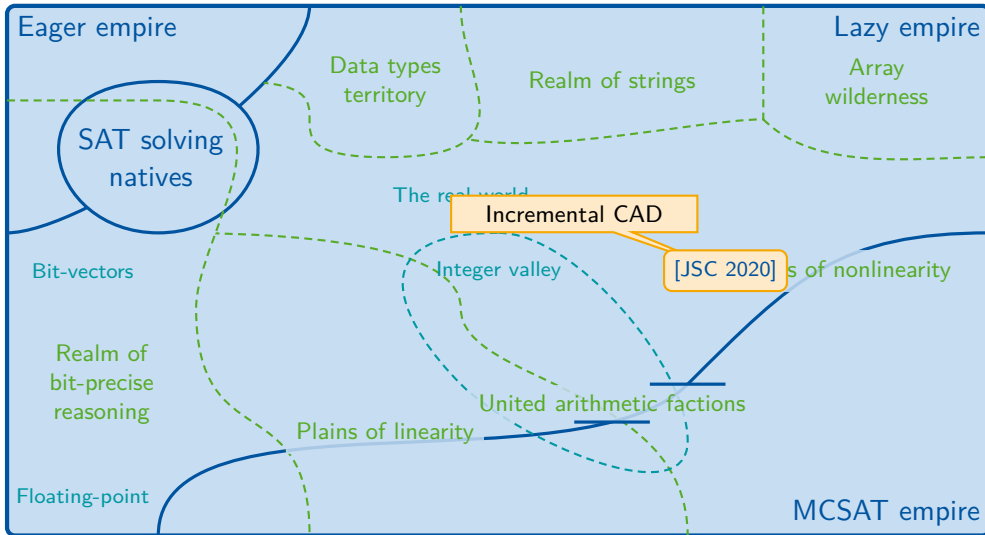
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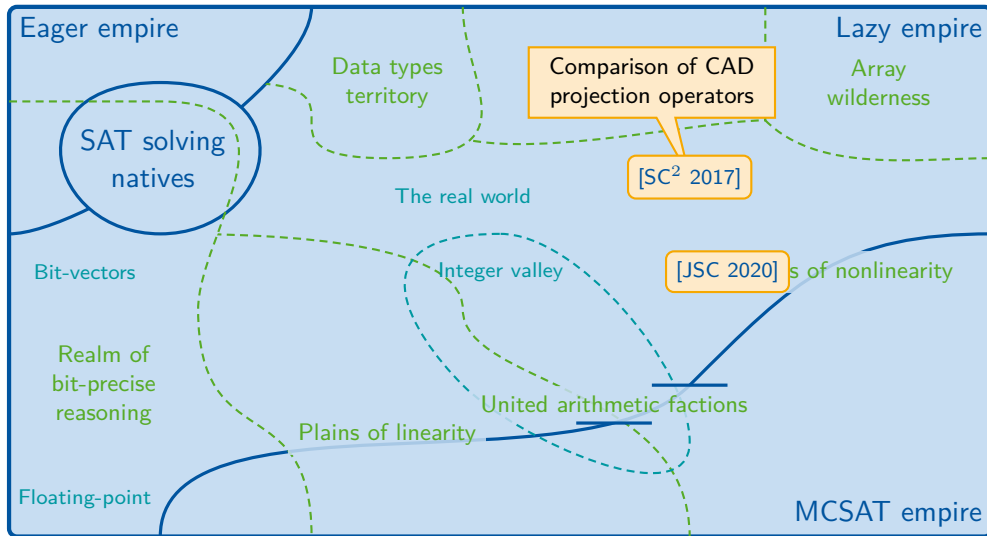
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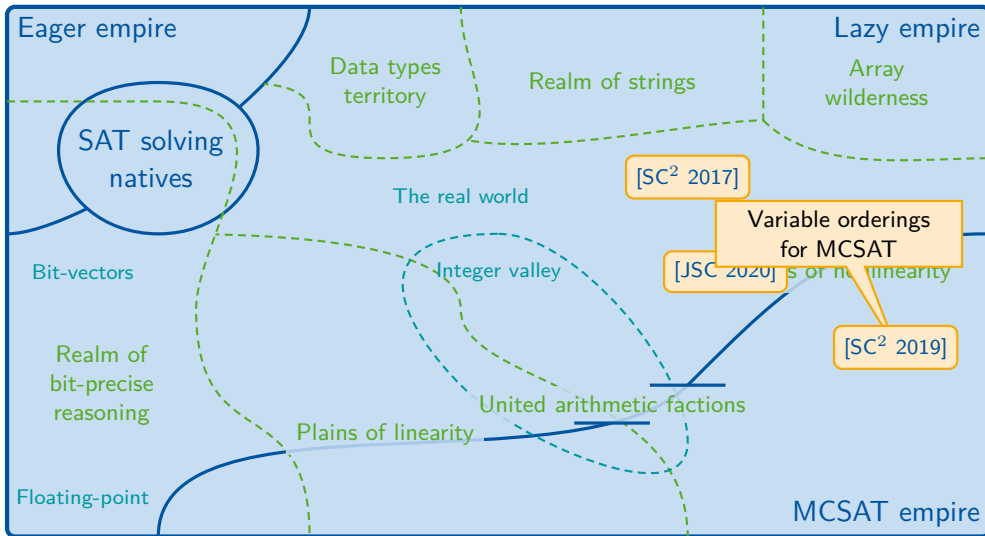
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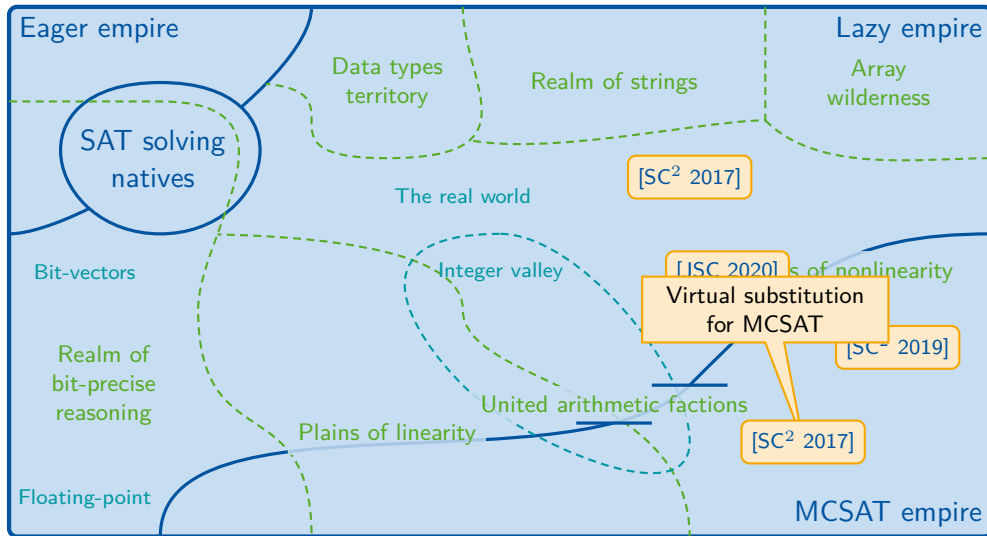
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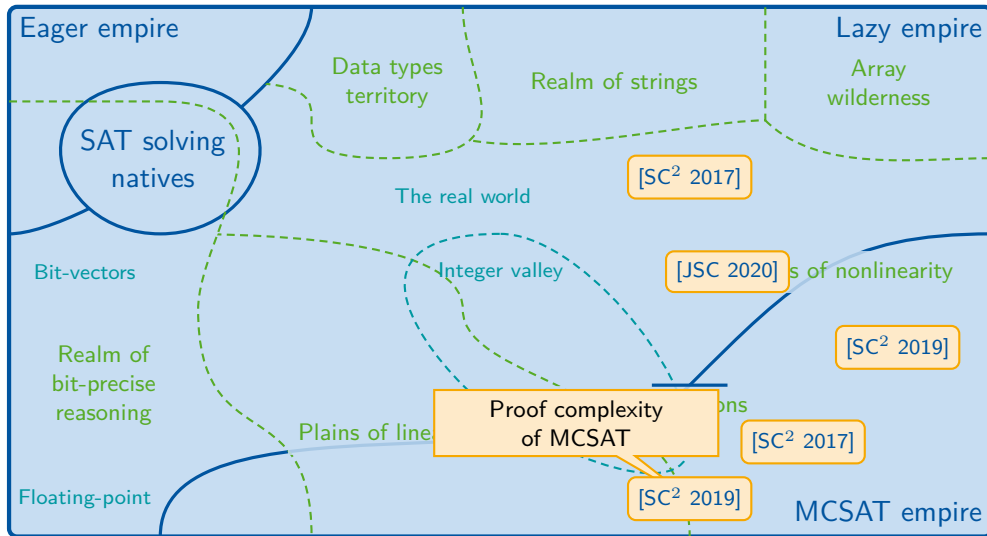
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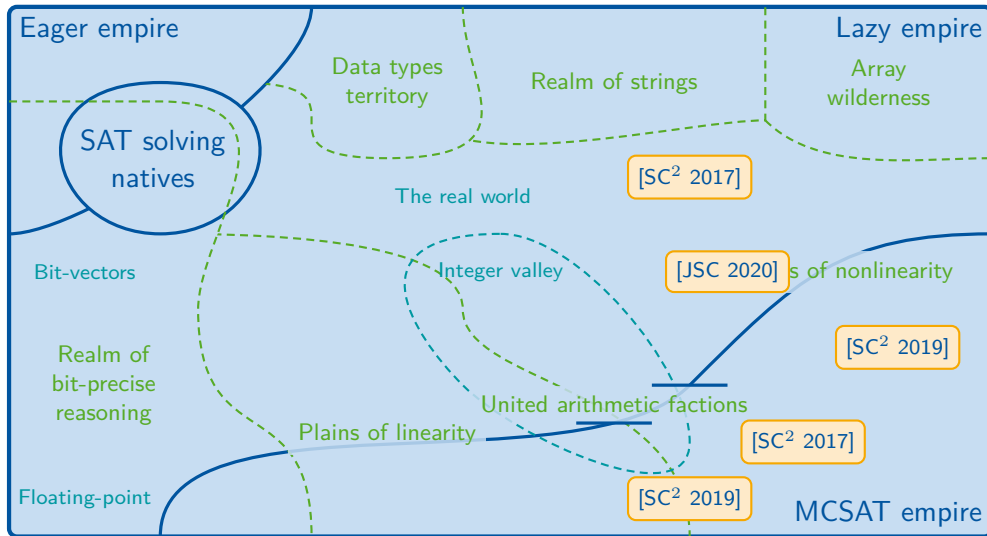
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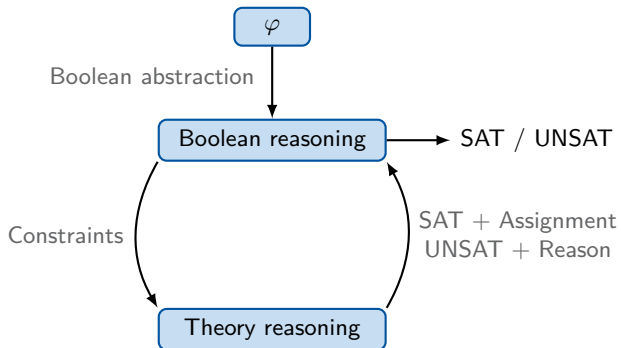
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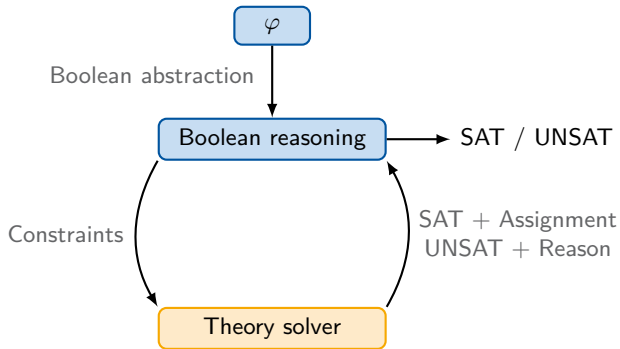
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Lazy SMT solving

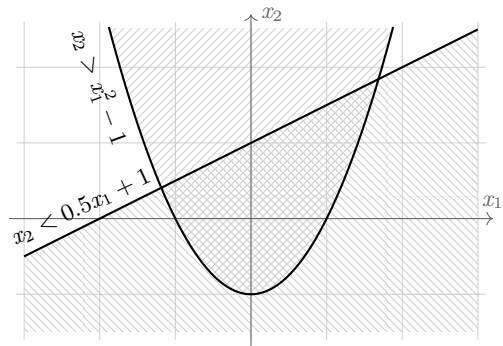


Lazy SMT solving



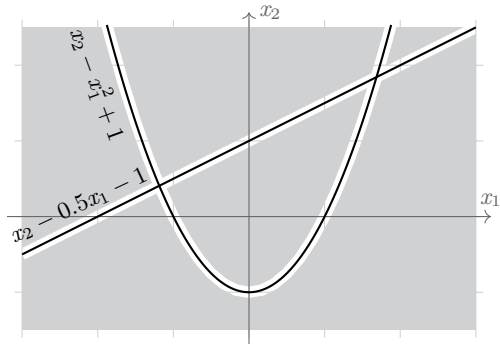
Core ideas of CAD

- Input: set of constraints



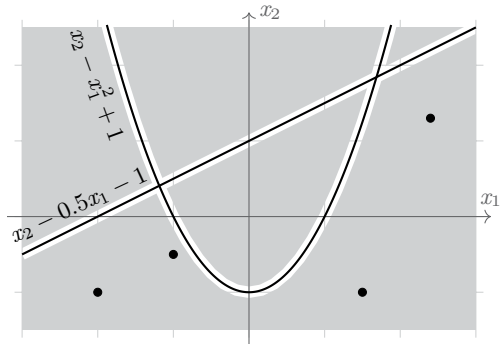
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- Input: set of constraints
- Consider **polynomials** of these constraints
- Identify **sign-invariant** regions



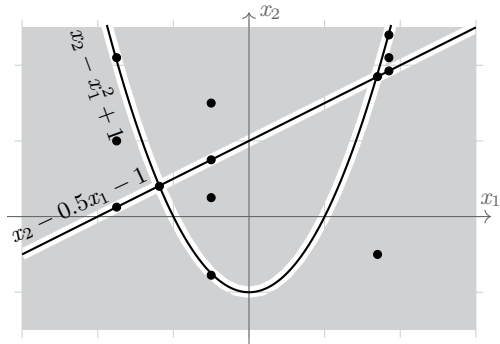
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- Sign-invariance (of polynomials) implies **truth-invariance** (of the formula)
- All samples in one region are **equivalent**



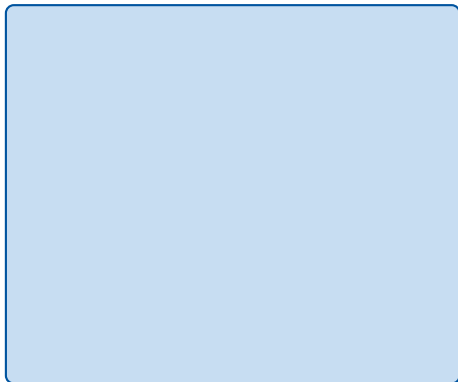
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- All samples in one region are **equivalent**
- Construct **one sample** per region
- Evaluate samples on constraints

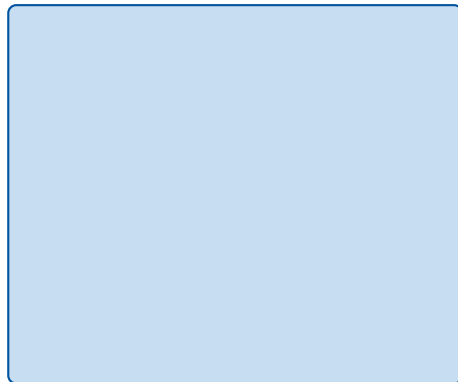


CAD in a nutshell

Set of constraints

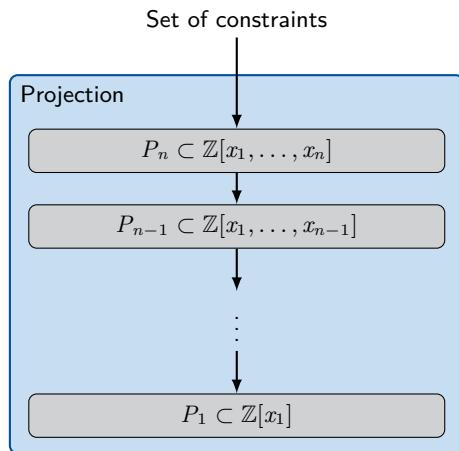


SAT / UNSAT

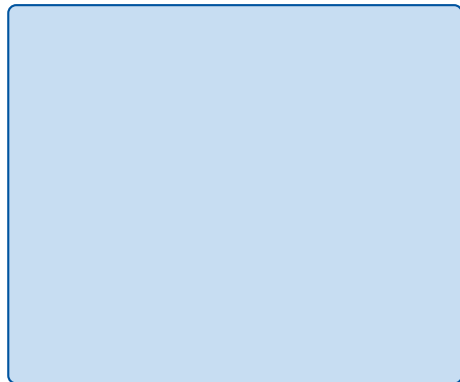


[Collins 1975]

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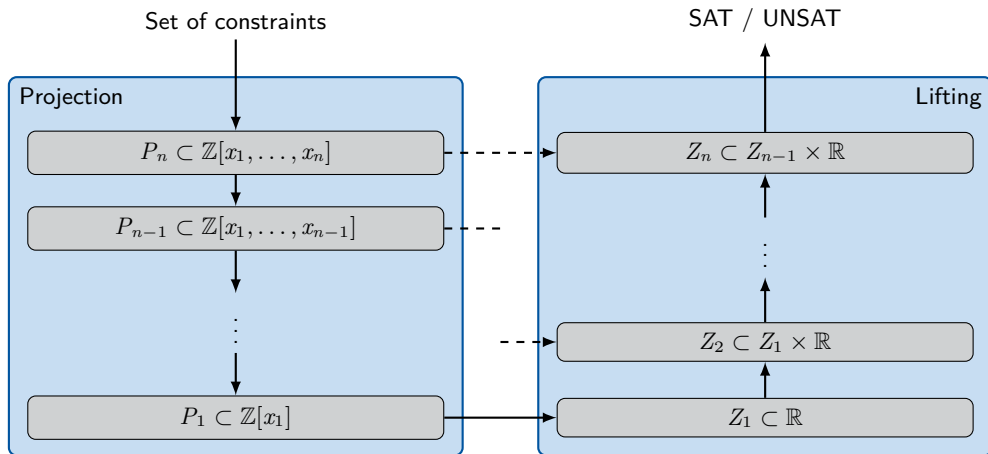


