

# Lecture 5: Satisfiability modulo theories

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Slides adapted from Prof. Sayan Mitra's slides in Fall 2021

Some of the slides for this lecture are adapted from slides by Clark Barrett

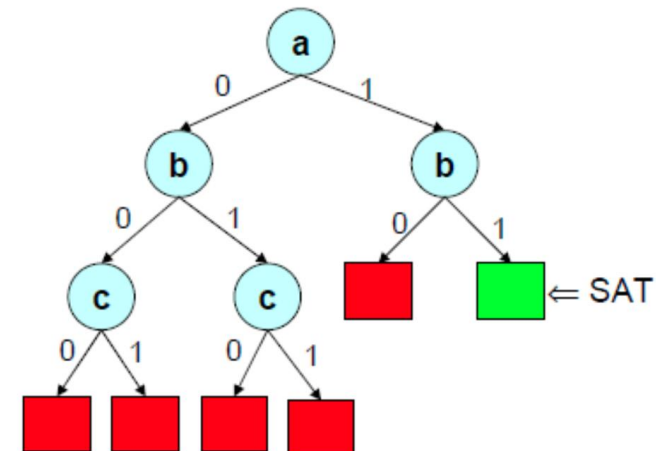
# Previous lectures

- Boolean satisfiability problem

$$\alpha := (\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_4) \wedge (\neg x_2 \vee \neg x_3 \vee \neg x_4)$$

DPLL algorithm (backtracking + unit propagation + pure literal assignment)

```
function DPLL( $\alpha$ )  
    unit-propagate  
    pure-literal-assign  
    check-stopping-conditions  
     $l \leftarrow$  choose-literal( $\alpha$ );  
return (DPLL( $\alpha \wedge \{l\}$ ) or  
        DPLL( $\alpha \wedge \{\neg l\}$ ));
```



# Today

- Satisfiability modulo theories (SMT): more general satisfiability problem
  - Theories, models, decision procedures
  - Uninterpreted Functions
  - Difference Logic
- Brief z3 tutorial (see notebook)

# Satisfiability modulo theories

- SAT: Given a *well-formed formula* in propositional logic, determine whether there exists a satisfying solution

$$\alpha := (\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_4) \wedge (\neg x_2 \vee \neg x_3 \vee \neg x_4)$$

- A *satisfiability modulo theory (SMT)* problem is a generalization of SAT in which some of the binary variables are replaced by predicates over a suitable set of non-binary variables

