

Exotic Semiring Constraints

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Example: Arctic Semiring Constraints

arctic semiring: domain $A = \{-\infty\} \cup \mathbb{N}$,

$x \oplus y = \max(x, y)$, $x \otimes y = x + y$.

example constraint system:

$$(a_{11} \geq 0) \wedge (b_{11} \geq 0) \wedge \bigwedge_{\substack{i \in \{1,2\} \\ j \in \{1,2\}}} \left\{ \begin{array}{l} (c_{ij} = (a_{i1} \otimes b_{1j}) \oplus (a_{i2} \otimes b_{2j})) \\ \wedge (((a_{i1} \otimes a_{1j}) \oplus (a_{i2} \otimes a_{2j}) > (c_{i1} \otimes a_{1j}) \oplus (c_{i2} \otimes a_{2j})) \\ \vee (((a_{i1} \otimes a_{1j}) \oplus (a_{i2} \otimes a_{2j}) = -\infty) \\ \wedge ((c_{i1} \otimes a_{1j}) \oplus (c_{i2} \otimes a_{2j}) = -\infty))) \\ \wedge (b_{ij} \geq b_{i1} \otimes b_{1j} \oplus b_{i2} \otimes b_{2j}) \end{array} \right.$$

imagine this with 100 ... 1000 unknowns

hard for DPLL(T) because \oplus introduces disjunctions,

$-\infty$ introduces case distinctions (in \oplus and \otimes)

Our solution: unary bit-blasting.

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Where do these constraints occur?

the framework is *exotic semirings*, examples:

- ▶ arctic $\{-\infty\} \cup \mathbb{Z}$, max, +
- ▶ tropical $\mathbb{N} \cup \{+\infty\}$, min, +
- ▶ fuzzy $\{-\infty\} \cup \mathbb{N} \cup \{+\infty\}$, min, max

applications:

- ▶ formal languages (star height problem) (Imre Simon 1988)
- ▶ idempotent analysis
- ▶ disjunctive invariants in static analysis
- ▶ automated analysis of termination of programs (modelled as rewriting systems)

Arctic Matrices for Termination

$\mathcal{R} = \{ a a \rightarrow a b a \}$ and $\mathcal{S} = \{ b \rightarrow b b \}$

to show relative termination of \mathcal{R} w.r.t. \mathcal{S}

(no $\mathcal{R} \cup \mathcal{S}$ -derivation with infinitely many \mathcal{R} steps)

interpret symbols by matrices $a \mapsto A, b \mapsto B$

with $A_{1,1} \geq 0 \wedge B_{1,1} \geq 0 \wedge (A^2 >_0 ABA) \wedge (B \geq B^2)$.

where $(x >_0 y)$ is $x > y \vee (x = -\infty = y)$

matrix dimension 2 gives constraints from intro slide where c_{ij} is contents of $C = AB$,

$$\text{one solution is } A = \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix}, B = \begin{pmatrix} 0 & -\infty \\ 0 & -\infty \end{pmatrix}.$$

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Topics of this talk

- ▶ definition and motivation of exotic semiring constraints
- ▶ solving by translation (QF_LIA, QF_IDL)
- ▶ solving by unary bitblasting:
 - ▶ naturals, integers, exotic numbers
- ▶ implementation, empirical evaluation
 - ▶ the “killer” example
 - ▶ comparison of different approaches

Translation to QF_LIA, QF_IDL

Represent exotic number as pair of “(minus/plus) infinity” (a boolean) and “contents” (a number).

arctic multiplication (plus): $(m, c) = \otimes_k(m_k, c_k)$ iff
 $(m = \bigvee_k m_k) \wedge (\neg m \rightarrow (c = \sum_k c_k))$.

arctic addition (max): $(m, c) = \oplus_k(m_k, c_k)$ iff
 $(m = \bigwedge_k m_k) \wedge (\neg m \rightarrow \bigwedge_k (\neg m_k \rightarrow c \geq c_k))$
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For fuzzy semiring constraints, operations are min and max (and no +), so encoding goes to QF_IDL.