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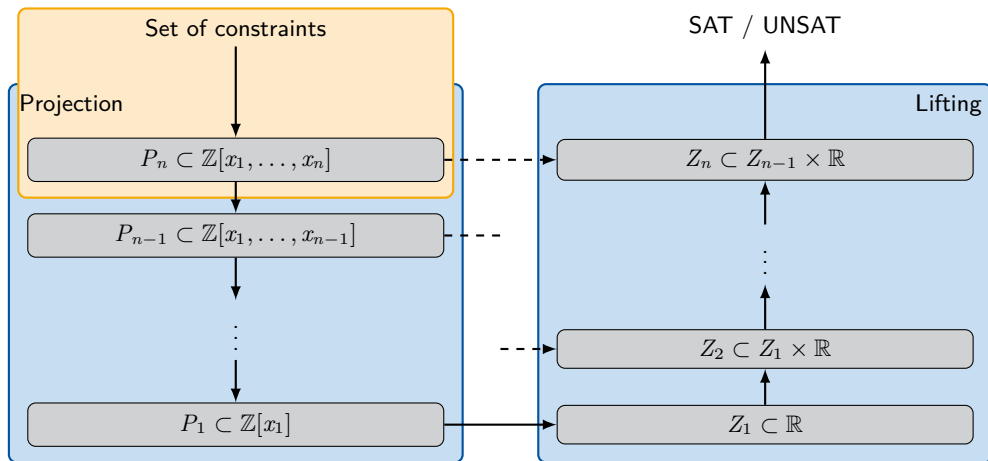
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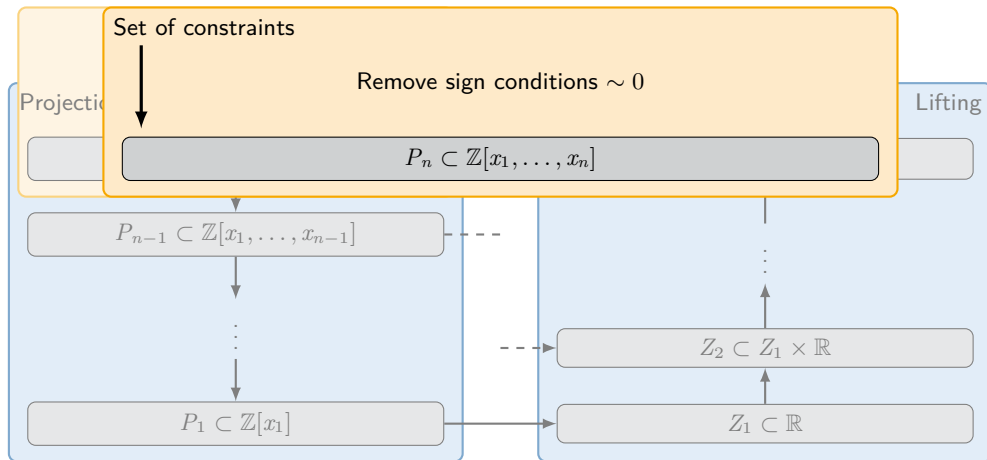
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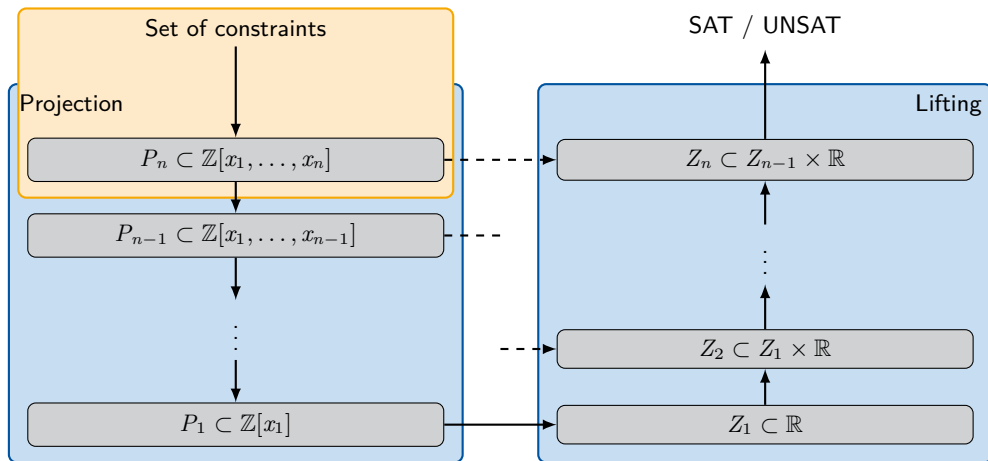
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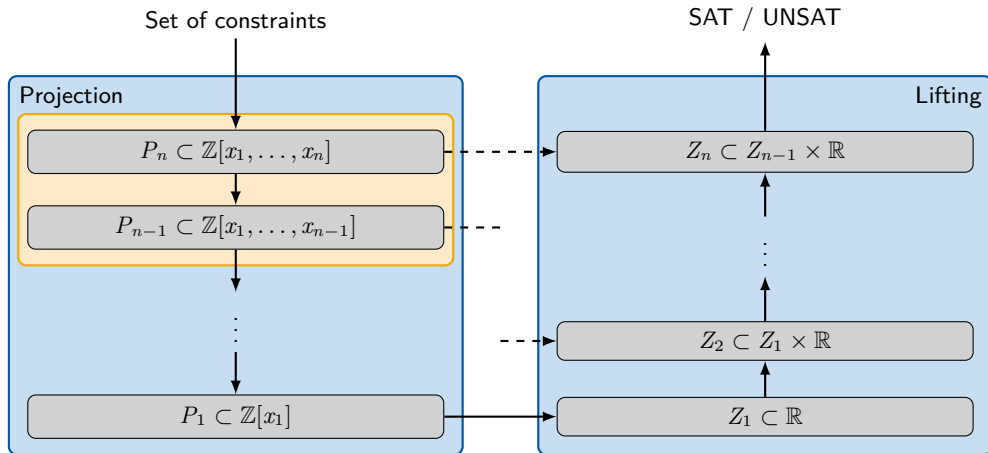
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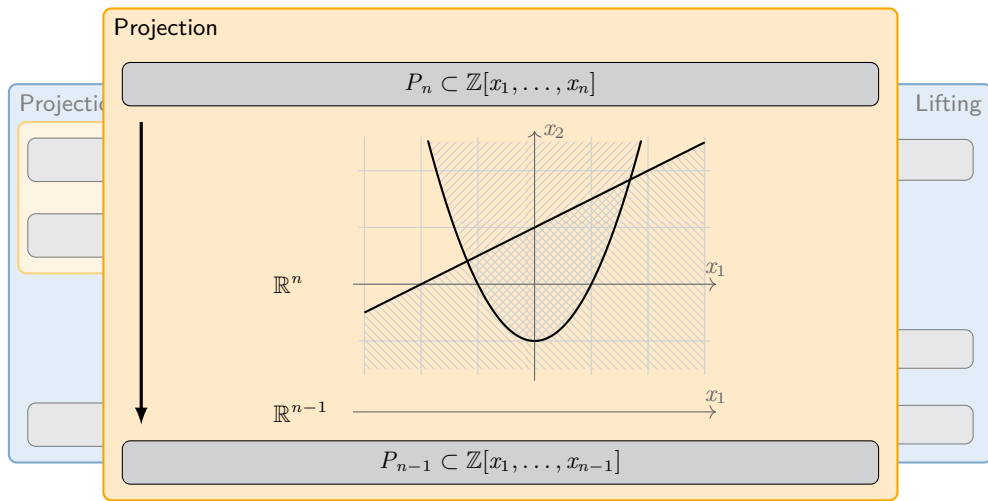
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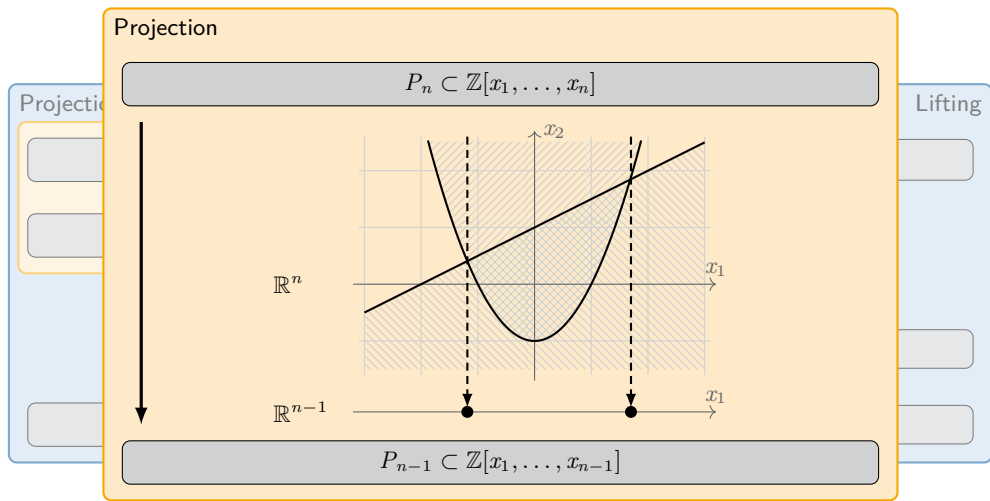
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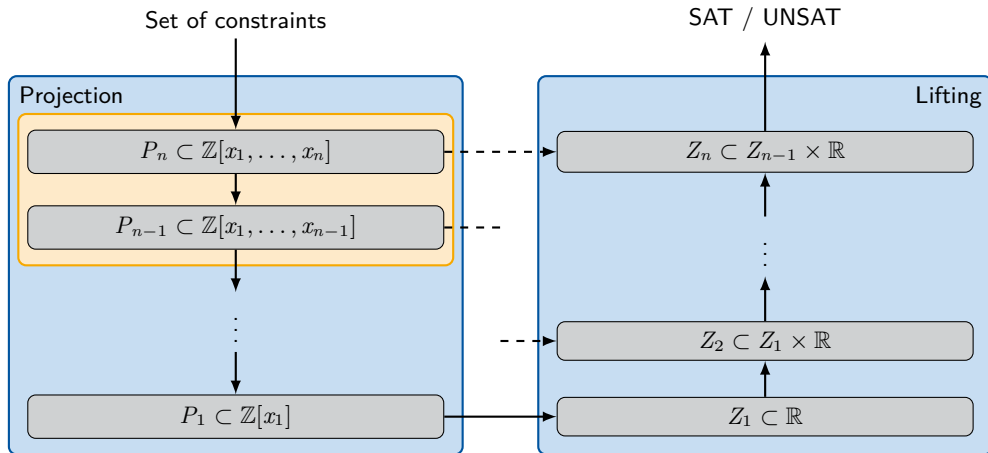
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CAD in a nutshell



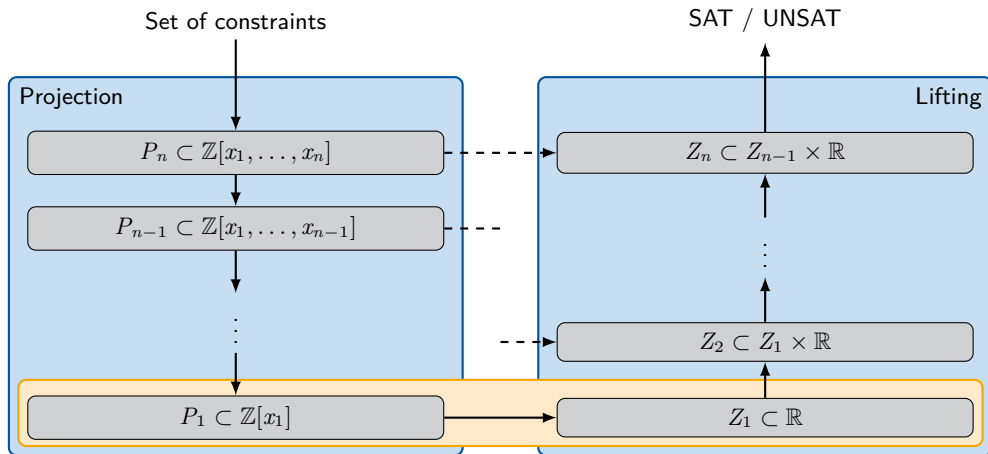
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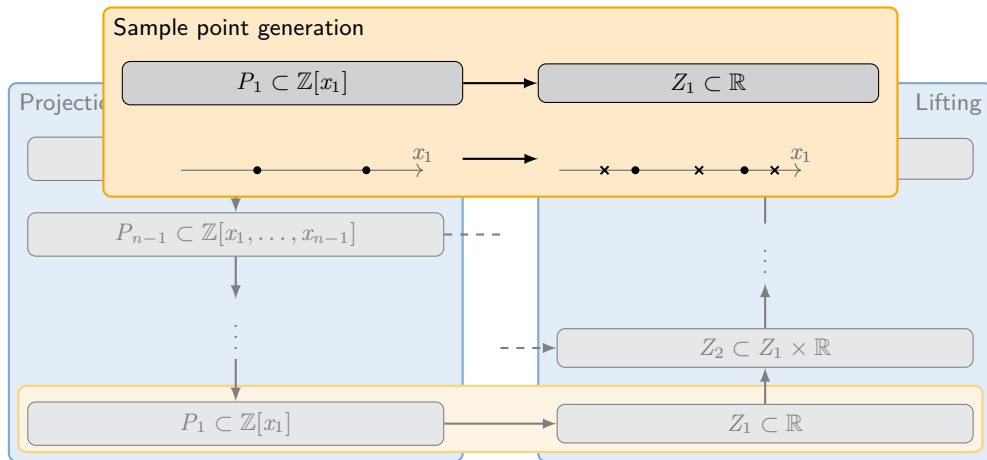
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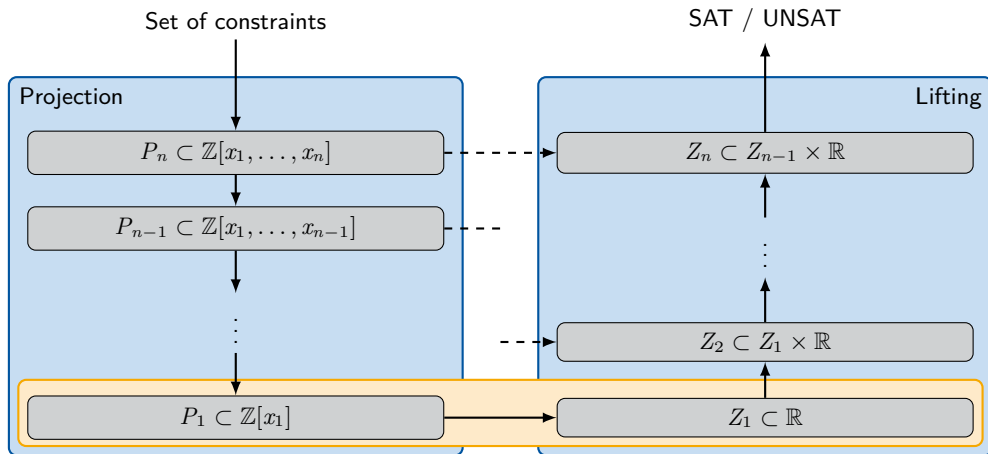
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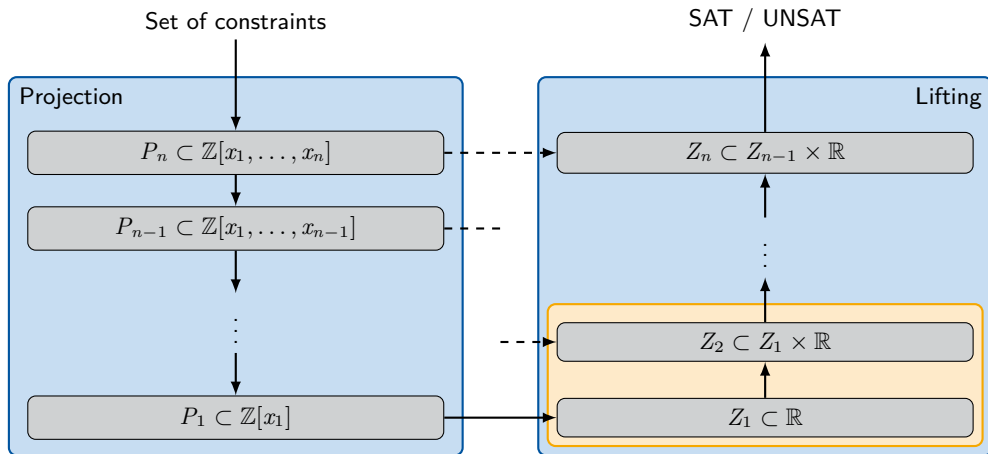
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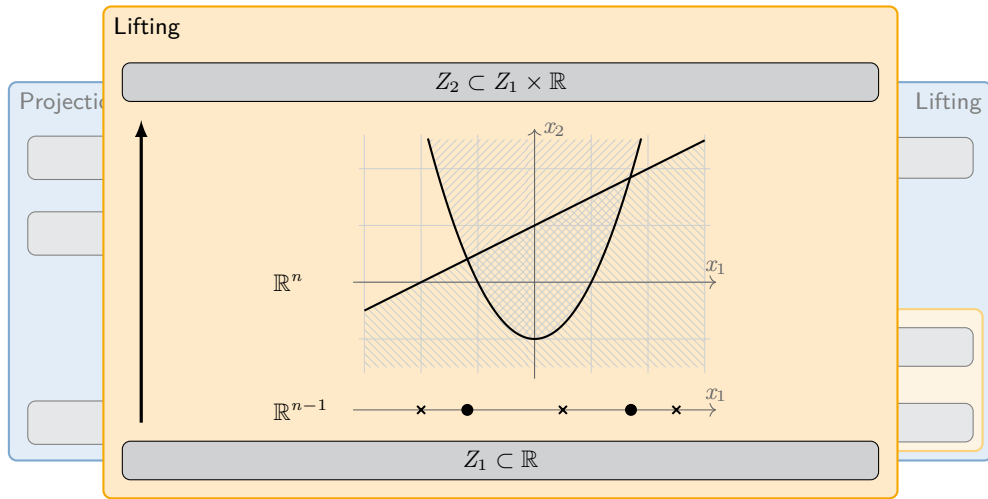
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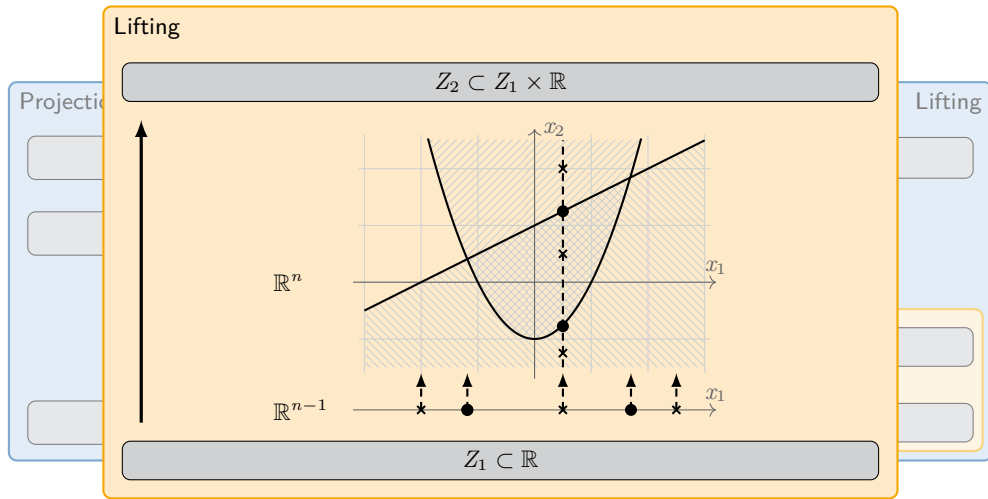
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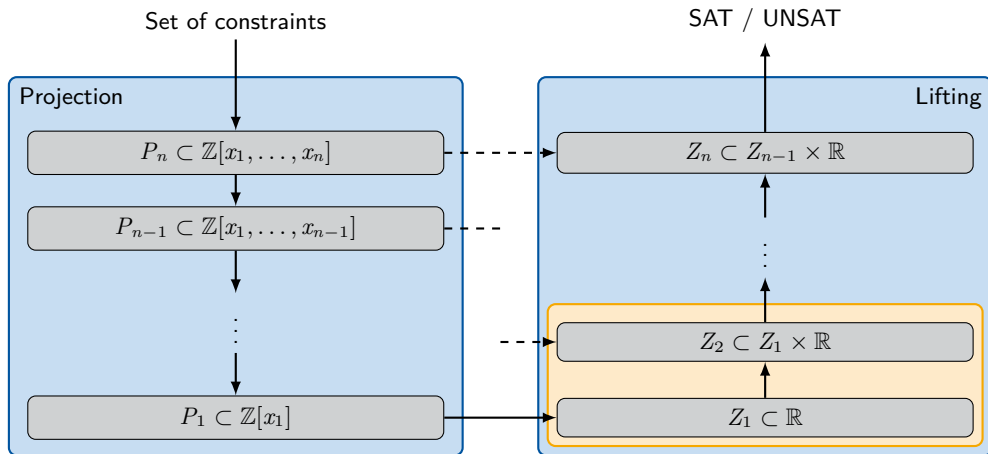
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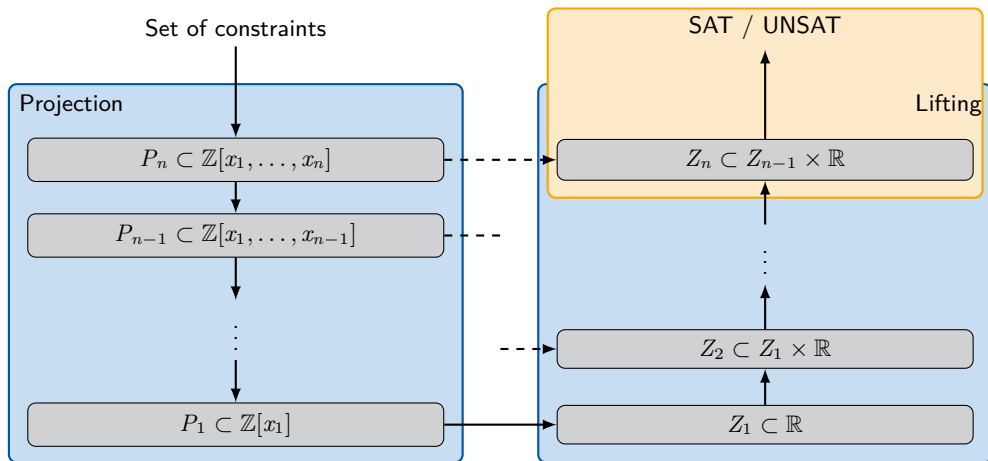
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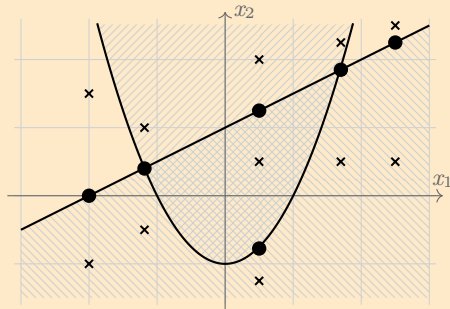
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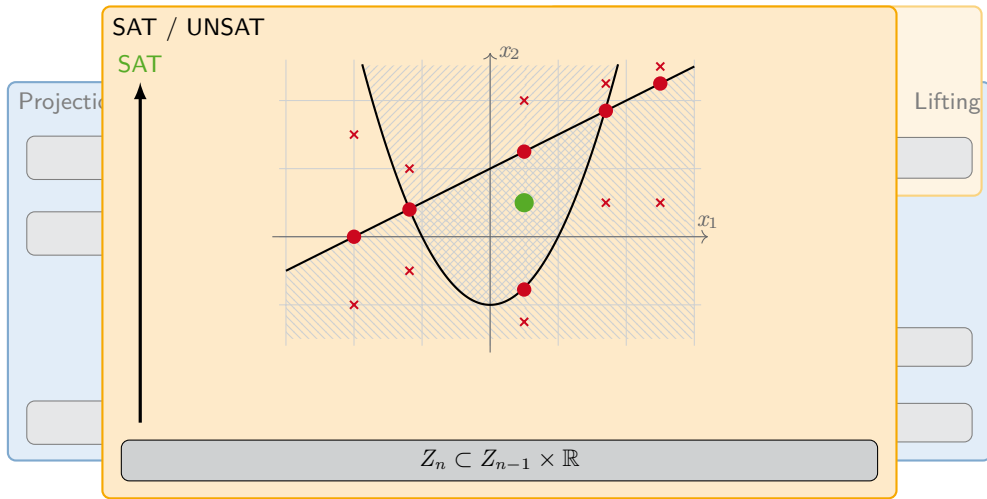
SAT / UNSAT