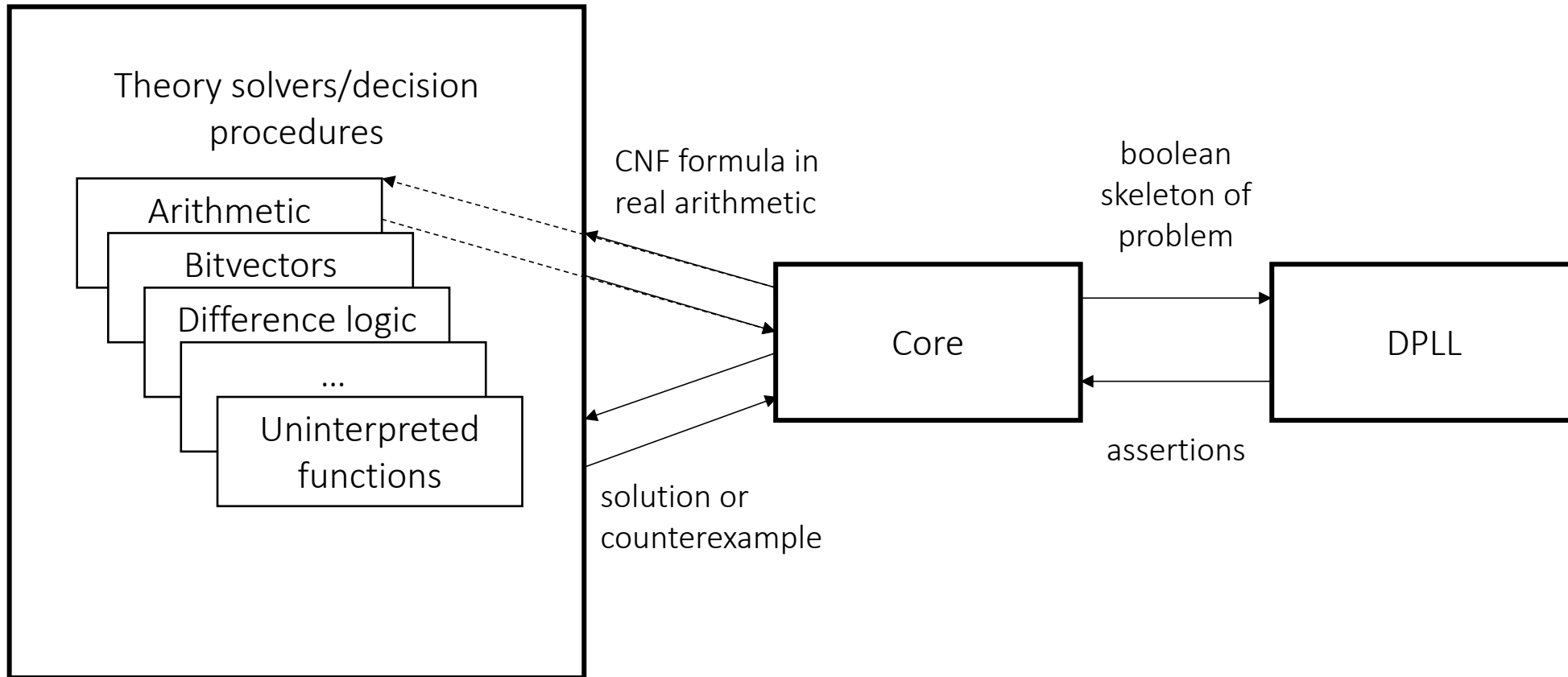
# Satisfiability modulo theories

- SAT: Given a *well-formed formula* in propositional logic, determine whether there exists a satisfying solution

$$\alpha := (\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_4) \wedge (\neg x_2 \vee \neg x_3 \vee \neg x_4)$$

- A *satisfiability modulo theory (SMT)* problem is a generalization of SAT in which some of the binary variables are replaced by predicates over a suitable set of **non-binary variables**
- $\phi_1(w, x, y, z) := (x - y = 5) \wedge (z - y \geq 2) \wedge (z - x > 2) \wedge (w - x = 2)$
- $\phi_2(x, y, z) := (3x^2 - 4y + 5z \leq 5) \wedge (-2x + 5z^3 \leq 7)$
- $\phi_1$  is a predicate in *difference logic* in which the variables are real-valued, and the clauses are constructed with standard comparison operations  $>$ ,  $\geq$ ,  $=$  and  $-$  (minus)
- $\phi_2$  is a predicate in real arithmetic

# Architecture of an SMT solver



# *⟨model theory⟩*

A short overview of theories, models, decision procedures

# What is a theory in mathematical logic?

- When we talk about **well-formed formulas with non-binary variables**, we have to say exactly what type of formulas are allowed
- and, what it means for assignments to *satisfy such formulas*
- This brings us to some basic notions in mathematical logic
  - **theory** --- what does a well-formed formula look like ?
  - **models** --- what does it mean to satisfy a formula?



# Building up a theory

First, we define the syntax for writing **formulas**

A *signature*  $\Sigma = (\Sigma_F, \Sigma_P, V)$

- $\Sigma_F$  : set of *function symbols*, e.g.,  $\{+, -, f, g, \sin, \dots\}$
- $\Sigma_P$  : set of *predicate symbols*, e.g.,  $\{<, >, <=, \dots\}$ 
  - *arity* of each function:  $\text{arity}: \Sigma_F \rightarrow \mathbb{N}$
  - *0 arity* functions are constants
- $V$ : set of *variables*

*Terms*( $\Sigma, V$ ): combines variables and functions

- Elements of  $V$  are terms
- If  $t_1, \dots, t_k \in \text{Terms}(\Sigma, V)$  and  $f \in \Sigma_F$  with arity  $k$ , then  $f(t_1, \dots, t_k) \in \text{Terms}(\Sigma, V)$
- *Ground terms* are terms without variables

