## **Exotic Semiring Constraints**

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#### SMT12

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arctic semiring: domain  $A = \{-\infty\} \cup \mathbb{N}$ ,  $x \oplus y = \max(x, y), x \otimes y = x + y$ .

example constraint system:

$$(a_{11} \geq 0) \land (b_{11} \geq 0) \land \\ \begin{pmatrix} (c_{ij} = (a_{i1} \otimes b_{1j}) \oplus (a_{i2} \otimes b_{2j})) \\ \land (((a_{i1} \otimes a_{1j}) \oplus (a_{i2} \otimes a_{2j}) > (c_{i1} \otimes a_{1j}) \oplus (c_{i2} \otimes a_{2j})) \\ \lor (((a_{i1} \otimes a_{1j}) \oplus (a_{i2} \otimes a_{2j}) = -\infty) \\ \land ((c_{i1} \otimes a_{1j}) \oplus (c_{i2} \otimes a_{2j}) = -\infty))) \\ \land (b_{ij} \geq b_{i1} \otimes b_{1j} \oplus b_{i2} \otimes b_{2j}) \end{pmatrix}$$

imagine this with 100 ... 1000 unknowns

hard for DPLL(T) because  $\oplus$  introduces disjunctions,  $-\infty$  introduces case distinctions (in  $\oplus$  and  $\otimes$ )

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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)

 $\mathcal{R} = \{ aa \rightarrow aba \} \text{ and } \mathcal{S} = \{ b \rightarrow bb \}$  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,

one solution is 
$$A = \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix}$$
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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

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

arctic multiplication (plus): 
$$(m, c) = \bigotimes_k (m_k, c_k)$$
 iff  $(m = \bigvee_k m_k) \land (\neg m \to (c = \sum_k c_k))$ .  
arctic addition (max):  $(m, c) = \bigoplus_k (m_k, c_k)$  iff  $(m = \bigwedge_k m_k) \land (\neg m \to \bigwedge_k (\neg m_k \to c \ge c_k))$ 

For fuzzy semiring constraints, operations are min and max (and no +), so encoding goes to QF\_IDL.

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#### translation QF\_LIA to SAT (sound, incomplete):

- ▶ restrict to finite domain {0, 1, ..., B}
- number  $x \Rightarrow$  monotone list of booleans  $[x_1, \ldots, x_B]$  where  $x_i \leftrightarrow (x \ge i)$ , e.g., 3 = [1, 1, 1, 0, 0, 0, 0, 0]
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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:  $a \ge b \iff \bigwedge_k \{a_k \leftarrow b_k\}$
- ▶ by negation:  $a > b \iff \bigvee_{k} \{a_k \land \neg b_k\}$
- this is equivalent (by monotonicity) and we think it is better, since it is a conjunction: a > b ⇔ a₁ ∧ ¬b<sub>k</sub> ∧ ∧<sub>k</sub>{a<sub>k+1</sub> ← b<sub>k</sub>}

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- ► max/min by point-wise ∨/∧ (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

$$\bigwedge_{\substack{0 < i \leq k, \\ 0 < j \leq k'}} \left\{ 
\begin{array}{l}
(a_i \to c_i) \wedge (\neg a_i \to \neg c_{k'+i}) \wedge \\
(b_j \to c_j) \wedge (\neg b_j \to \neg c_{k+j}) \wedge \\
(a_i \wedge b_j \to c_{i+j}) \wedge (\neg a_i \wedge \neg b_j \to \neg c_{i+j-1})
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#### **Extensions**

- integers: shift the encoding.
   transform x ∈ {-k + 1,...,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
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- low level: boolean equipropagation [MCLS11] (Bee)
   (detect instances of x ↔ y or x ↔ ¬y and propagate)
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   (Matchbox)
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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

$$\begin{array}{c} 0_{\#}0^{3} \geq 1_{\#}01^{2} \wedge 1_{\#}0^{2}1 \geq 0_{\#}010 \\ \wedge \ 0^{4} \geq \ 101^{2} \wedge 10^{2}1 \geq 0^{2}10 \\ \wedge \ (0^{4} >_{0} 101^{2} \ \lor \ 10^{2}1 >_{0} 0^{2}10). \end{array}$$

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## **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) outcome: unary is better
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