$$Z_n \subset Z_{n-1} \times \mathbb{R}$$

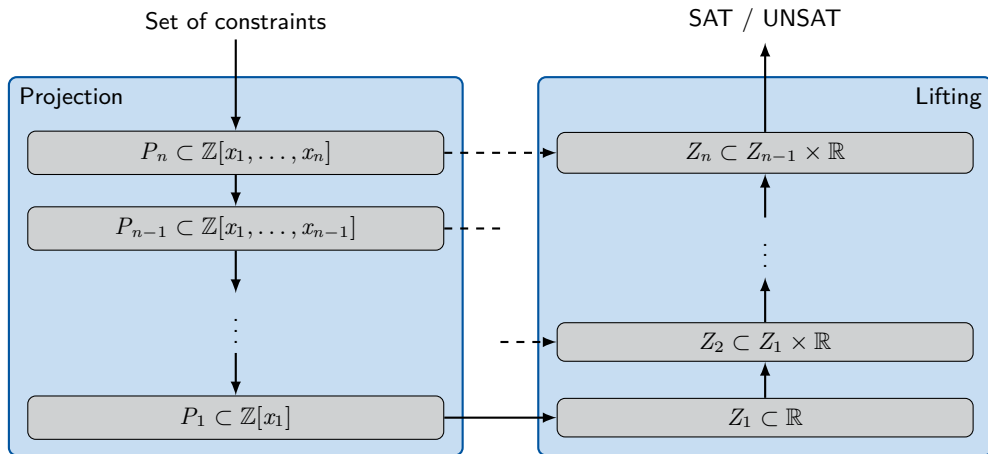
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CAD in a nutshell



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Towards incrementality

~ 2010: How can we solve nonlinear real arithmetic in SMT?

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- CAD [established for quantifier elimination](#)
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Our goal: make CAD [as efficient as possible for SMT](#)

Towards incrementality

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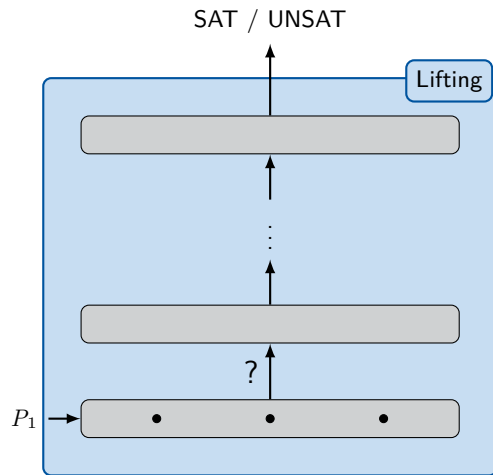
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Our goal: make CAD *as efficient as possible for SMT*

- Incrementality: *reuse* previously computed results
- Backtracking: remove part of the input *efficiently*
- Reasons for unsatisfiability: *small* infeasible subsets

Incremental lifting

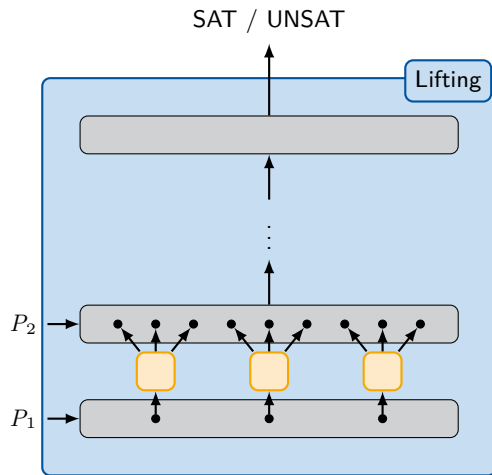
[JSC 2020]



[Collins + 1991]

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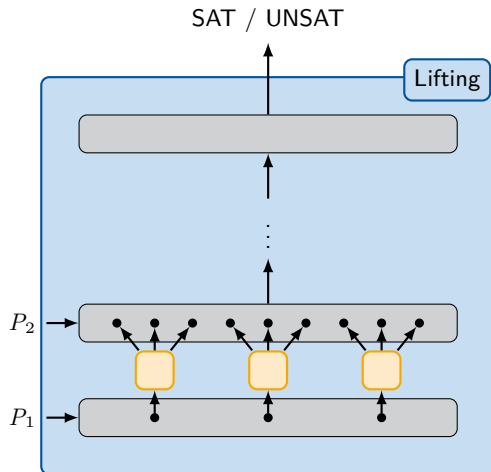
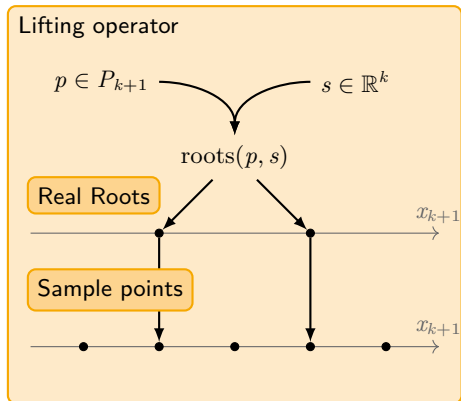
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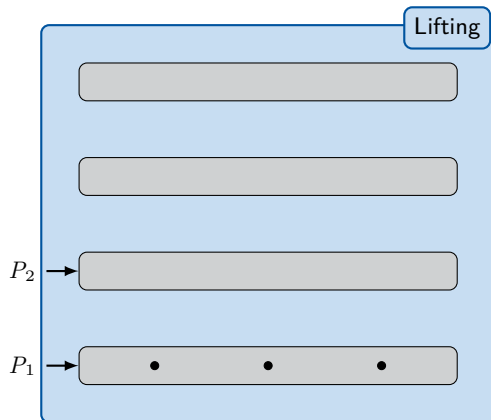
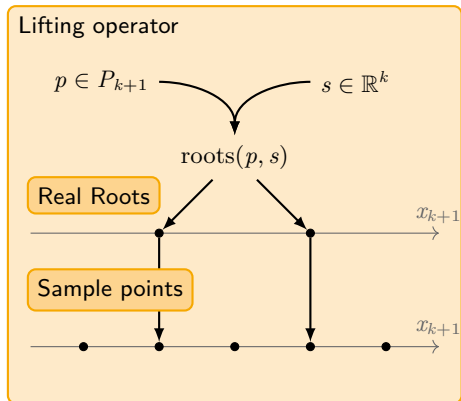


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SAT / UNSAT

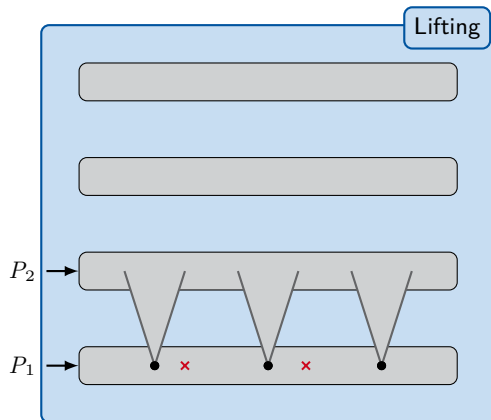
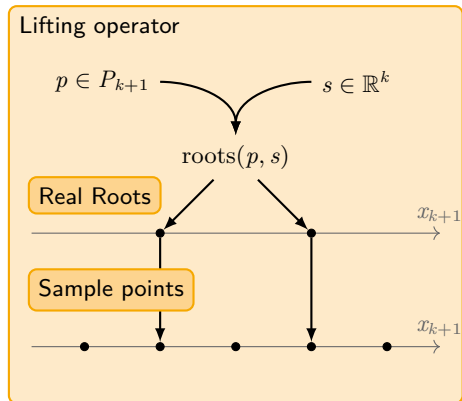


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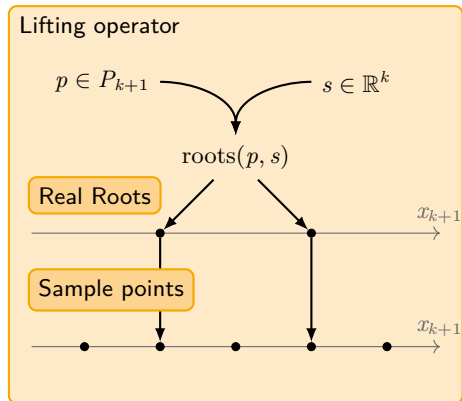
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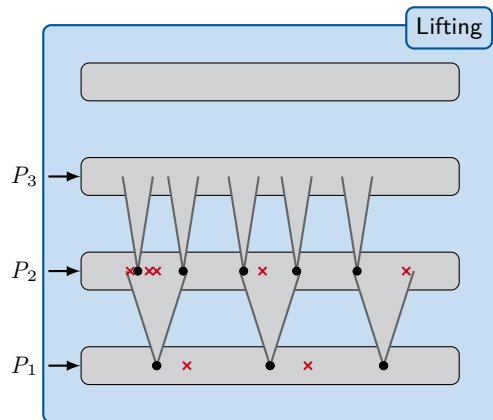
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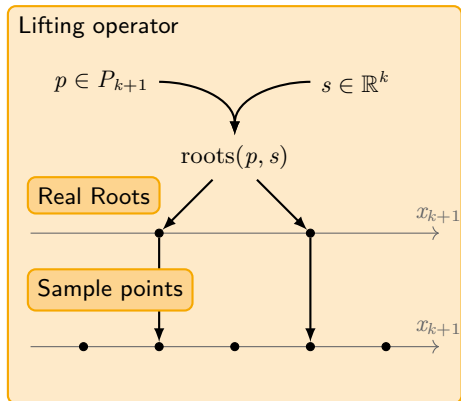
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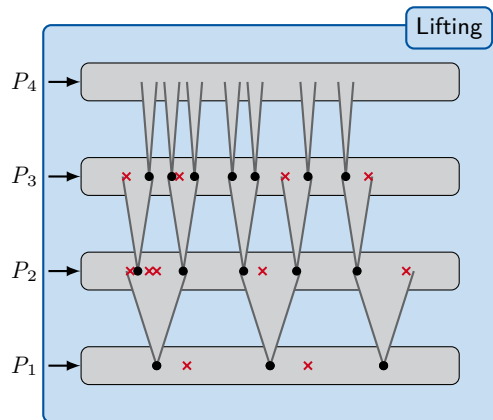
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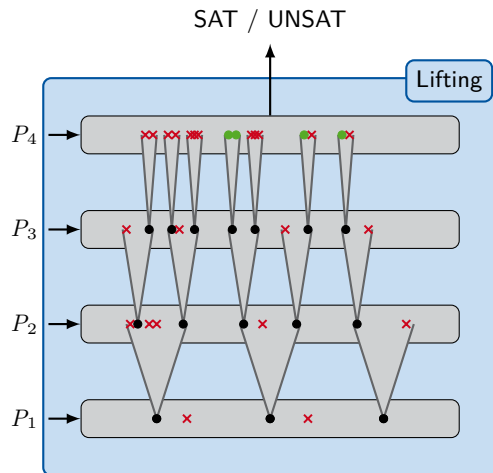
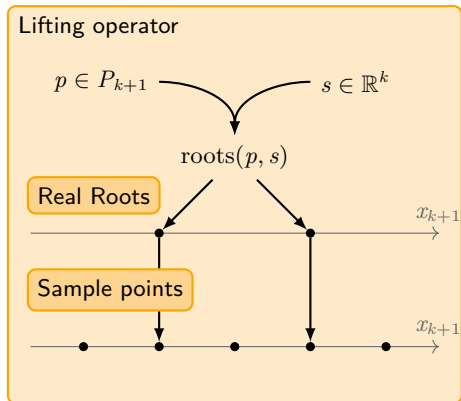
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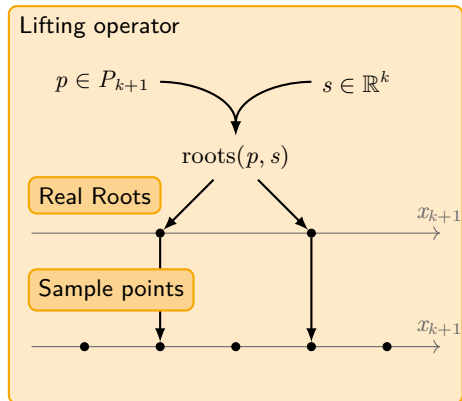
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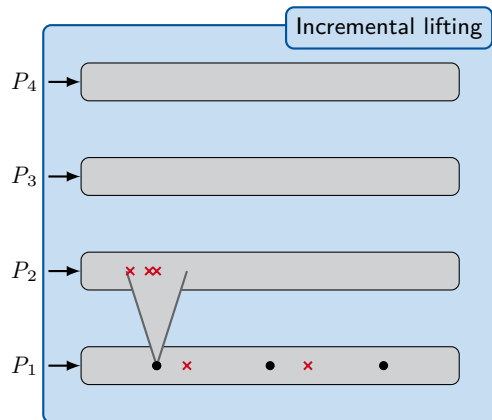
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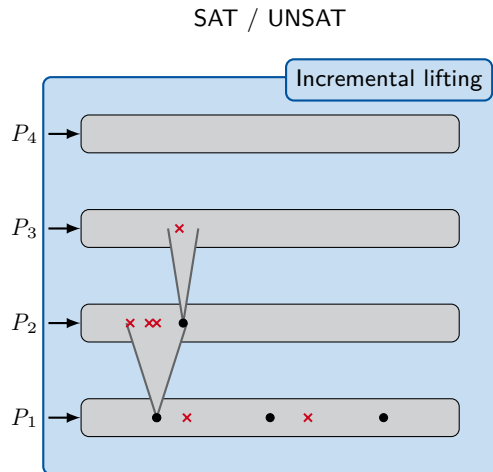
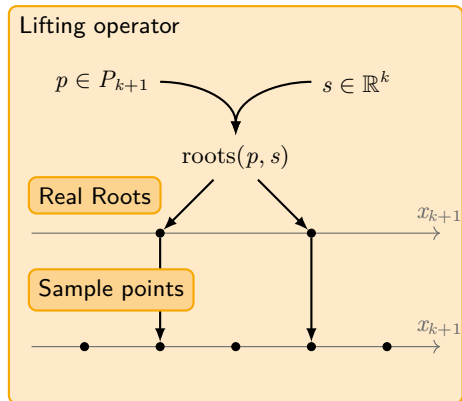
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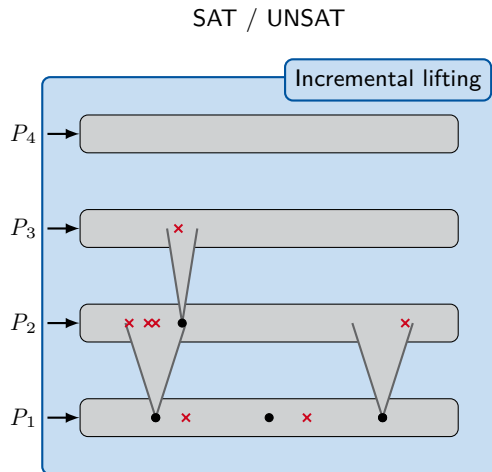
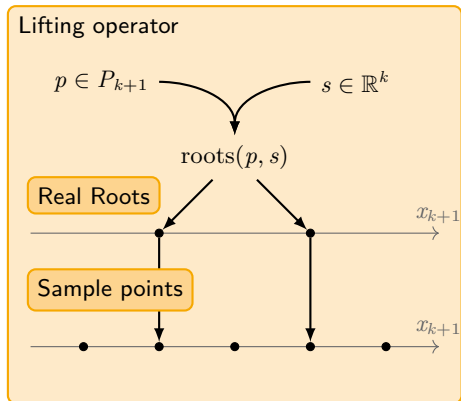
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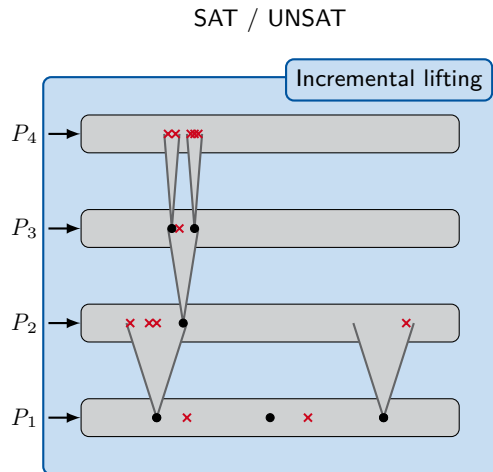
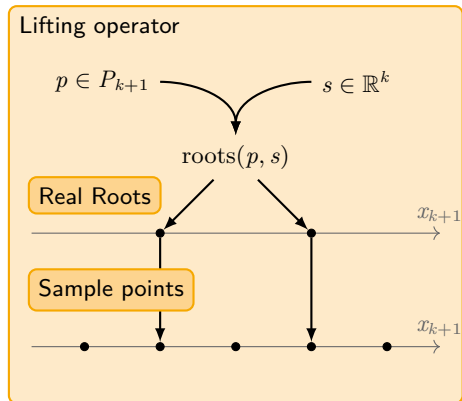
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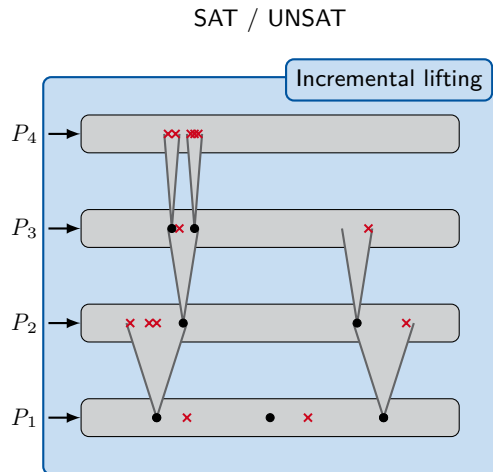
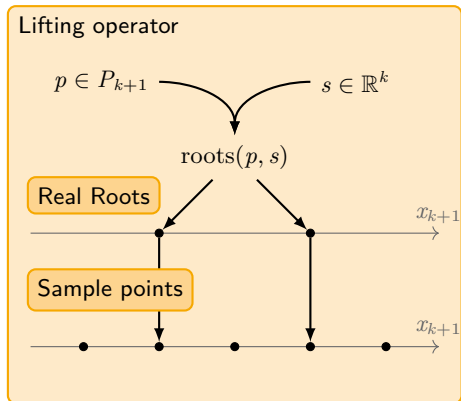
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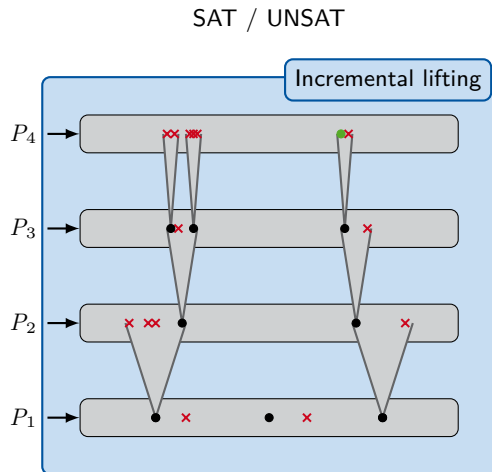
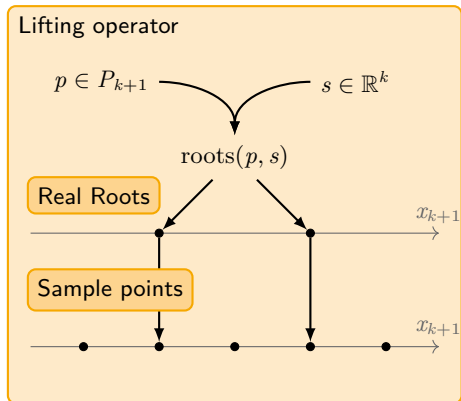
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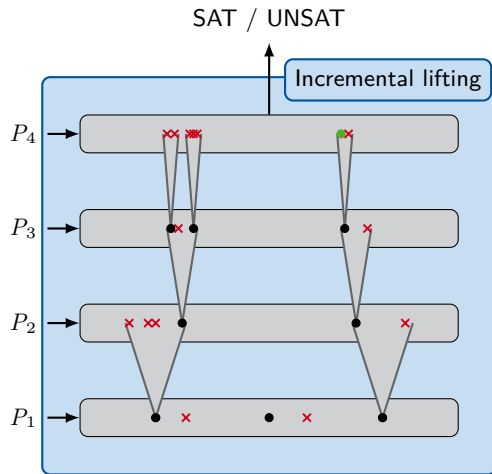
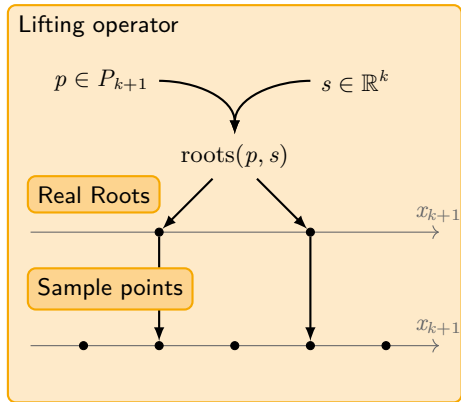
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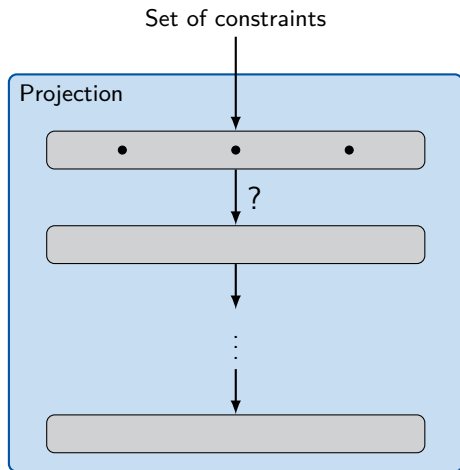
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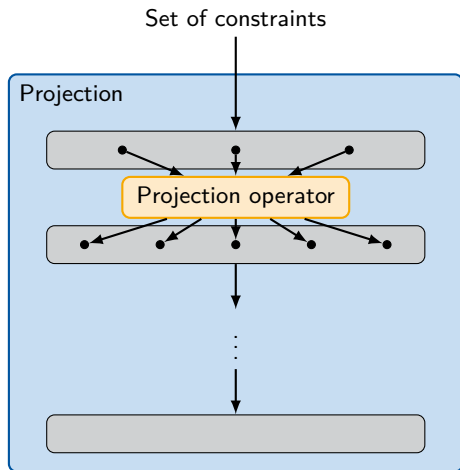
Incremental projection