- $\Sigma_F = \{0, +\}, \Sigma_P = \{<\}$
- $\text{arity}(0) = 0$
- $\text{arity}(+) = 2$
- $\text{arity}(<) = 2$
- $V = \{x, y, z\}$
  
- Terms defined by this signature are  $x, y, z, + (x, y), + (+ (x, y), 0), 0, \dots$

# Terms to Formulas

- **Atomic formulas**  $AF$ : combines terms and predicates
  - True, False
  - If  $t_1, \dots, t_k \in Terms(\Sigma, V)$  and  $p \in \Sigma_p$  with arity  $k$ , then  $p(t_1, \dots, t_k) \in AF(\Sigma, V)$
  - A *literal* is an AF or its negation
  - Set of all atomic formulas  $AF(\Sigma, V)$
- **Quantifier free formulas**  $QFF(\Sigma, V)$ 
  - $AF$
  - if  $\phi_1, \phi_2 \in QFF$  then
    - $\neg\phi_1 \in QFF, \phi_1 \wedge \phi_2 \in QFF, \phi_1 \vee \phi_2 \in QFF, \phi_1 \rightarrow \phi_2 \in QFF$
  - Set of all quantifier free formulas  $QFF(\Sigma, V)$
- **First order formulas** is the set of quantifier free formulas under universal and existential quantifiers
  - **Bound variables** are those that are attached to quantifiers
  - **Free variables**: variables not bound
- **Sentence**: First order formula with no free variables
- **Theory** $(\Sigma, V)$  set of all sentences over  $(\Sigma, V)$

AF examples:

- $x < y$
- $+ (x, y) = + (y, x)$

QFF examples:

- $+ (x, y) = 0 \wedge x > y$

First order formulas:

- $\forall x, \exists y: + (x, y) = 0$
- $\forall x, \exists y: x < y$
- $\forall x, \exists y: + (x, y) = x$

Sentence:

- $\exists x: + (x, 1) = x$

# Models for theories

This notion of model from **mathematical logic** is not to be confused with the notion of a model for a computational or physical process

- A *model* gives meanings or *interpretations* to formulas in theory  $T$
- A model  $M$  for  $T = \text{Theory}(\Sigma, V)$  has to define
  - A **domain**  $|M|$
  - interpretations of all **functions** and **predicate** symbols
  - $M(f): |M|^n \rightarrow |M|$  if  $\text{arity}(f) = n$
  - $M(p) \subseteq |M|^n$  if  $\text{arity}(p) = n$
  - Assignment  $M(x) \in |M|$  for every variable  $x \in V$
- A **formula  $\phi$  is true in  $M$**  if it evaluates to true under the given interpretations over domain  $M$

# Example

A *model* gives meanings or *interpretations* to formulas in theory  $T$

Example model for  $\Sigma = \{0, +, <, \{x, y\}\}$

$$|M| = \{a, b, c\}$$

$$M(0) = a$$

$$M(<) = \{\langle a, b \rangle, \langle a, c \rangle, \langle b, c \rangle\}$$

$$\text{if } M(x) = a, M(y) = b$$

$$\text{then } M(+ (x, y)) \text{ is } M(+)(M(x), M(y)) =$$

$$M(+)(a, b) = b$$

$$M(+ (+ (x, y), y)) = c$$

Define formula  $\phi: = \forall x \exists y + (x, y) = x$

$$M \models \forall x \exists y + (x, y) = x$$

$M(+)$	$a$	$b$	$b$
$a$	$a$	$b$	$c$
$b$	$b$	$c$	$a$
$c$	$c$	$a$	$b$

We say that the model  $M$  *T-satisfies* the formula  $\phi$

# Decision procedures

Given a theory  $T$  a **theory solver** or a **decision procedure** for  $T$  takes as input a set of **literals**  $\phi$  (atomic propositions) and determines whether  $\phi$  is  **$T$ -satisfiable**, that is,

$\exists$  a model  $M$  such that  $M \models \phi$ ?