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Unary Bit-Blasting: Order Encoding

translation QF_LIA to SAT (sound, incomplete):

- ▶ restrict to finite domain $\{0, 1, \dots, B\}$
- ▶ number $x \Rightarrow$ monotone list of booleans $[x_1, \dots, x_B]$ where $x_i \leftrightarrow (x \geq i)$,
e.g., $3 = [1, 1, 1, 0, 0, 0, 0]$
- ▶ arithmetical operations \Rightarrow boolean functions

unary encoding is highly redundant,
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Order-Encoded Operations

should prefer conjunctive encodings.
Example: comparison of k -bit unary numbers,
e.g., $a = \langle 1, 1, 1, 0 \rangle, b = \langle 1, 1, 0, 0 \rangle$.

- ▶ this is easy: $a \geq b \iff \bigwedge_k \{a_k \leftarrow b_k\}$
- ▶ by negation: $a > b \iff \bigvee_k \{a_k \wedge \neg b_k\}$
- ▶ this is equivalent (by monotonicity)
and we think it is better, since it is a conjunction:
 $a > b \iff a_1 \wedge \neg b_k \wedge \bigwedge_k \{a_{k+1} \leftarrow b_k\}$

no hard evidence.
also depends on whether we want to assert $a > b$,
or have it as a subformula, e.g., assert $((a > b) \vee \dots)$

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- ▶ max/min by point-wise \vee/\wedge (linear formula size).
- ▶ addition: let $a = \langle a_1, \dots, a_k \rangle, b = \langle b_1, \dots, b_{k'} \rangle$. Then $a + b = c$ where $c = \langle c_1, \dots, c_{k+k'} \rangle$ with

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Extensions

- ▶ integers: shift the encoding. transform $x \in \{-k + 1, \dots, k\}$ to $x + k$ and encode as natural. keep min and max, modify + (shift back)
- ▶ exotic numbers: use one extra bit for $-\infty, +\infty$ (either one for arctic and tropical, both for fuzzy) keep monotonicity, \Rightarrow keep min and max
- ▶ Overflows: are not allowed (otherwise unsound) Either increase bit width as needed (in addition), or keep bit width and assert " \neg overflow".

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(detect instances of $x \leftrightarrow y$ or $x \leftrightarrow \neg y$ and propagate)
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The “Killer” Example

Termination benchmark SRS/Gebhardt/19
{0000 \rightarrow 1011, 1001 \rightarrow 0010} (open since 2006)

is terminating since tropical matrix constraint system

$$\begin{aligned} 0_{\#}0^3 &\geq 1_{\#}01^2 \wedge 1_{\#}0^21 \geq 0_{\#}010 \\ \wedge 0^4 &\geq 101^2 \wedge 10^21 \geq 0^210 \\ \wedge (0^4 >_0 101^2 \vee 10^21 >_0 0^210). \end{aligned}$$

is solvable for 8×8 , minisat needs one hour.

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{0000 \rightarrow 1011, 1001 \rightarrow 0010} (open since 2006)
is terminating since tropical matrix constraint system

$$\begin{aligned} 0_{\#}0^3 &\geq 1_{\#}01^2 \wedge 1_{\#}0^21 \geq 0_{\#}010 \\ \wedge 0^4 &\geq 101^2 \wedge 10^21 \geq 0^210 \\ \wedge (0^4 >_0 101^2 \vee 10^21 >_0 0^210). \end{aligned}$$

is solvable for 8×8 , minisat needs one hour.

Experimental Results

using solvers satchmo-smt, Bee, Z3
on exotic constraints from termination problems

- ▶ 3 bit binary vs. 7 bit unary (equal range)
outcome: unary is better
- ▶ Z3 (with DPLL(Simplex)?) vs. unary (with iterative deepening = increasing bit width)
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- ▶ unary: straightforward (satchmo-smt) vs. preprocessed (Bee)
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(minisat preprocessor will run anyway)

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Lots of disjunctions and booleans.

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