[JSC 2020]



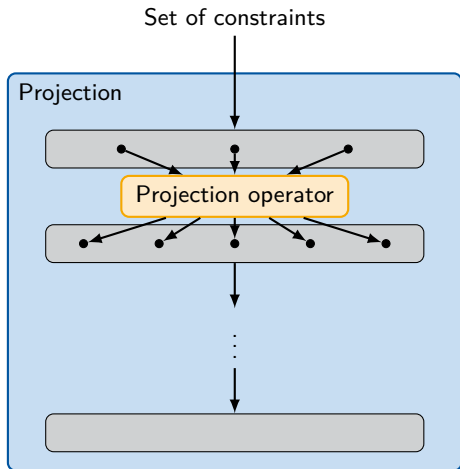
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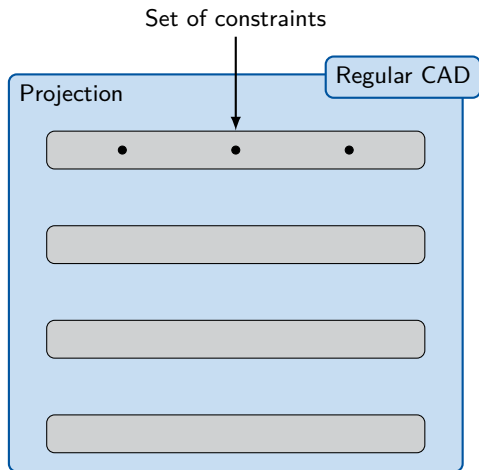


Projection operator

$$Proj(P) = \{Proj(p) \mid p \in P\} \\ \cup \{Proj(p, q) \mid p, q \in P\}$$

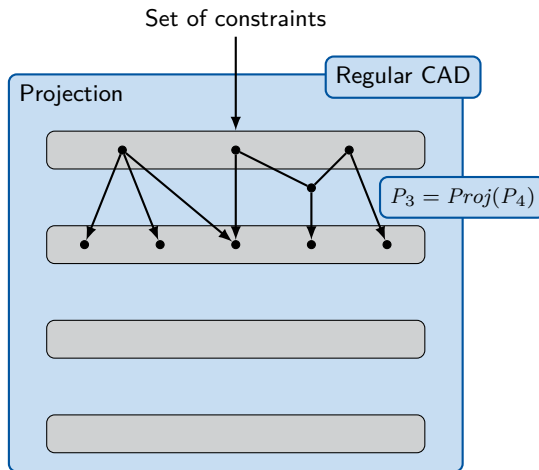
Incremental projection

[JSC 2020]



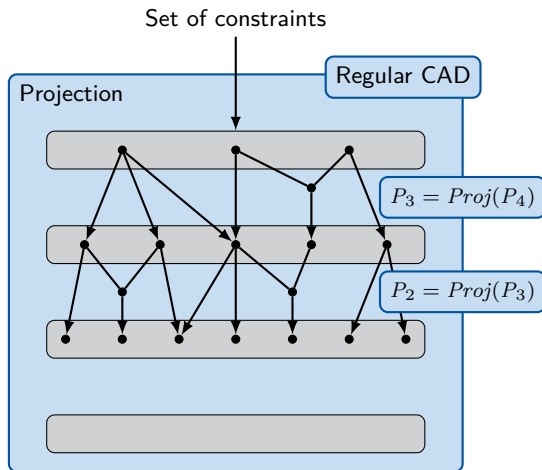
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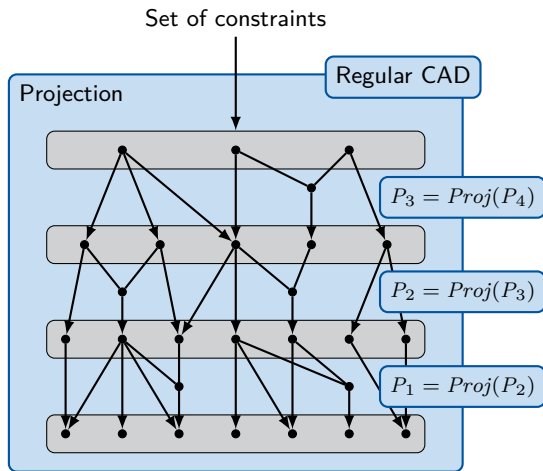
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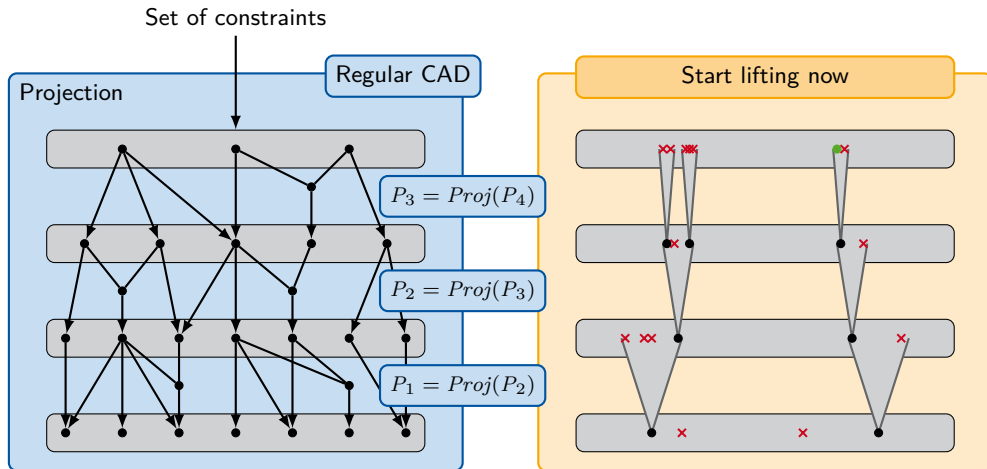
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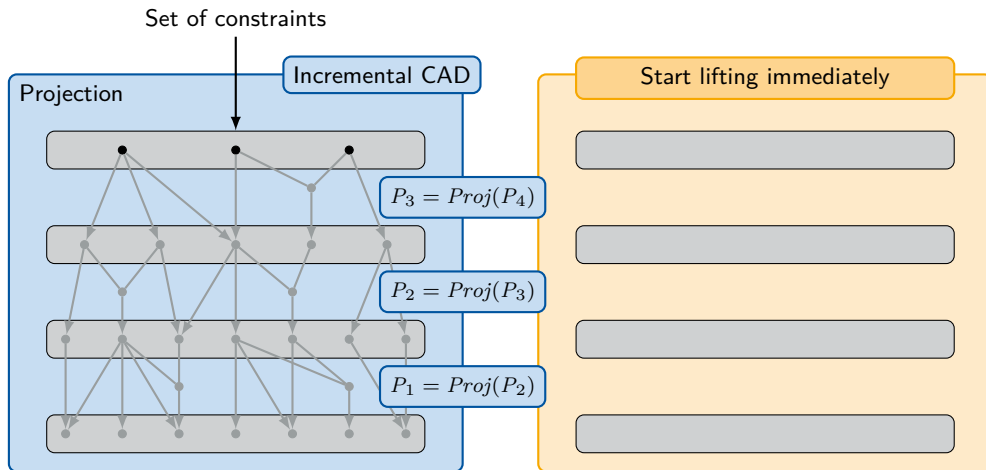
Incremental projection

[JSC 2020]



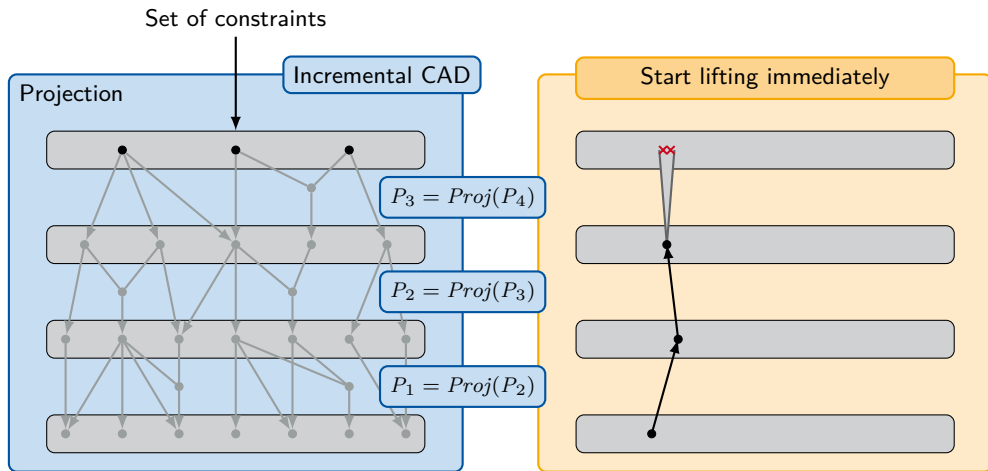
Incremental projection

[JSC 2020]



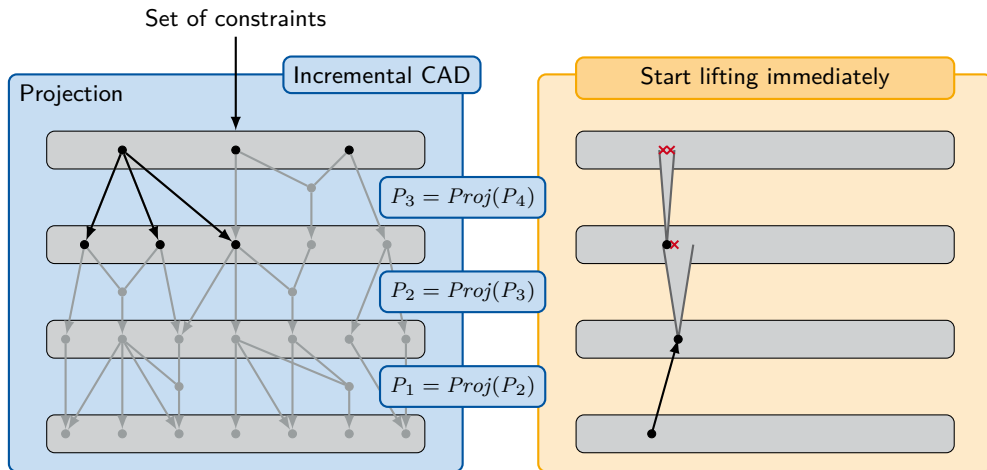
Incremental projection

[JSC 2020]



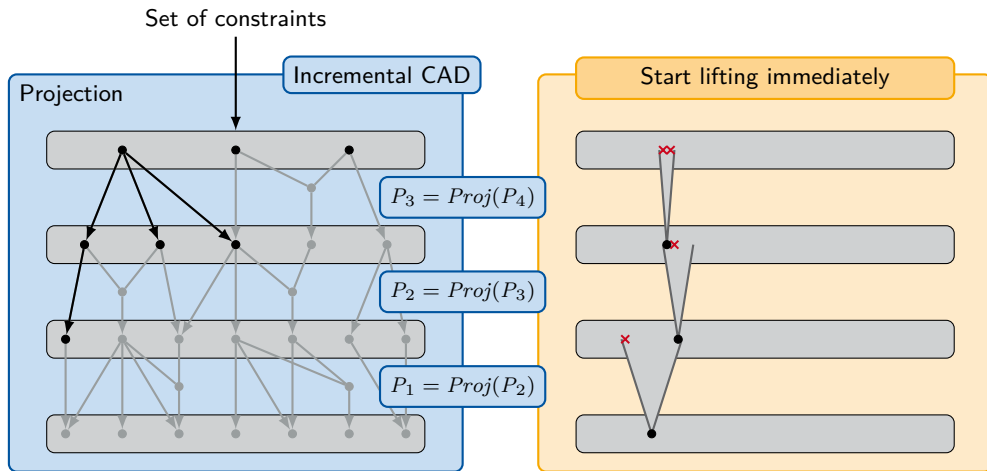
Incremental projection

[JSC 2020]



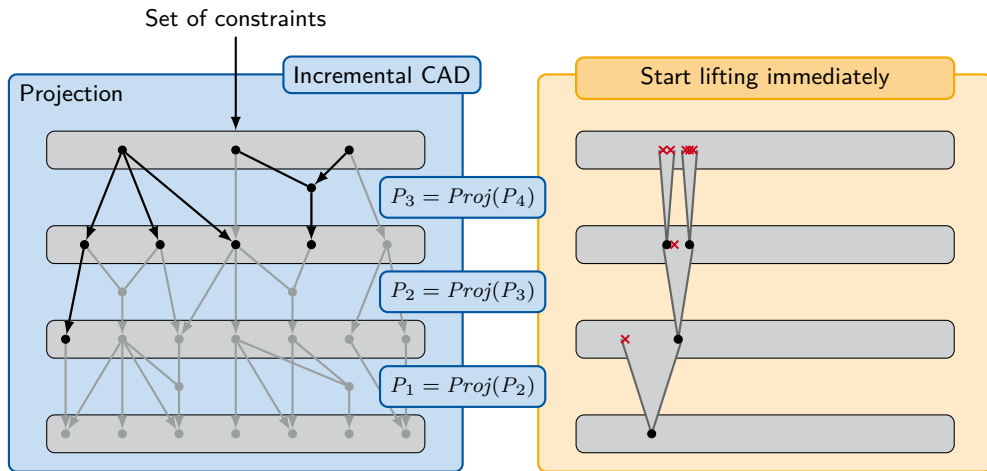
Incremental projection

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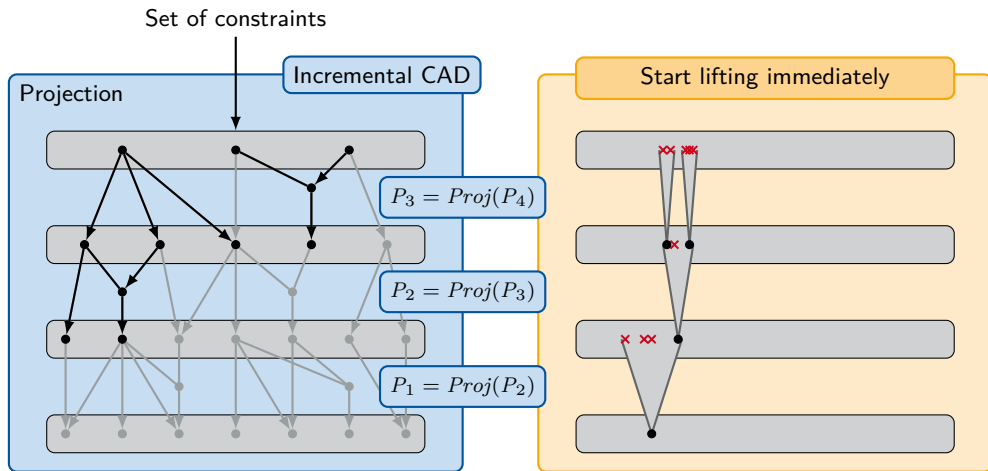
Incremental projection