# $\langle \setminus \text{model } \textit{theory} \rangle$

A short overview of theories and models in mathematical logic

# Example theories

- (Real) Linear arithmetic
  - $4x - 3y + 6z \leq 10, x + y - z \leq 1;$
- Real nonlinear arithmetic
  - $4x^2 + 6y - 9z^3 \leq 5$
- Bit vectors
- Arrays
  - $x'[i] = x[i] + 1$
- Uninterpreted functions (UF)  $\Sigma_F := \{f, g, \dots\}, \Sigma_P := \{=\}, V := \{x_i\}$ 
  - $x_1 = x_2 \wedge x_3 \neq x_2 \wedge f(x_3) \neq f(x_2)$
- Difference logic  $\Sigma_F := \{1, 2, 3, \dots, - \}, \Sigma_P := \{<, \leq, =, >, \geq\}, V := \{x_i\}$ 
  - $x_1 - x_2 \geq k$ , where  $\geq \in \{<, \leq, =, >, \geq\}$

# Uninterpreted functions

Useful for abstractly reasoning about programs

- $\Sigma_F := \{f, g, \dots\}$ ,  $\Sigma_P := \{=\}$ ,  $V := \{x_i\}$

Literals are of the form  $x_1 = x_2 \wedge x_3 \neq x_2 \wedge f(x_3) \neq f(x_2)$

We know nothing about  $f, g, \dots$  except for its name and arity

How to check its satisfiability?



# Decision procedure for Uninterpreted functions (UF)

$$\phi = (x_1 = x_2) \wedge (x_2 = x_3) \wedge (x_4 = x_5) \wedge (x_5 \neq x_1) \wedge (F(x_1) \neq F(x_3))$$

## Decision procedure

1. Put all variables and function instances in their own classes
2. If  $t_1 = t_2$  is a literal then merge the classes containing them; do this repeatedly
3. If  $t_1$  and  $t_2$  are terms in the same class then merge classes containing  $F(t_1)$  and  $F(t_2)$ ; repeat
4. If  $t_1 \neq t_2$  is a literal in  $\phi$  and they belong to the same class then return unsat else return sat  $t_1$  and  $t_2$

# Decision procedure for Uninterpreted functions (UF)

Initial classes  $\phi = (x_1 = x_2) \wedge (x_2 = x_3) \wedge (x_4 = x_5) \wedge (x_5 \neq x_1) \wedge (F(x_1) \neq F(x_3))$

Classes  $\{x_1\} \{x_2\} \{x_3\} \{x_4\} \{x_5\} \{F(x_1)\} \{F(x_3)\}$

$\{x_1, x_2, x_3\} \{x_4, x_5\} \{F(x_1)\} \{F(x_3)\}$

$\{x_1, x_2, x_3\} \{x_4, x_5\} \{F(x_1), F(x_3)\}$

*Unsat*

# Difference Logic (conjunctive fragment)

A useful fragment of linear arithmetic

$$\Sigma_F := \{1, 2, \dots, -\}$$

$$\Sigma_P := \{<, \leq, =, \neq, >, \geq\}$$

Literals are of the form  $x_1 - x_2 \cong k$ , where  $\cong \in \{<, \leq, =, >, \geq\}$

$x_1, x_2$  are Integers or rational variables

Example:  $\phi = (x - y = 5) \wedge (z - y \geq 2) \wedge (z - x > 2) \wedge (w - x = 2) \wedge (z - w < 0)$

Satisfiability problem: checking whether this formula is consistent

# An Application: Job shop scheduling problem

Given a finite set of  $n$  jobs. Each job  $i$  of which consists of a chain of operations  $(m_1^i, d_1^i), (m_2^i, d_2^i), \dots$ . There is a finite set of  $m$  machines  $M = \{m_1, m_2, \dots, m_m\}$ , each of which can handle at most one operation at a time.

The problem of finding a shortest schedule---allocation of machine time to jobs---can be formulated in DL.