[JSC 2020]



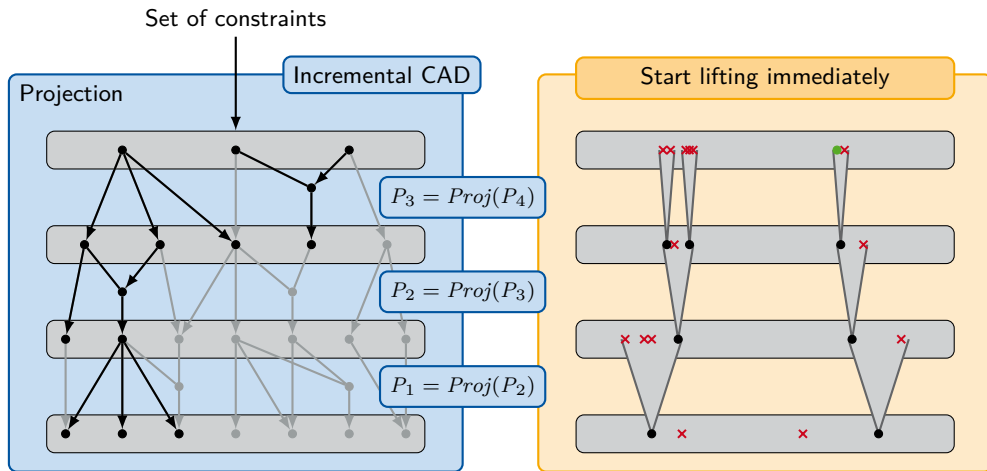
Incremental projection

[JSC 2020]



Incremental projection

[JSC 2020]



Heuristics & Additions

[JSC 2020]

■ Level of incrementality

Trade-off between potential [gains](#) and [bookkeeping](#)

■ Projection operators

Collins, Hong, McCallum, Lazard, Brown

[\[SC² 2017\]](#)

■ Variable ordering

[\[England⁺ 2014\]](#) [\[Huang⁺ 2014\]](#)

■ Scheduling of projection and lifting

Which polynomials and sample points to use first

■ Equational constraints

Exploits [equalities](#) to reduce projection

[\[McCallum 1999\]](#) [\[SC² 2018\]](#)

■ Factorization of polynomials

Ensures correctness, can reduce projection

■ Minimal infeasible subsets

Allow SAT solver to learn [smaller conflicts](#)

[\[Jaroschek⁺ 2015\]](#)[\[Hentze 2017\]](#)

Experimental results

- Timeout: 120s
- Benchmarks: QF_NRA from SMT-LIB

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Solver	SAT	UNSAT	overall
Naive	4285	3496	7781
None	4293	3507	7800
Simple	4281	3889	8170
Full	4328	4250	8578
Full-hide	4328	4255	8583

- Incrementality is [worth the effort](#)
- Bookkeeping [not that costly](#)

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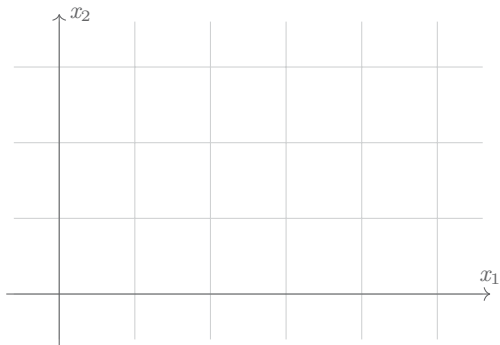
- Incrementality is **worth the effort**
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Projection operators

Solver	SAT	UNSAT	overall
Collins	4292	4150	8442
Hong	4301	4189	8490
McCallum	4320	4216	8536
Lazard	4322	4229	8551
Brown	4328	4250	8578

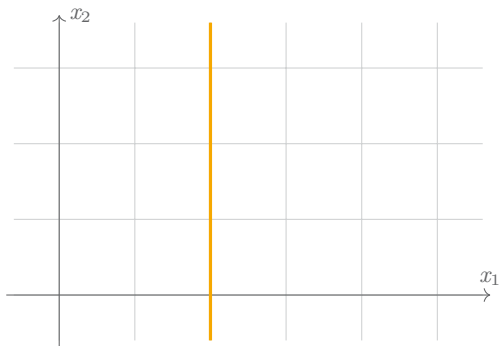
- **Confirms** conventional wisdom
- Margins **may be surprising**

Model-Constructing Satisfiability Calculus



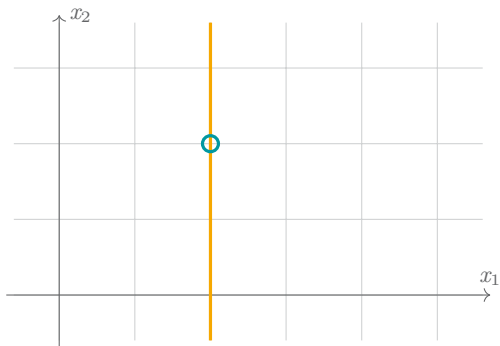
[Jovanović + 2012] [Moura + 2013]

Model-Constructing Satisfiability Calculus



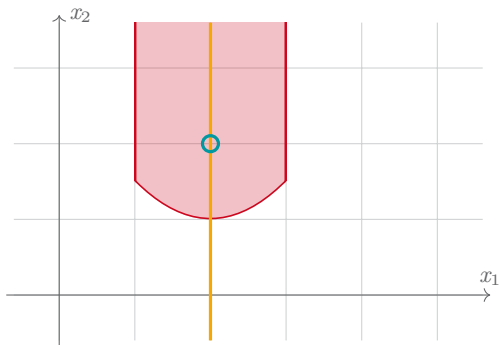
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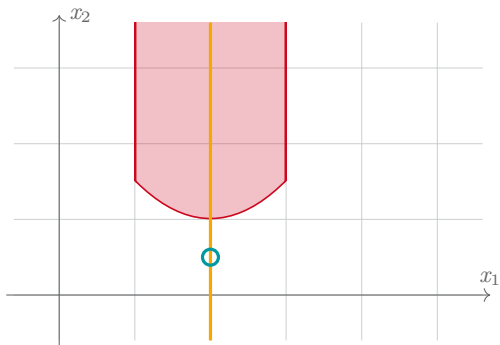
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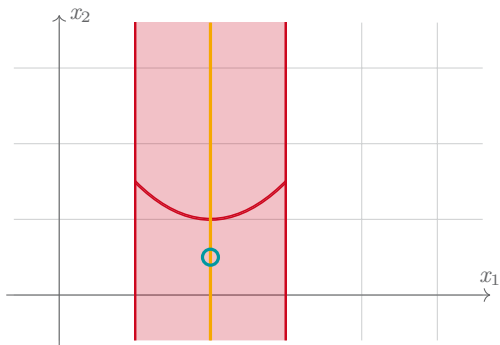
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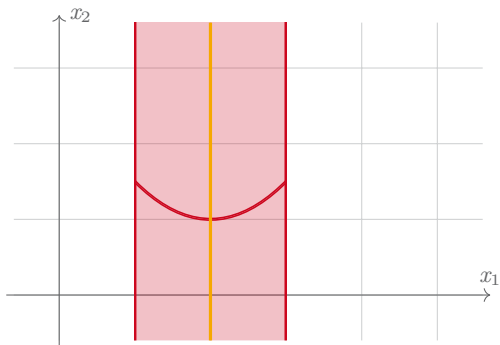
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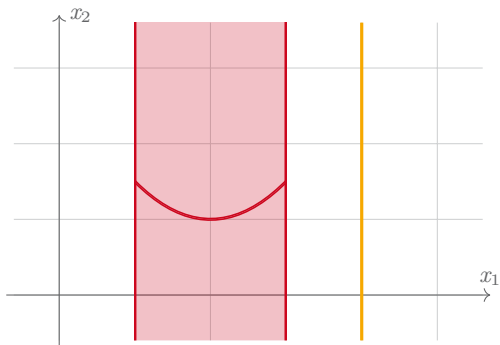
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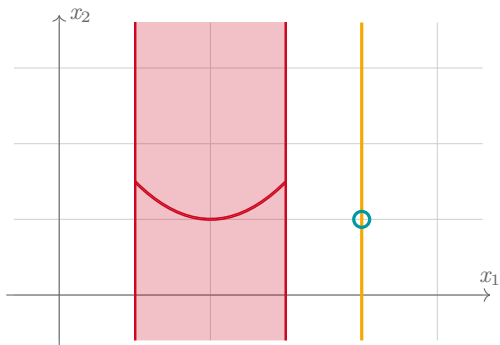
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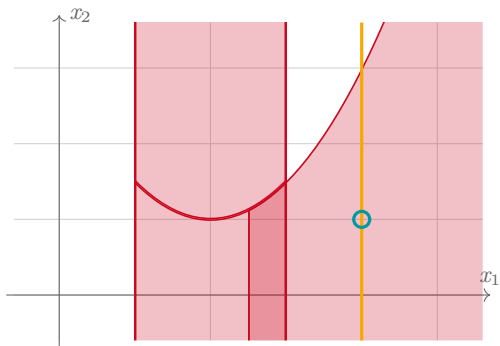
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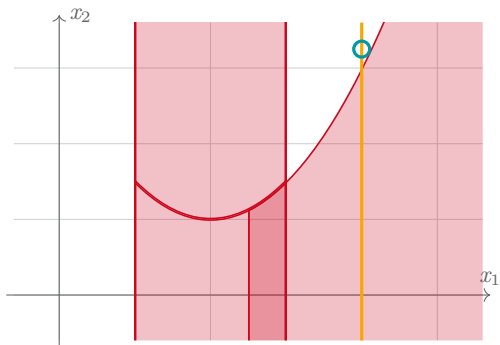
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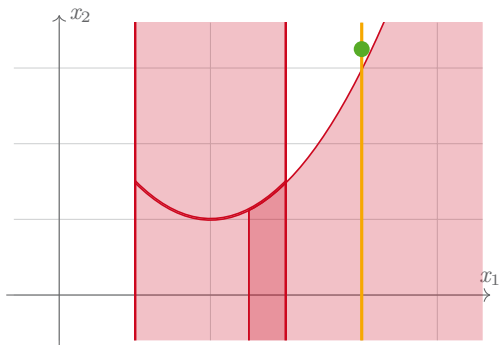
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Model-Constructing Satisfiability Calculus



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Model-Constructing Satisfiability Calculus



[Jovanović + 2012] [Moura + 2013]

Model-Constructing Satisfiability Calculus

(Lazy) SMT solver

[Jovanović + 2012] [Moura + 2013]

Model-Constructing Satisfiability Calculus

(Lazy) SMT solver

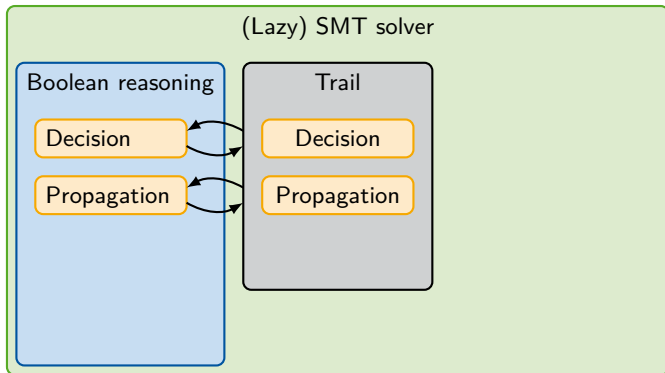
Boolean reasoning

Decision

Propagation

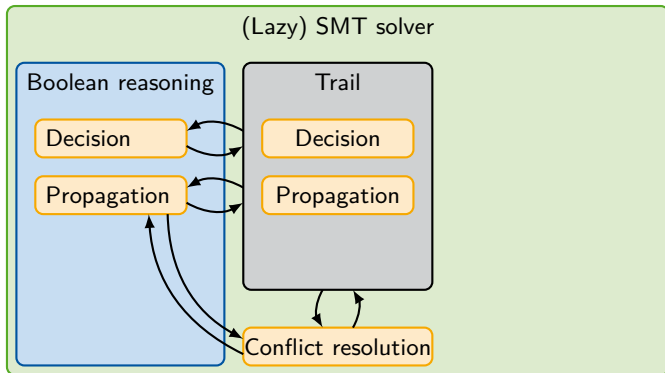
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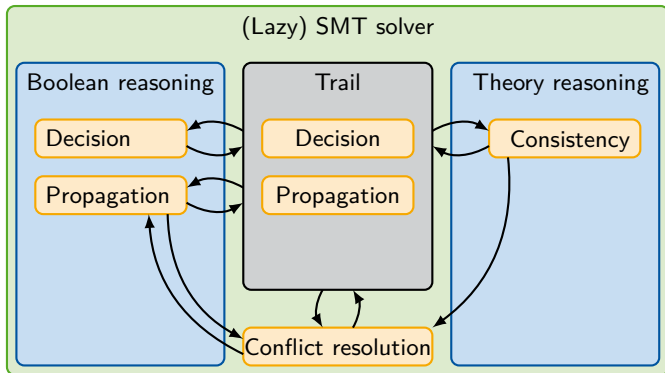
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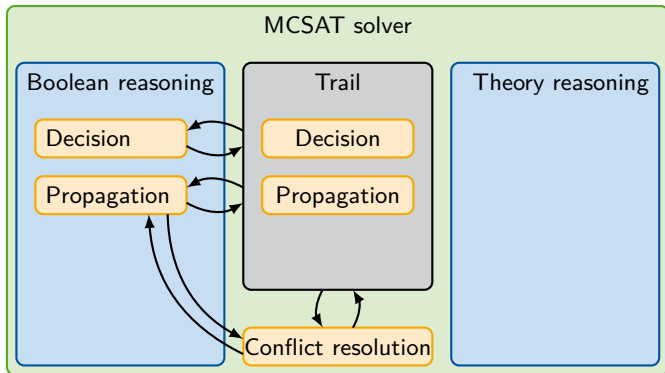
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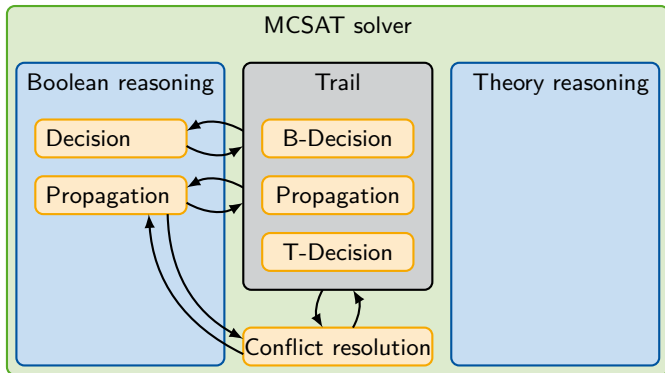
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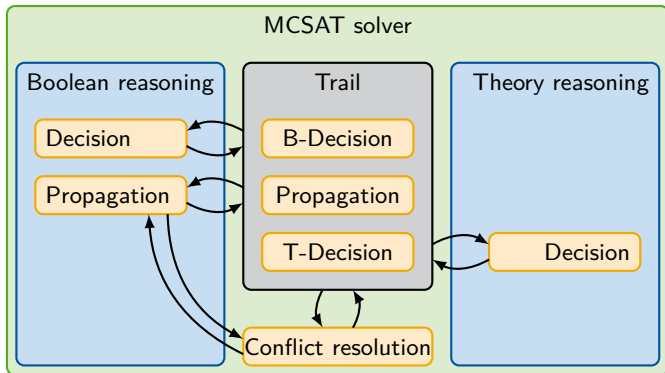
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Model-Constructing Satisfiability Calculus



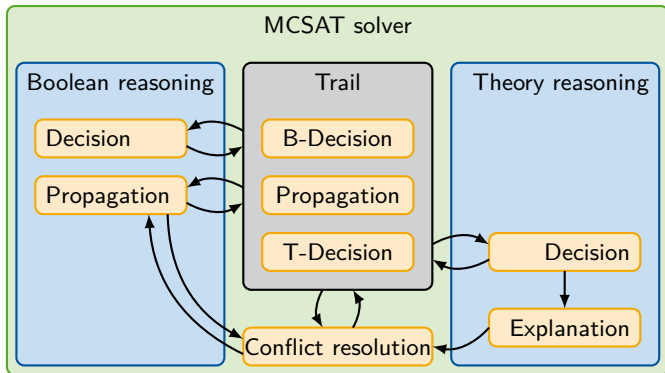
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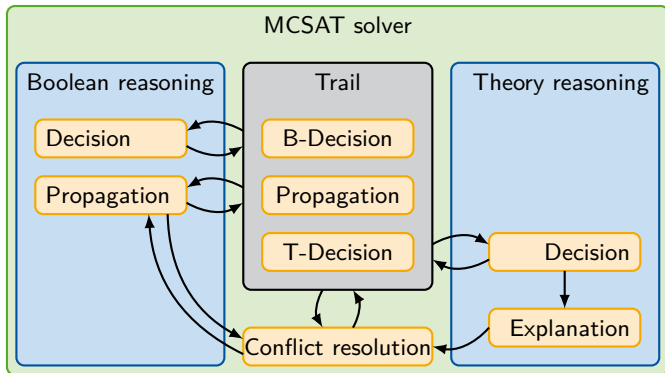
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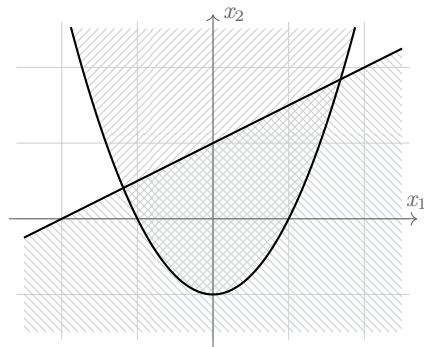


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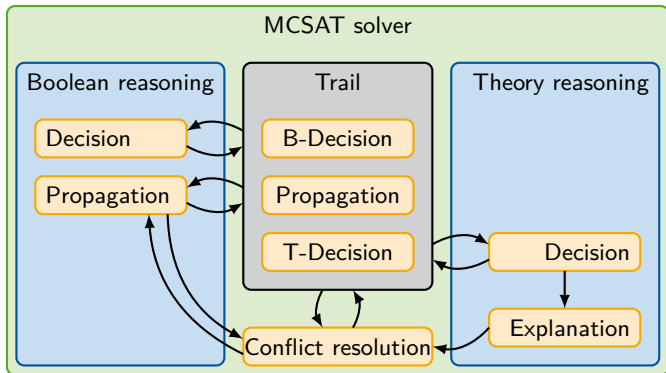