# Decision procedure for Difference logic

$$\phi = (x - y = 5) \wedge (z - y \geq 2) \wedge (z - x > 2) \wedge (w - x = 2) \wedge (z - w < 0)$$

Decision procedure:

Convert each literal (AF) to  $x_1 - x_2 \leq c$  form:

# Decision procedure for Difference logic

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Convert each literal (AF) to  $x_1 - x_2 \leq c$  form:

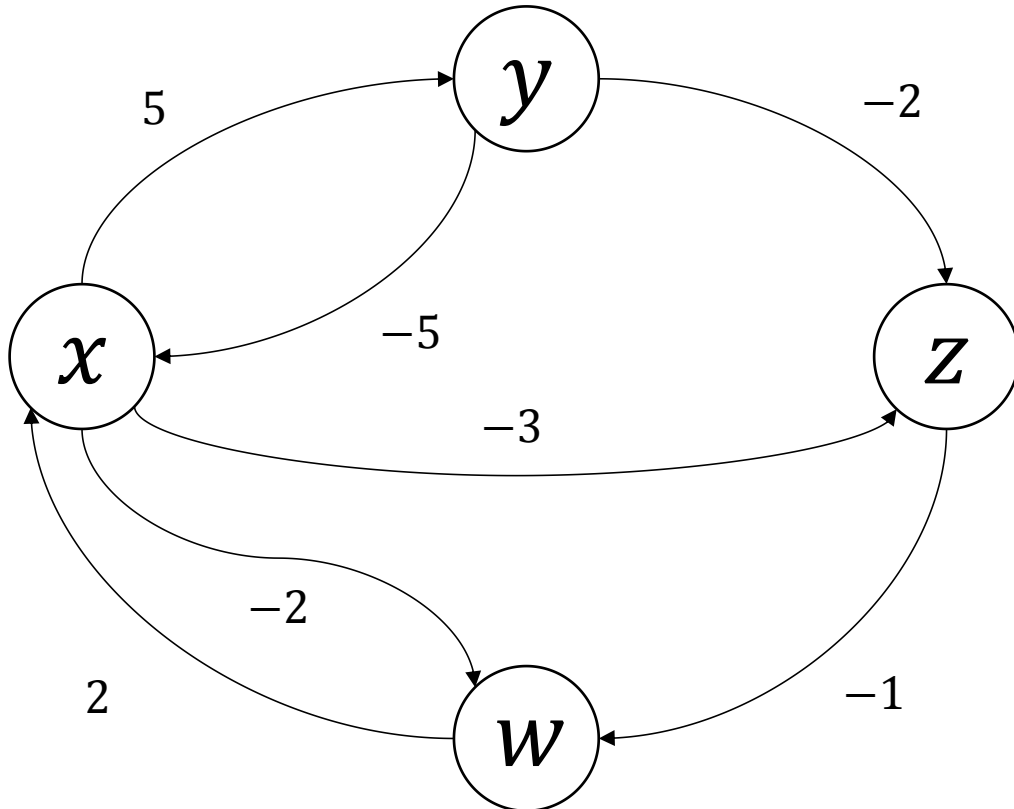
$$\begin{aligned} \phi' = & (x - y \leq 5) \wedge (y - x \leq -5) \\ & \wedge (y - z \leq -2) \wedge \\ & (x - z \leq -3) \wedge \\ & (w - x \leq 2) \wedge (x - w \leq -2) \\ & (z - w \leq -1) \end{aligned}$$

For integer domain  $(x_1 - x_2 < k)$  is replaced by  $(x_1 - x_2 \leq k - 1)$

How to check satisfiability or consistency of formula  $\phi'$ ?

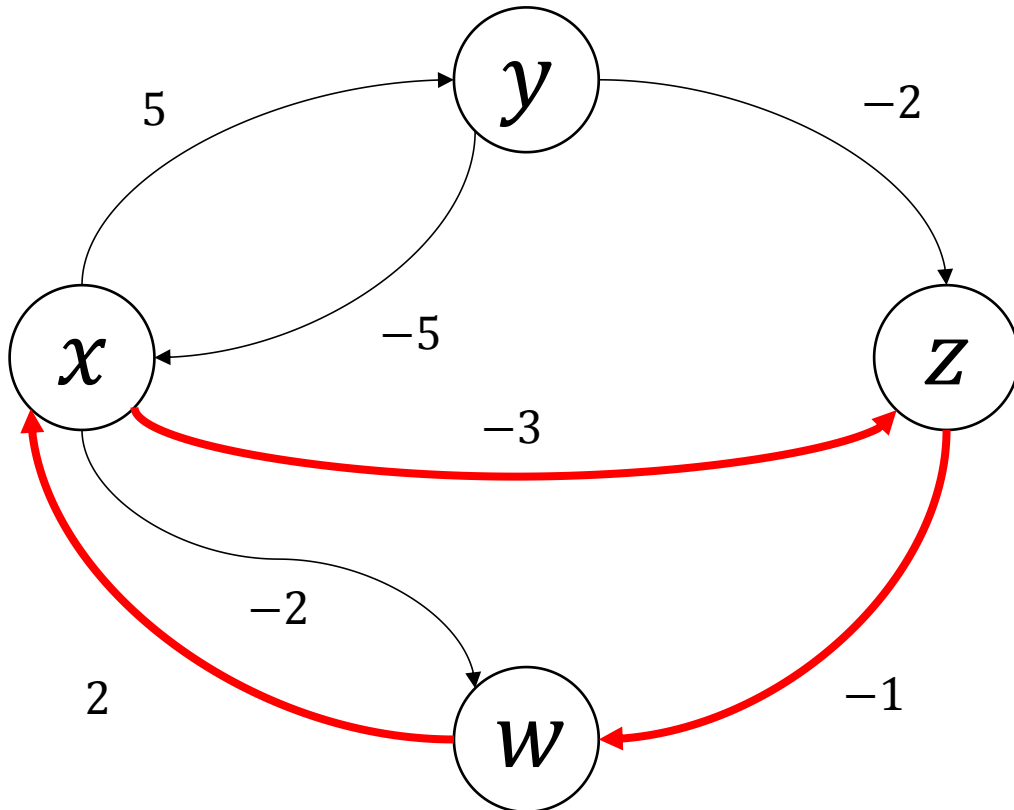
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Construct a graph  $G_{\phi'}$ , with edge from  $x \rightarrow^c y$  for each literal  $x - y \leq c$  in  $\phi'$



$$\begin{aligned} \phi' = & (x - y \leq 5) \wedge (y - x \leq -5) \\ & \wedge (y - z \leq -2) \wedge \\ & (x - z \leq -3) \wedge \\ & (w - x \leq 2) \wedge (x - w \leq -2) \\ & (z - w \leq -1) \end{aligned}$$

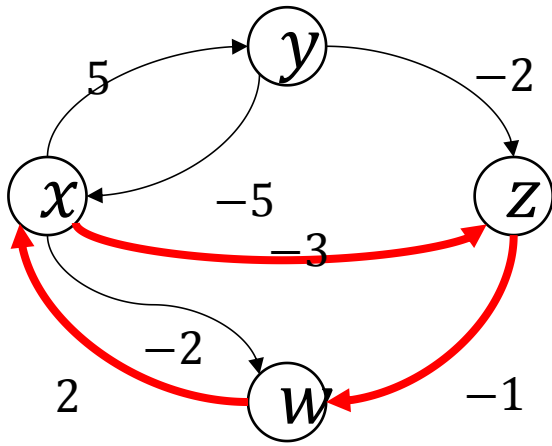
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**Proposition.**  $\phi$  is satisfiable iff  $G_{\phi'}$  is negative cycle free.

Proof. ( $\Leftarrow$ ) If there is a negative cycle then  $(x - z \leq -3)$ ;  $(z - w \leq -1)$ ;  $(w - x \leq 2)$  adding all up:  $(0 \leq -2)$  which is inconsistent.