Trail: $\llbracket c_1, c_2 \rrbracket$



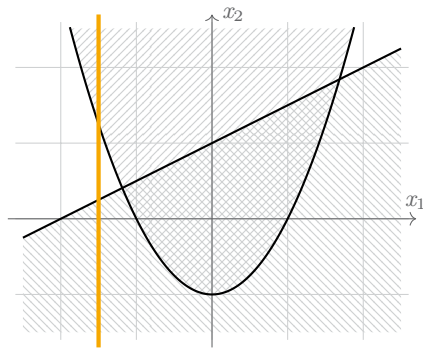
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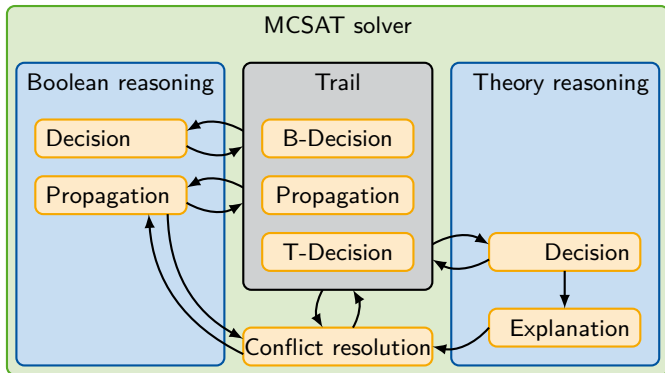
Guess: $x_1 \mapsto -1.5$

Trail: $\llbracket c_1, c_2, x_1 \mapsto -1.5 \rrbracket$



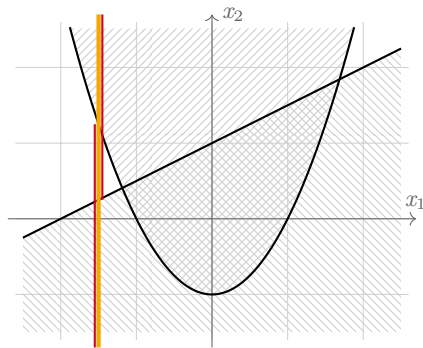
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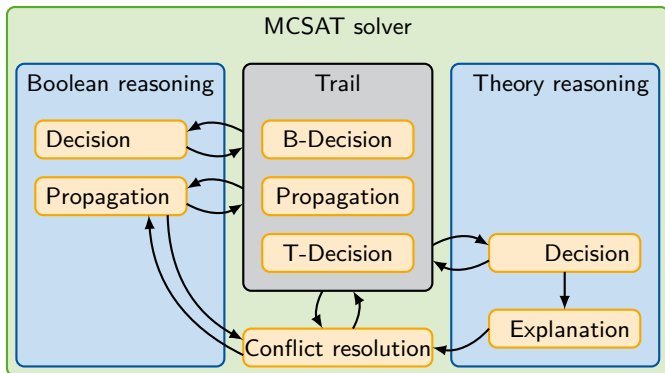
No assignment for x_2 . Reason: c_1, c_2

Trail: $\llbracket c_1, c_2, x_1 \mapsto -1.5 \rrbracket$



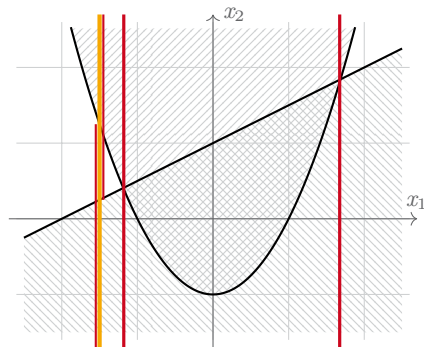
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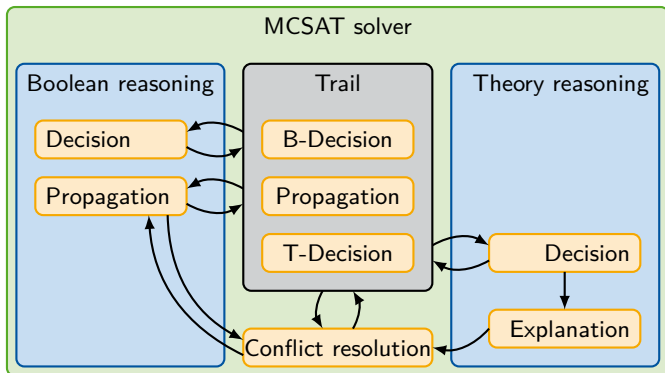
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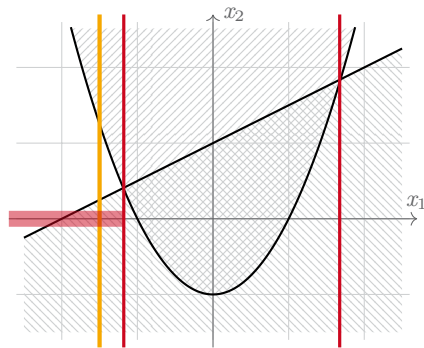
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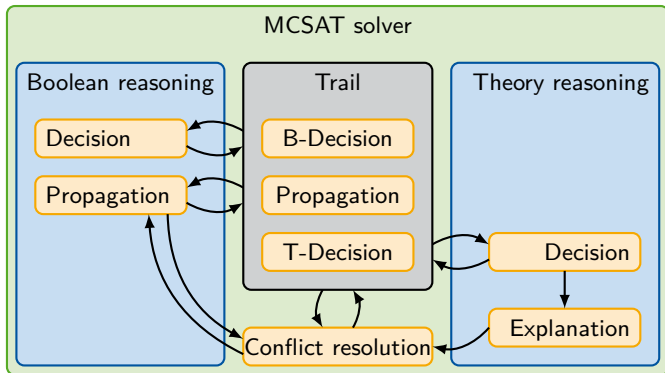
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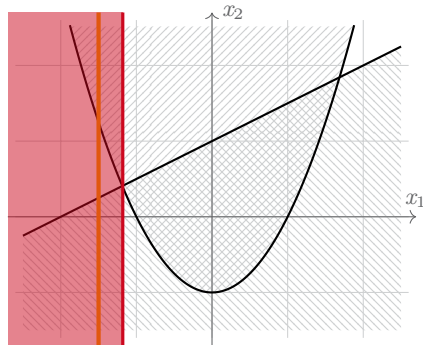
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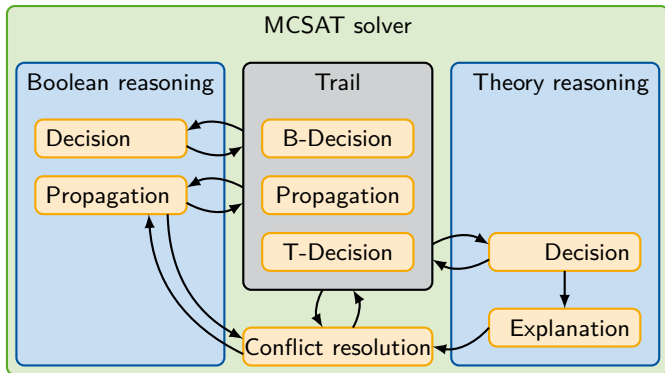
$$(c_1 \wedge c_2) \rightarrow (\alpha_2 \leq x_1)$$

Trail: $\llbracket c_1, c_2 \rrbracket$



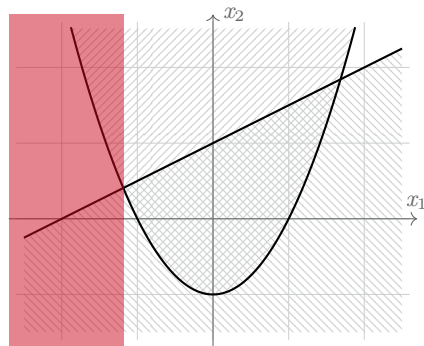
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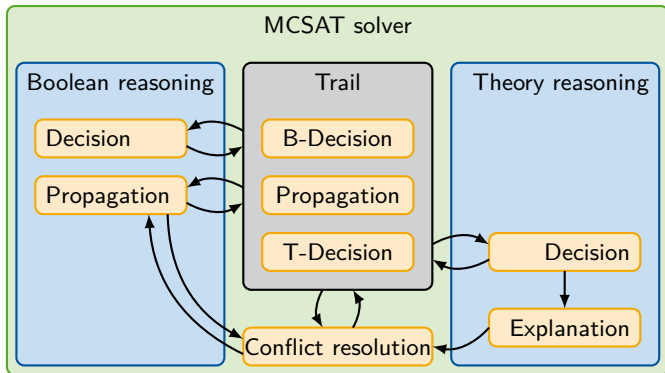
Conflict resolution

Trail: $\llbracket c_1, c_2 \rrbracket$



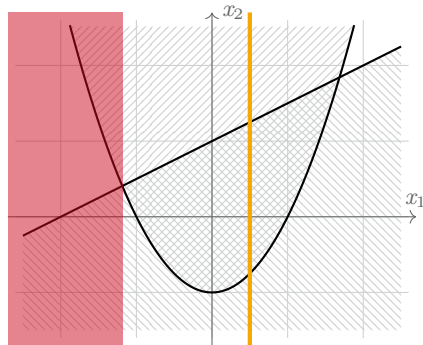
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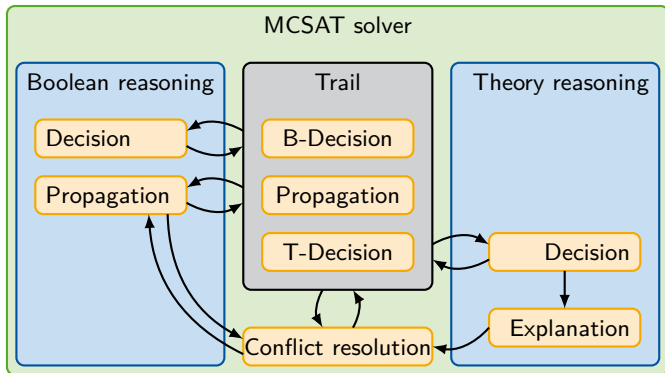
Guess: $x_1 \mapsto 0.5$

Trail: $[[c_1, c_2, x_1 \mapsto 0.5]]$



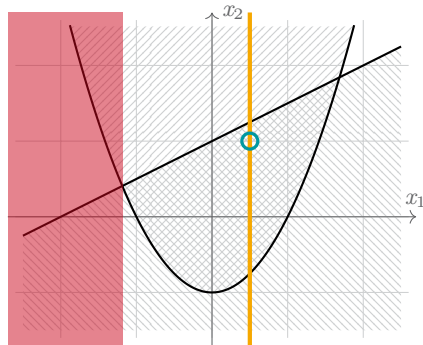
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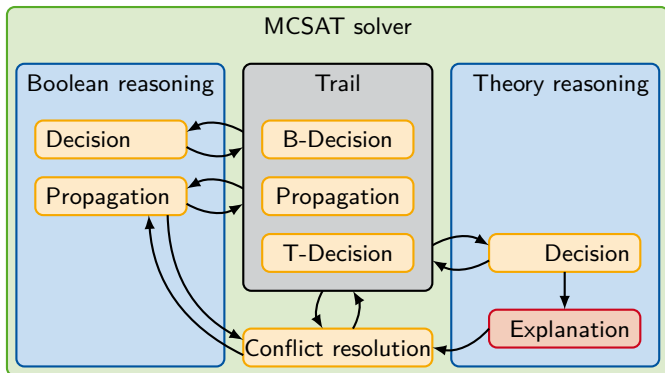
Guess: $x_2 \mapsto 1$

Trail: $\llbracket c_1, c_2, x_1 \mapsto 0.5, x_2 \mapsto 1 \rrbracket$



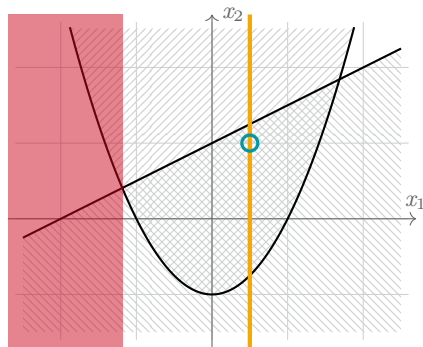
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Model-Constructing Satisfiability Calculus



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[Jovanović + 2012] [Moura + 2013]

Explanation functions

Definition (Explanation function)

Let $\mathcal{A} = \{x_1 \mapsto \alpha_1, \dots, x_k \mapsto \alpha_k\}$ and $C = \{c_1, \dots\}$ constraints over x_1, \dots, x_{k+1} .
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Model-based CAD [Jovanović⁺ 2012], Fourier-Motzkin [Jovanović⁺ 2013]

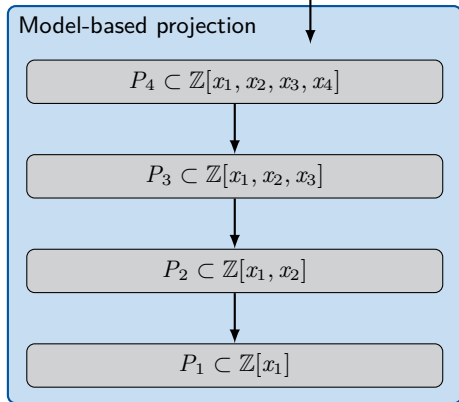
CAD-based / NLSAT

Conflicting constraints

[Jovanović + 2012]

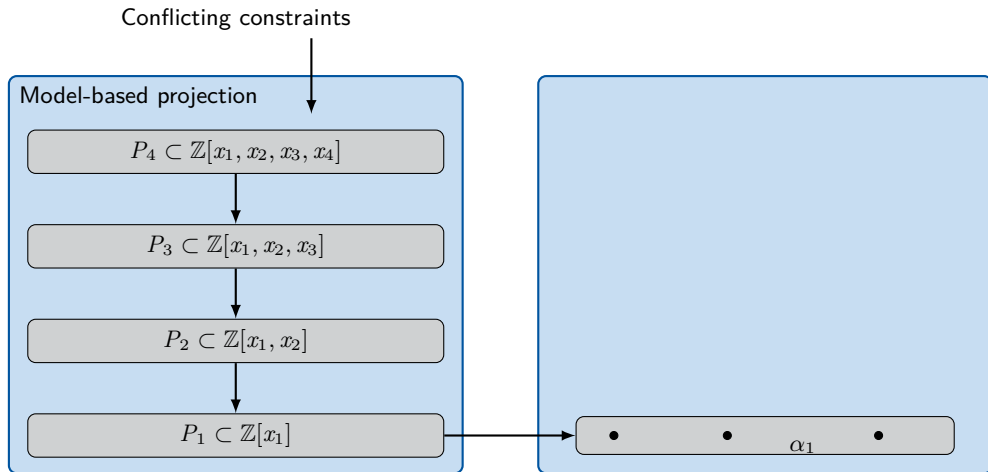
CAD-based / NLSAT

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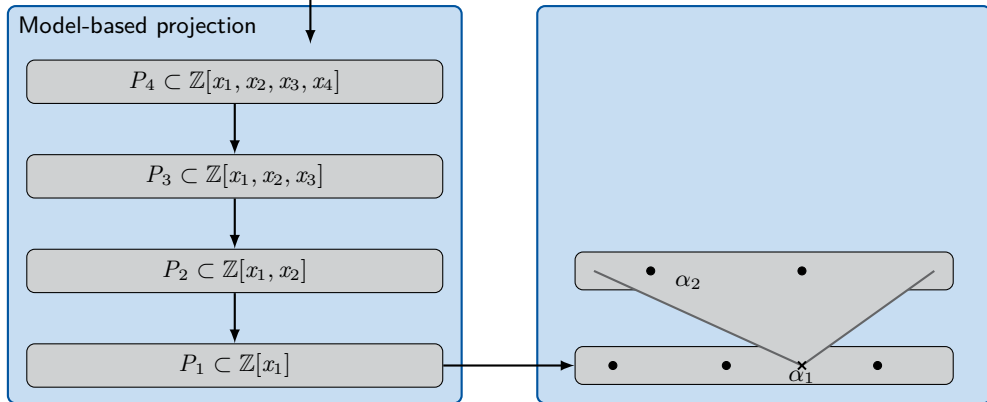
CAD-based / NLSAT



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CAD-based / NLSAT

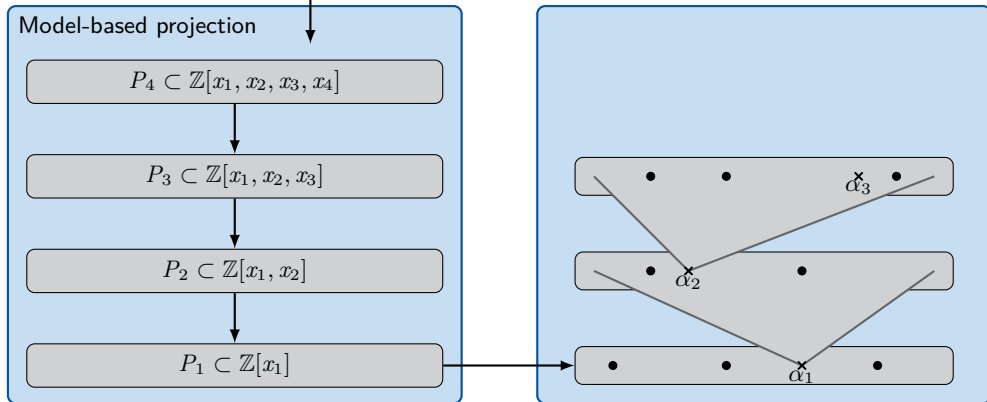
Conflicting constraints



[Jovanović+ 2012]

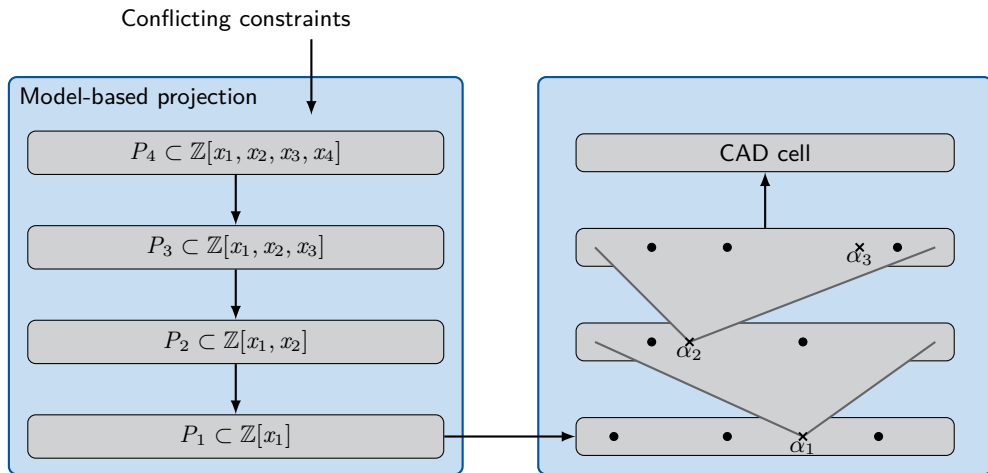
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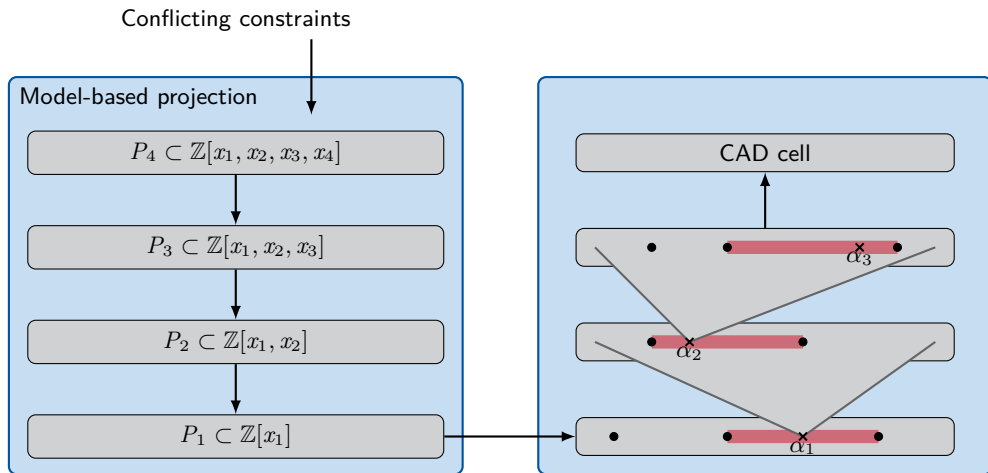
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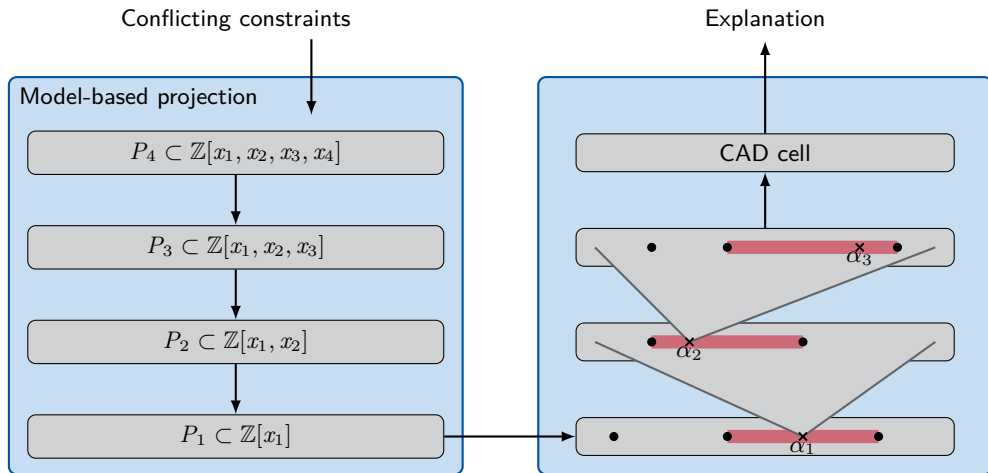
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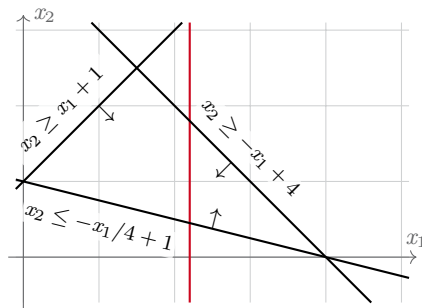
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Fourier-Motzkin

[Bartolomé 2018]

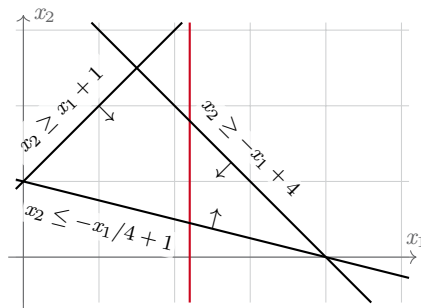


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Conflicting constraints



[Jovanović + 2013]

Fourier-Motzkin

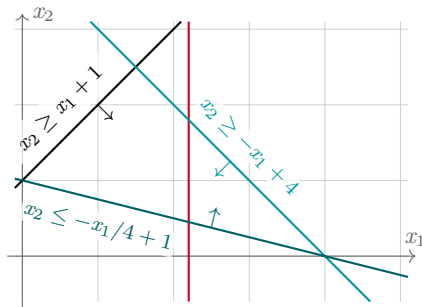
[Bartolomé 2018]

Conflicting constraints



Identify lower and upper bounds

$$-x_1 + 4 \leq x_2 \leq -x_1/4 + 1$$



[Jovanović + 2013]

Fourier-Motzkin

[Bartolomé 2018]

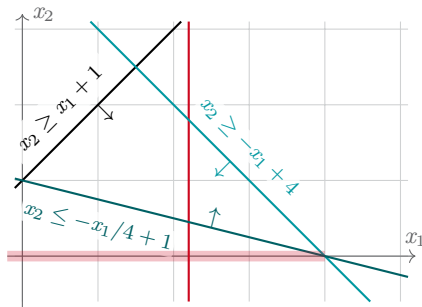
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Combine bounds

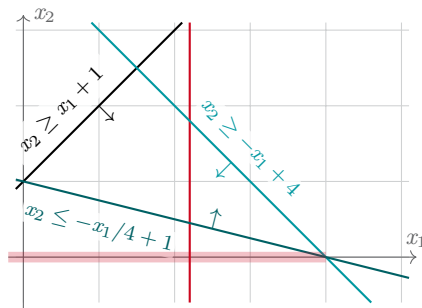
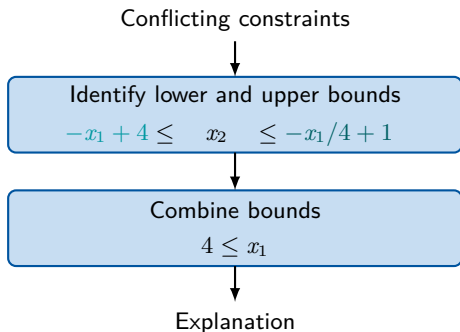
$$4 \leq x_1$$



[Jovanović + 2013]

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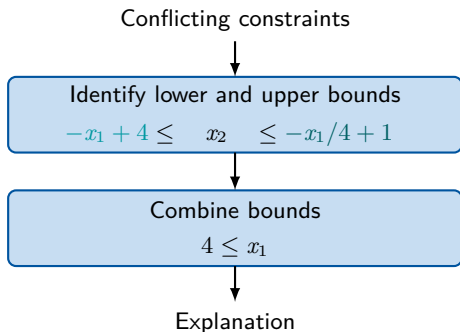
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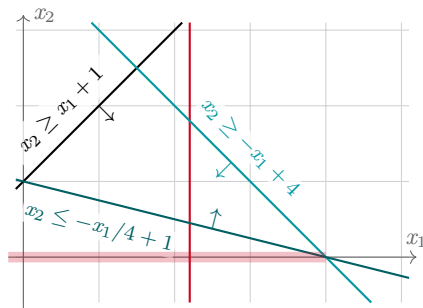
Fourier-Motzkin

[Bartolomé 2018]



- Strict inequalities
- Disequalities
- Nonlinear coefficients

[Jovanović + 2013]



$$2x_1 + 1 < x_2 < x_1 + 2 \quad x_1 \mapsto 2$$

$$2x_1 + 1 \leq x_2 \leq x_1 + 2 \wedge x_2 \neq 3 \quad x_1 \mapsto 1$$

$$x_1^2 + 3 \leq x_2 \leq 2x_1 \quad x_1 \mapsto 1$$

Other explanation functions

Virtual substitution [SC² 2017]

- Nonlinear but low-degree
- Flexible dimensionality of generated cells

[Brown 2013] [Brown + 2015]

Other explanation functions

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OneCell [Neuß 2018]

- CAD tailored to construct a single cell
- Larger cells (and thus excludes more)

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Composition of explanation functions

- SMT-RAT: modular composition
- Sequential & parallel

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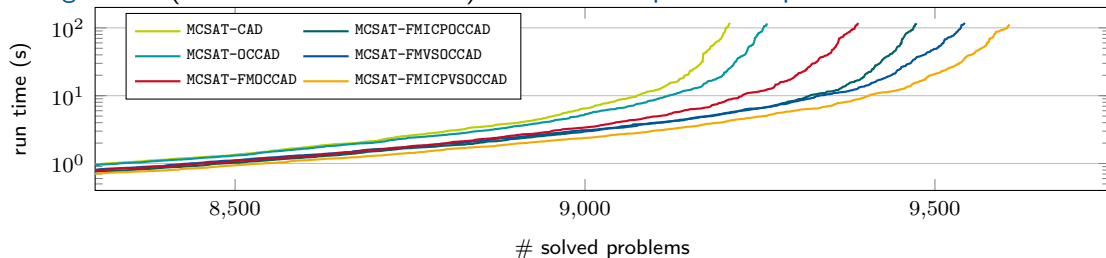
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Theory variable ordering

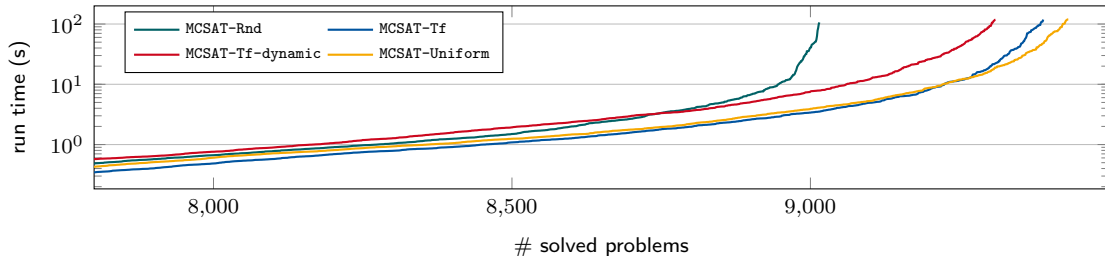
[SC² 2019]

- Well-known: variable ordering is crucial for CAD, VS, FM, ...
- Usually: **static** ordering for theory reasoning
- Goal: use **dynamic** ordering for MCSAT [Jovanović⁺ 2013]

Theory variable ordering

[SC² 2019]

- Well-known: variable ordering is crucial for CAD, VS, FM, ...
- Usually: **static** ordering for theory reasoning
- Goal: use **dynamic** ordering for MCSAT [Jovanović + 2013]
- Observation: MCSAT-Tf and MCSAT-Uniform perform best
- But: MCSAT-Tf-dynamic is worse



Comparison of lazy SMT and MCSAT

[SC² 2019]

- Two **similar proof systems**: lazy SMT and MCSAT
- Experience: MCSAT **better for nonlinear**
- Goal: **theoretic** comparison of lazy SMT and MCSAT

Comparison of lazy SMT and MCSAT

[SC² 2019]

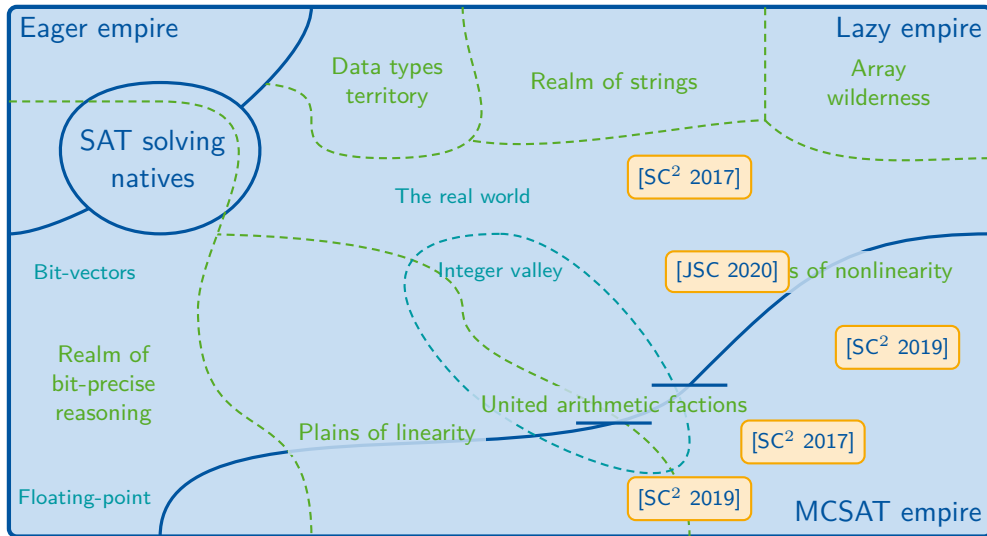
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Theorem

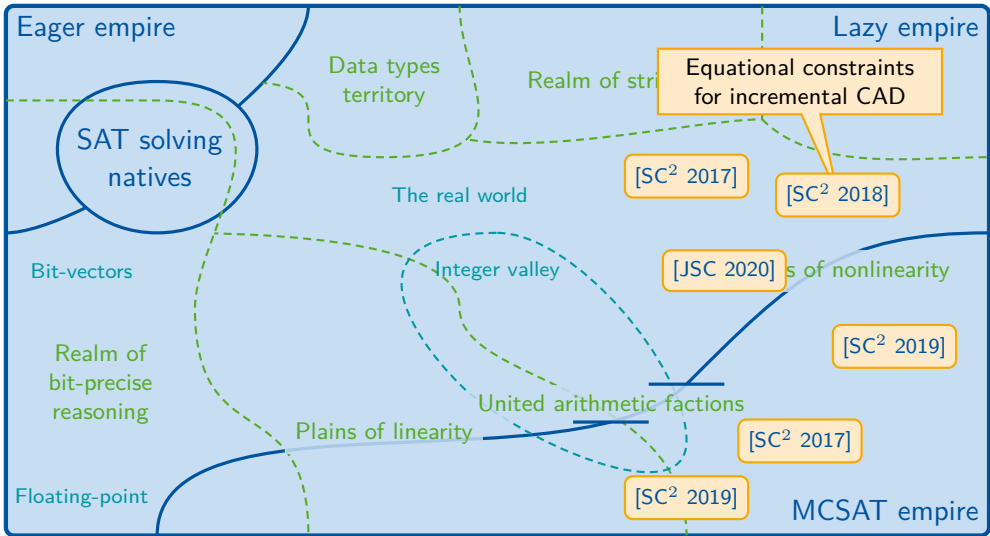
*Lazy SMT and MCSAT are **algorithmically equivalent****.