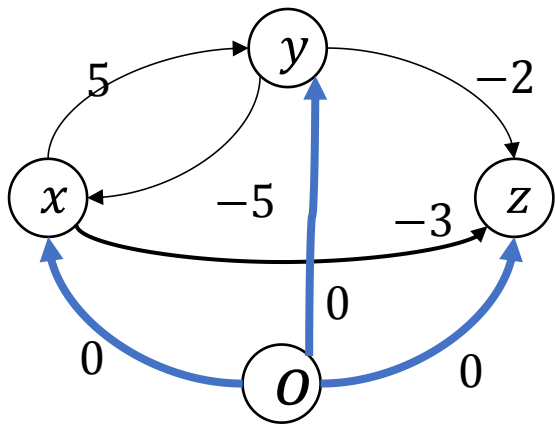
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( $\Rightarrow$ ) Let us assume that there is no negative cycle. We will construct a satisfying solution  $\sigma: V \rightarrow \mathbb{Z}$

Consider additional vertex  $o$  with  $o \xrightarrow{0} v$  edges for all  $v$



For each variable  $v$  define solution as the shortest distance from  $o$  to  $v$  (be aware of negative distances):  $\sigma(v) = -dist(o, v)$

Suppose FSOC,  $\sigma$  does not satisfy a literal  $x - y \leq k$  then

$$-dist(o, x) + dist(o, y) > k$$

$$dist(o, y) > k + dist(o, x)$$

$$dist(o, y) > dist(x, y) + dist(o, x)$$

violates definition of  $dist(o, y)$ !

$$\sigma(x) = -dist(o, x) = 5$$

$$\sigma(y) = -dist(o, y) = 0$$

$$\sigma(z) = -dist(o, z) = 8$$

# Summary of DP for Difference Logic

- Satisfiability check for conjunctive fragment of DL can be performed using Bellman-Ford algorithm in time  $O(|V| \cdot |E|)$
- Inconsistency/unsatisfiability explanations are negative cycles
- Amenable to incremental checks

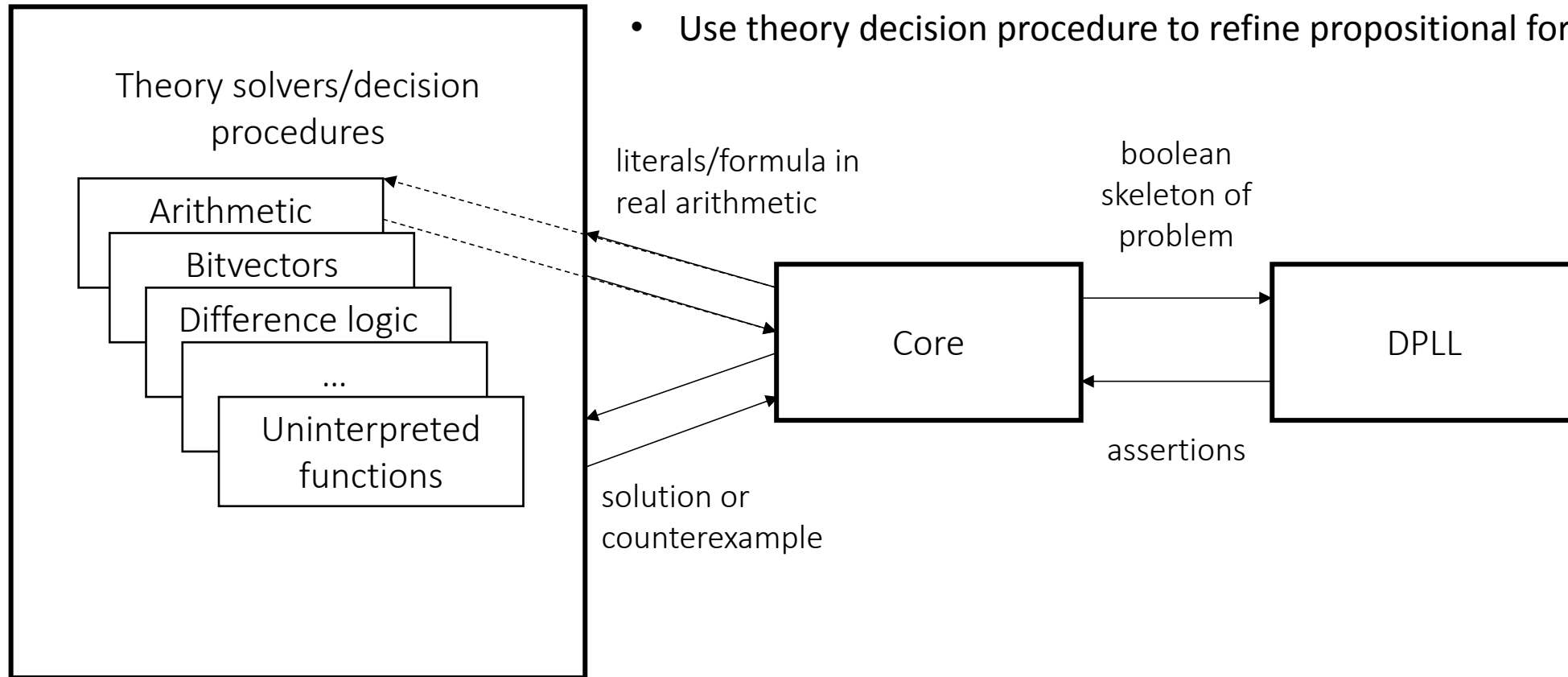
# Return to SMT

$$\phi \equiv (g(a) = c) \wedge (f(g(a)) \neq f(c) \vee g(a) = d) \wedge c \neq d$$

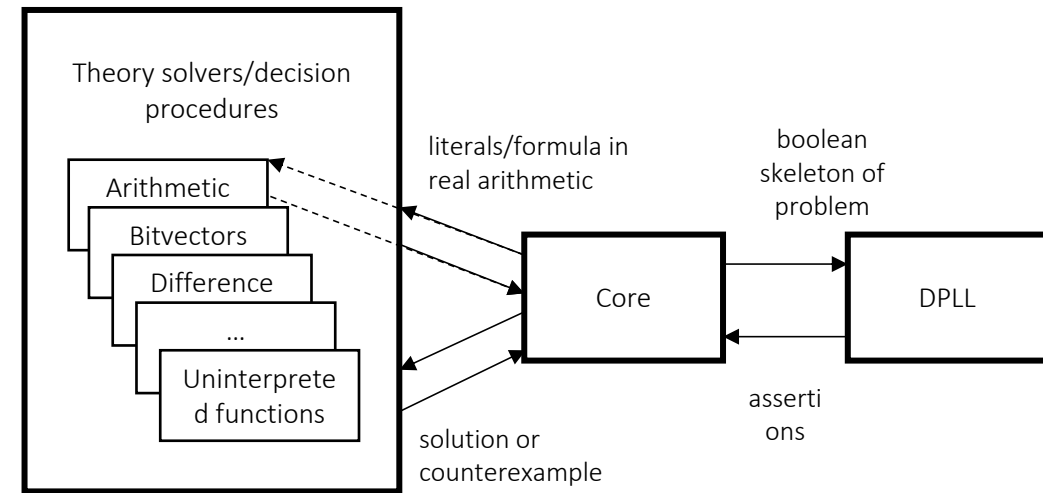
Several approaches, lazy approach:

- Abstract  $\phi$  to propositional form
- Feed to DPLL
- Use theory decision procedure to refine propositional formula a guide SAT

$$\phi \equiv x_1 \wedge (\neg x_2 \vee x_3) \wedge \neg x_4$$



$$\phi \equiv \underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$



- send  $\{1, \bar{2} \vee 3, \bar{4}\}$  to DPLL
- returns model  $\{1, \bar{2}, \bar{4}\}$
- UF solver concretizes to  $g(a) = c, f(g(a)) \neq f(c), c \neq d$
- checks this as UNSAT
- send  $\{1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4\}$  to DPLL
- returns model  $\{1, 2, 3, \bar{4}\}$
- UF solver concretizes and finds this to be UNSAT
- send  $\{1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee \bar{2} \vee \bar{3} \vee 4\}$  to DPLL
- returns UNSAT

# Assignments

- Learn z3
  - <https://ericpony.github.io/z3py-tutorial/guide-examples.htm>

## Readings

- Reading more about decision procedures
- Learn neural networks if you never used them