*: terms and conditions apply

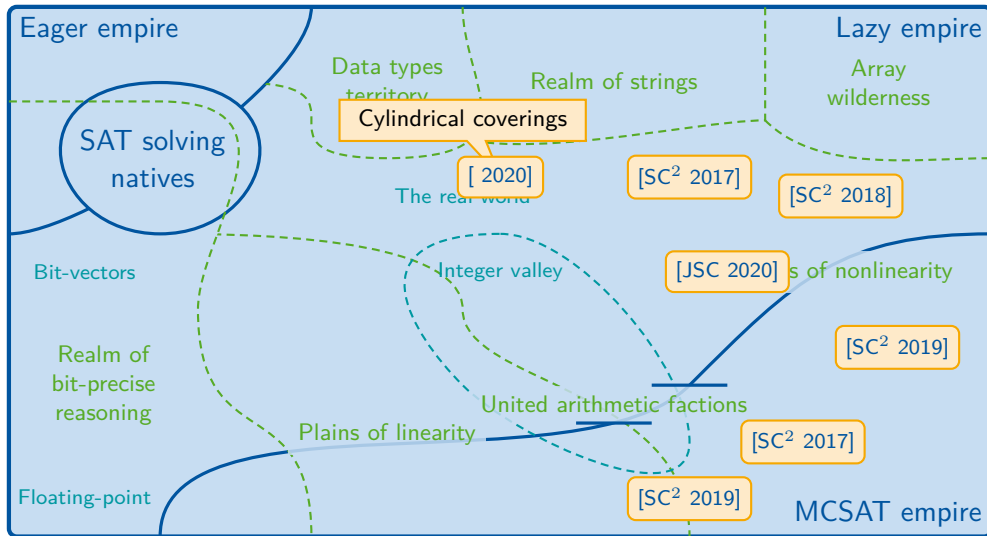
Summary



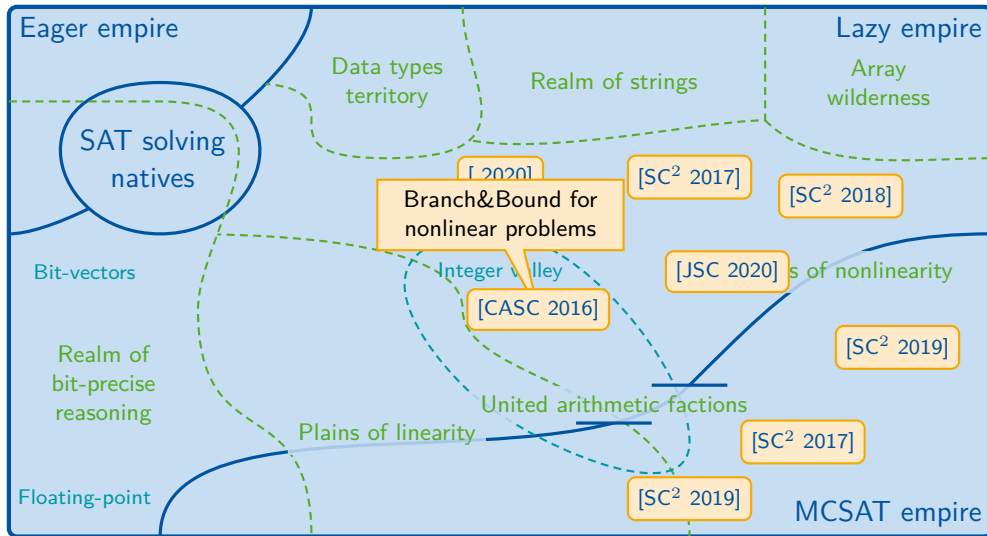
Summary



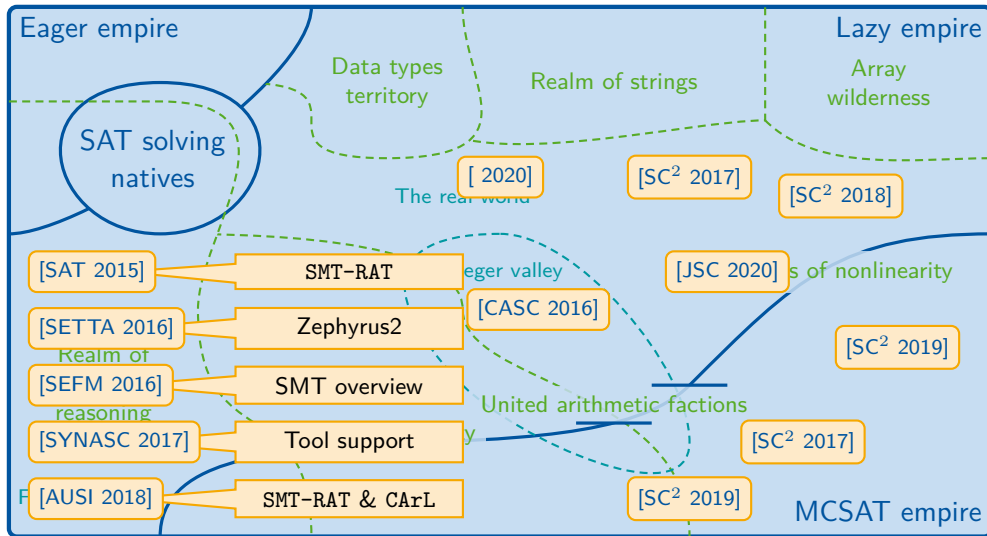
Summary



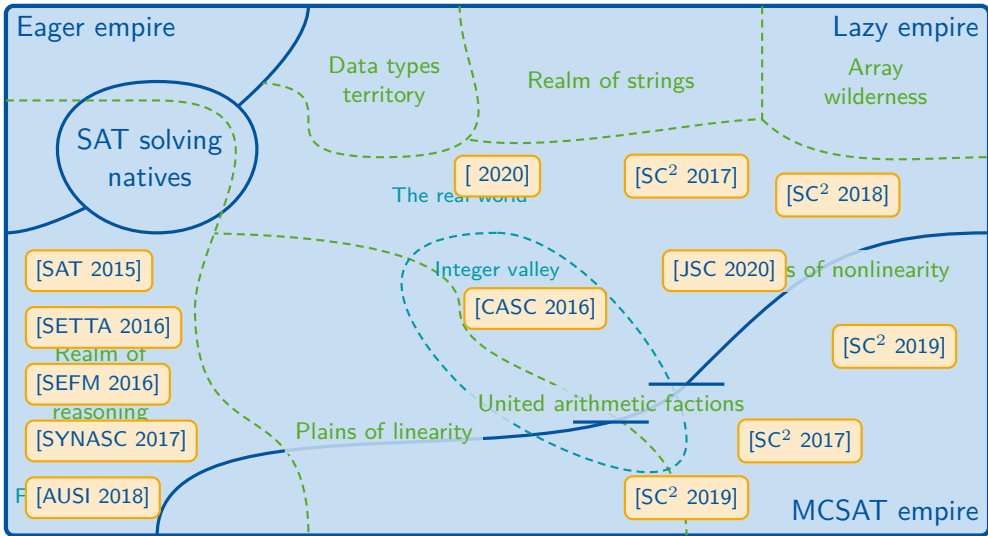
Summary



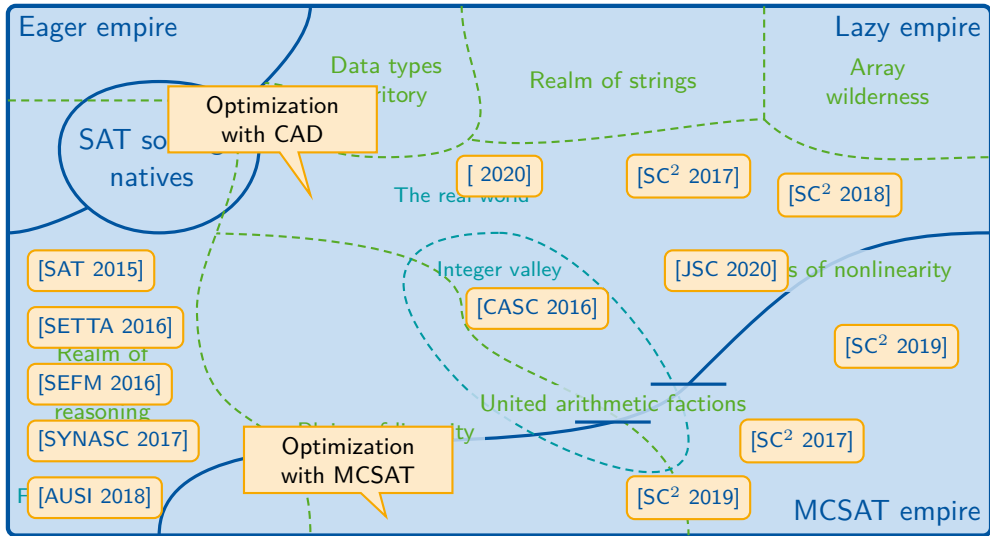
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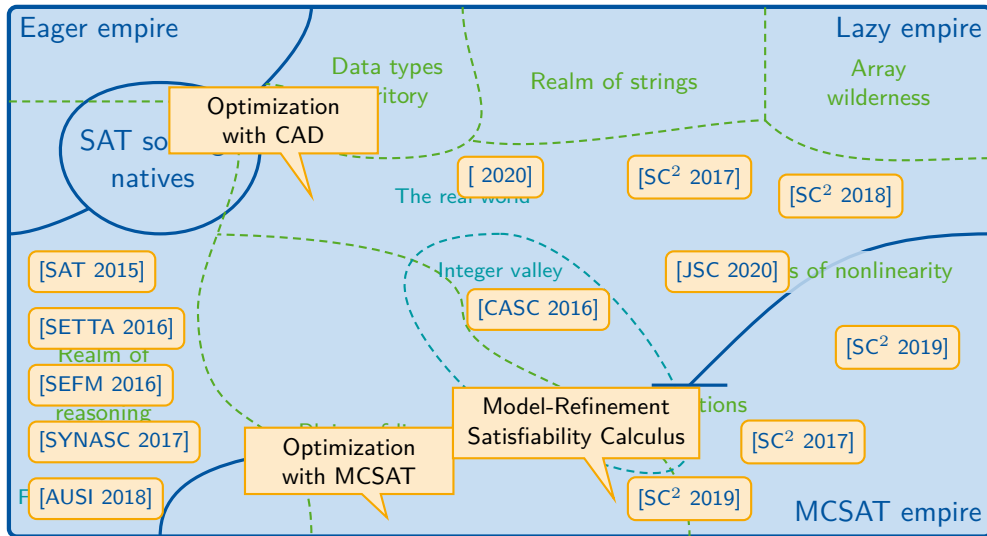
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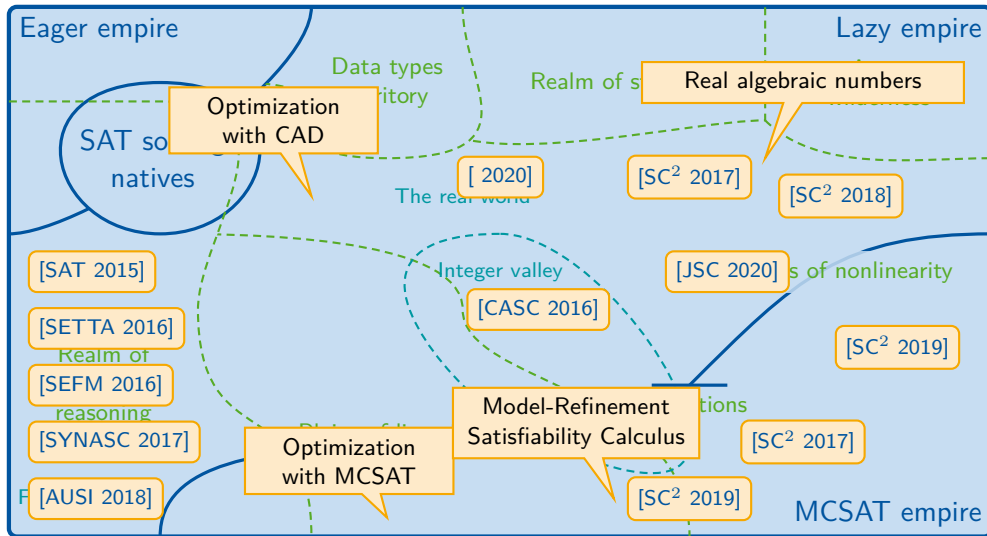
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Summary



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- CAD is **suitable for lazy SMT solving**
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Directions for future research:

- **Optimize CAD** integration: better scheduling, NuCAD, parallelization, ...
- **Novel adaptations** of CAD: CAC [2020], NuCAD [Brown 2015], ...
- **New methods** for MCSAT: explanations, orderings, assignment finders, ...
- **More applications**: optimization, dedicated heuristics, ...

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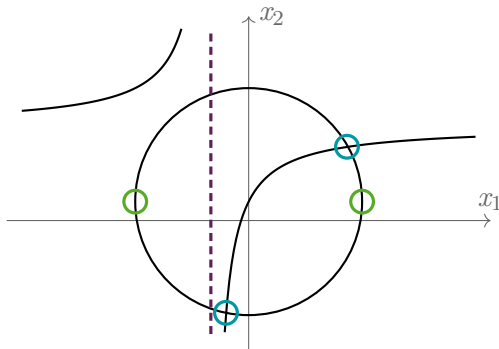
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CAD projection

Example: McCallum

$$\begin{aligned} Proj(P) = & \{ \text{coeffs}(p), \text{disc}(p) \mid p \in P \} \\ & \cup \{ \text{res}(p, q) \mid p, q \in P \} \end{aligned}$$



[McCallum 1985]

Minimal Infeasible Subsets

[Hentze 2017]

- Set of constraints $\{c_1, c_2, \dots\}$ is infeasible

[Jaroschek + 2015] [Chvátal 1979]

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- Set of constraints $\{c_1, c_2, \dots\}$ is infeasible
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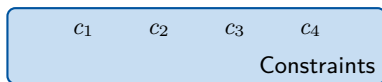
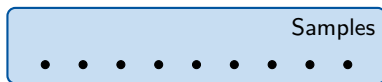
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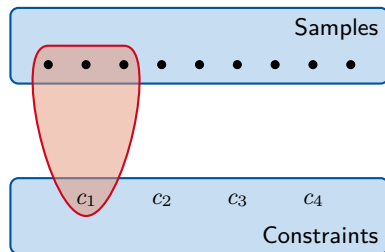


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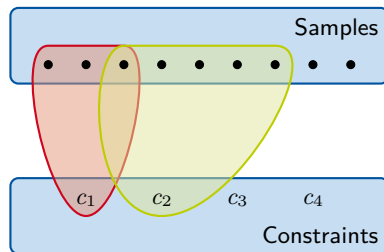


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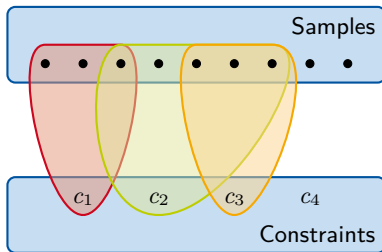


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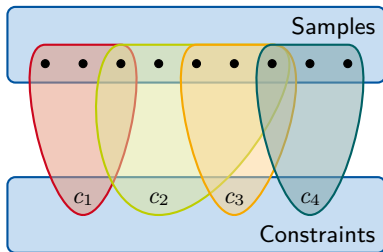


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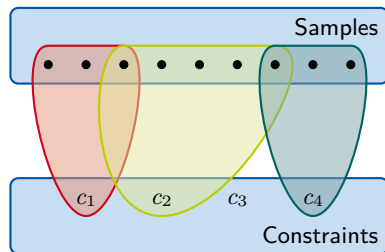


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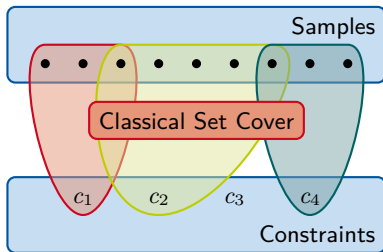


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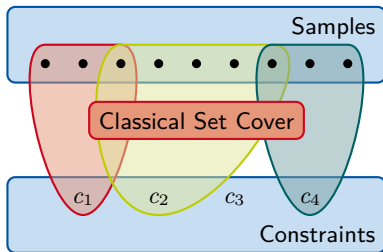


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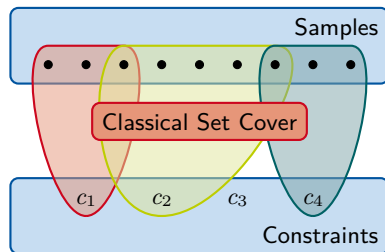


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- Significant improvements for MIS ...
- but no effect on solver performance



Solver	SAT	UNSAT	overall
Greedy-PP	4327	4249	8576
Greedy-Weighted	4329	4247	8576
Hybrid	4328	4248	8576
Trivial	4325	4251	8576
Exact	4328	4249	8577
Greedy	4328	4250	8578

[Jaroschek + 2015] [Chvátal 1979]

Using CAD for ...

...Quantifier Elimination

[Neuhäuser 2018]

$$(\exists x. \forall y. \varphi(x, y, z)) \Leftrightarrow \varphi'(z)$$

- **Validation** of our CAD implementation
- Show **reuse** of implementation
- Results comparable to QEPCAD B

[Brown 2003] [Fuhs + 2007]

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...Integer problems

[CASC 2016]

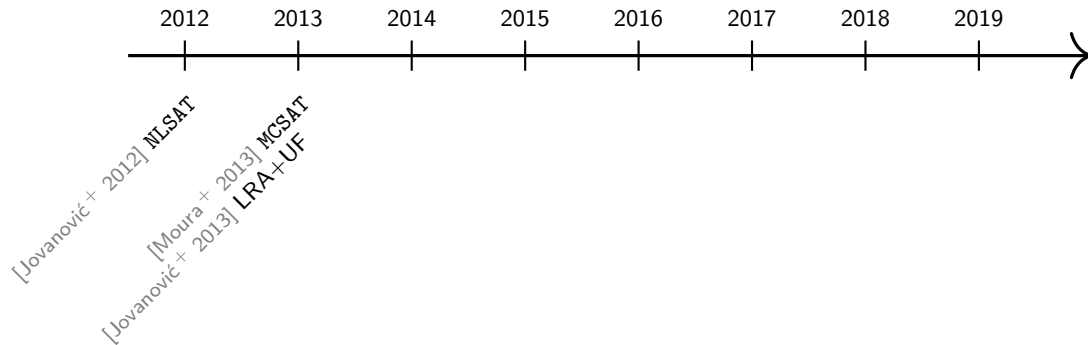
$$\exists x. \varphi(x) \quad x \in \mathbb{Z}$$

- Common approach: **bit-blasting**
mostly aims for **small domains**
- Our approach: **Branch & Bound**
mostly aims for **algebraically easy** problems & UNSAT

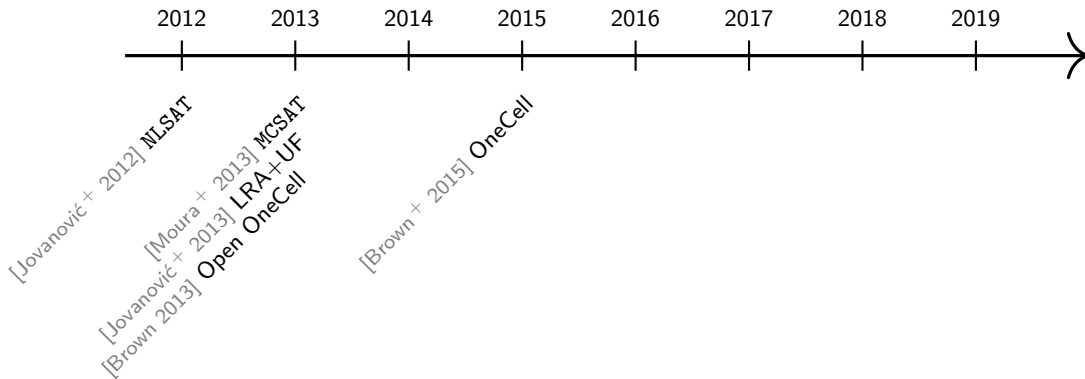
Solver	SAT	UNSAT	overall
NIA-B&B	1044	518	1562
NIA-Blast	4367	26	4393
NIA-Full	4490	508	4998

[Brown 2003] [Fuhs + 2007]

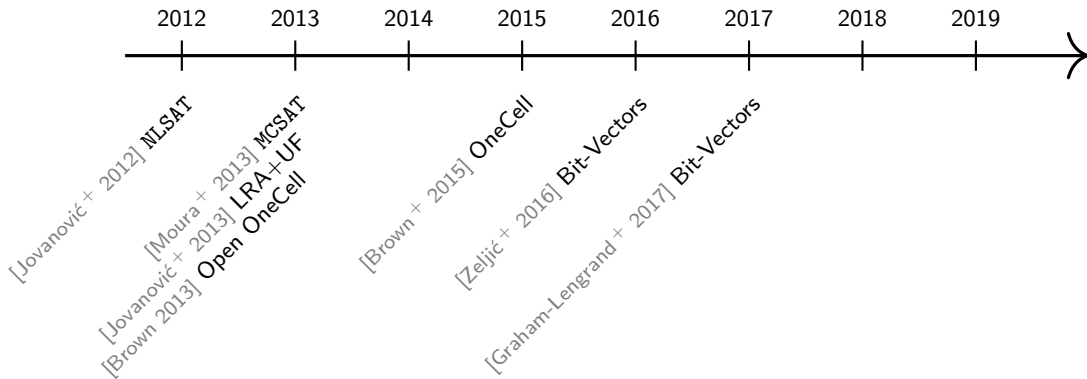
MCSAT history



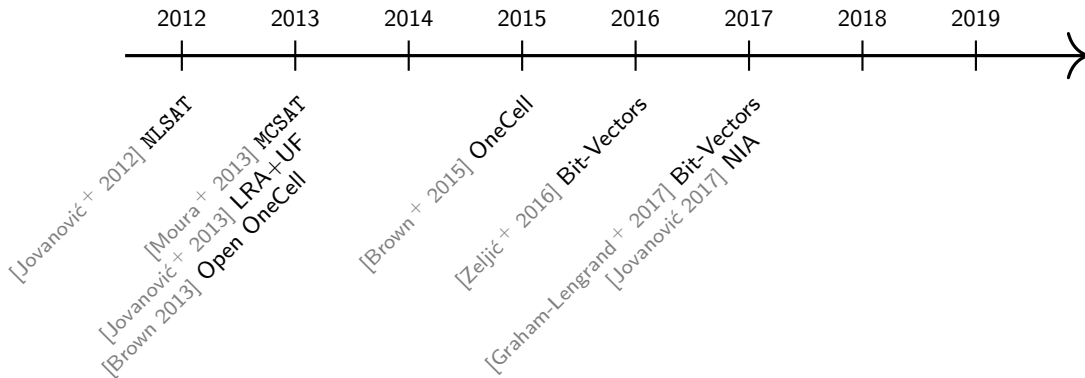
MCSAT history



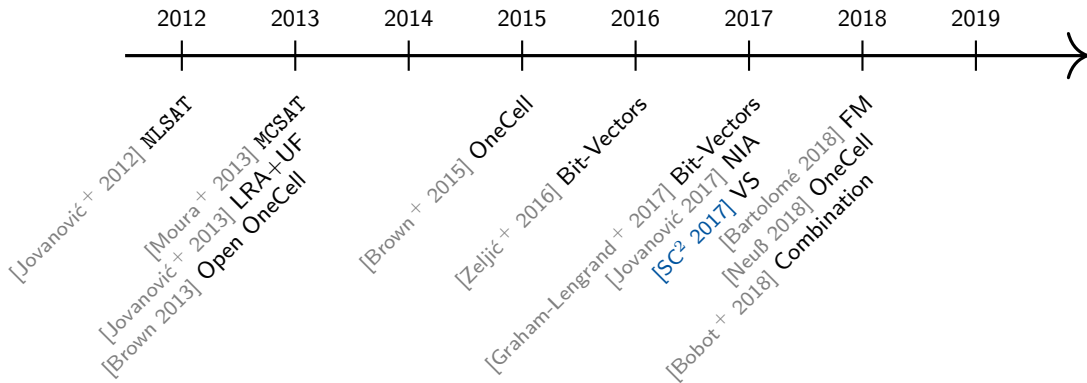
MCSAT history



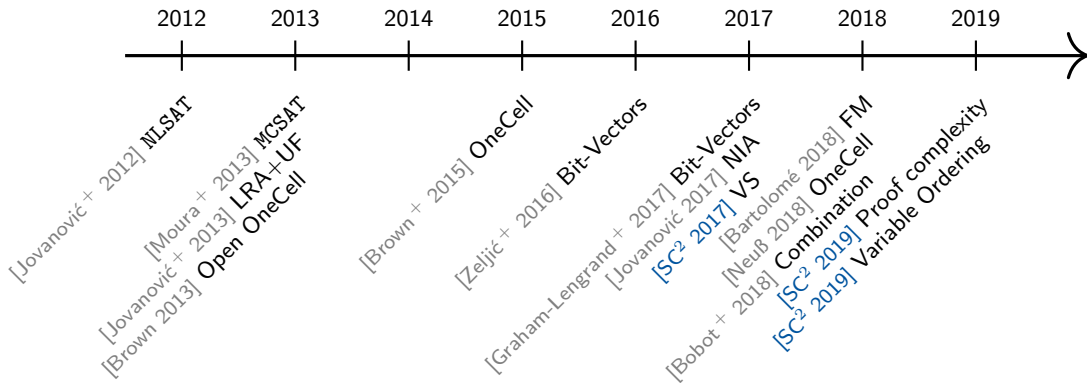
MCSAT history



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Theory decisions

Regular assignment finder

- Collect **univariate** constraints
- Compute **sign-invariant regions**
- Sample from **satisfying region**
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- **Larger look ahead**

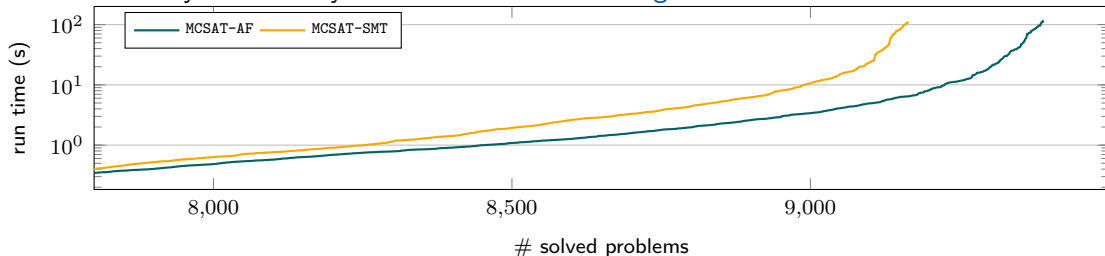
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[SC² 2019]

- Two **similar proof systems**: CDCL*(T) and MCSAT
- Experience: MCSAT **better for nonlinear**
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- **Proof complexity**: asymptotic length of shortest proof
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- The proof can be **completely different**
- **Algorithmic equivalence**: polynomial simulation of another method
- CDCL*(T) and MCSAT **simulate each other***
- The proofs are **logically equivalent**

*: terms and conditions apply