

# #16: How SAT/SMT Solvers Work

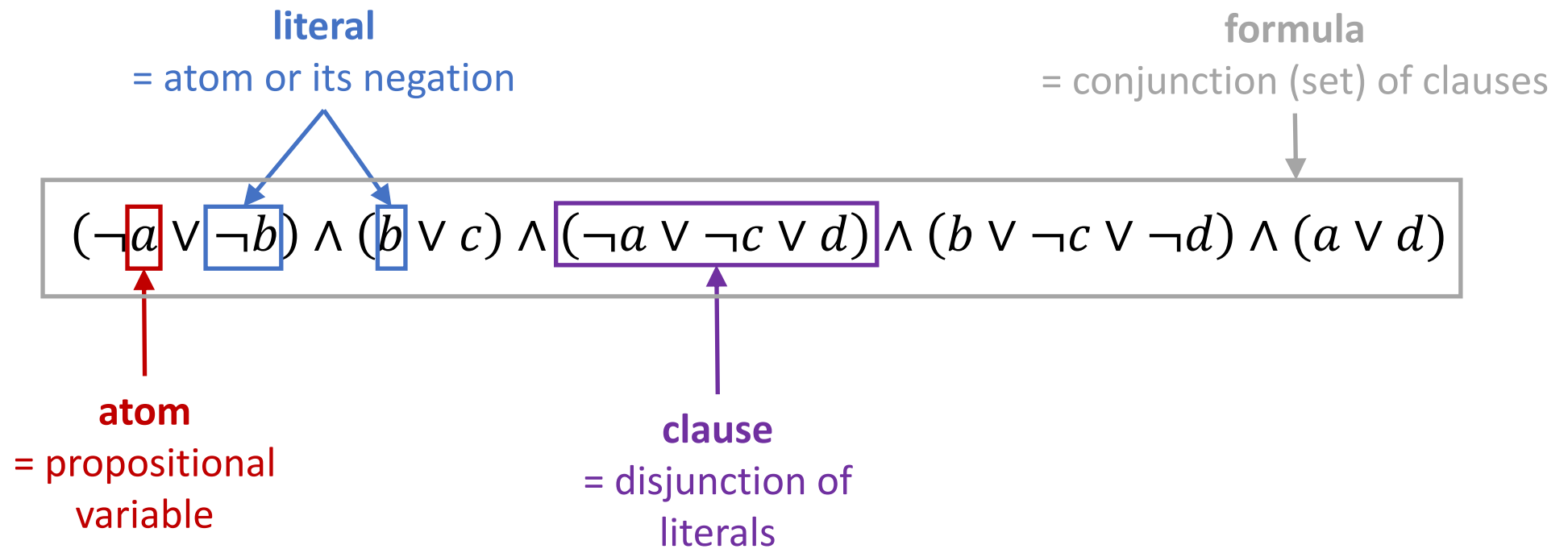
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EECS 700: Introduction to Program Synthesis



# The SAT problem

- **Input:** propositional formula in CNF



# The SAT problem

- **Problem:** find a *satisfying assignment* (also called a *model*)
  - or determine that the formula is *unsatisfiable*

$$(\neg a \vee \neg b) \wedge (b \vee c) \wedge (\neg a \vee \neg c \vee d) \wedge (b \vee \neg c \vee \neg d) \wedge (a \vee d)$$

a satisfying assignment:

$$\{a \mapsto 0, b \mapsto 1, c \mapsto 0, d \mapsto 1\}$$

can be written as a set of literals:

$$\{\neg a, b, \neg c, d\}$$

or as a formula:

$$\neg a \wedge b \wedge \neg c \wedge d$$

# Naive solution

$$(\neg a \vee \neg b) \wedge (b \vee c) \wedge (\neg a \vee \neg c \vee d) \wedge (b \vee \neg c \vee \neg d) \wedge (a \vee d)$$

- Build a truth table!
  - We can't do fundamentally better:  
it's an NP-complete problem
  - But we can do way better in practice  
for common instances

$2^{|P|}$

0000	0
0001	0
0010	0
0011	0
0100	0
0101	1
0110	0
0111	1
...	

# Intuition: Sudoku

- Easy vs hard: what's the difference?

7	9					3		
					6	9		
8				3			7	6
			9	6	5			2
		5	4	1	8	7		
4			7	2	3			
6	1			9				8
		2	3					
		9					5	4

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		9	7	4	8			
7								
	2		1		9			
		7				2	4	
	6	4		1		5	9	
	9	8				3		
			8		3		2	
								6
			2	7	5	9		

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- Most real-world SAT instances allow a lot of inference

# DPLL algorithm


[Davis, Putnam '60]

[Davis, Logemann, Loveland '62]

- **State:** current model  $M$  (a sequence of annotated literals)

$$M = \boxed{a^d} \neg b \ c$$

decision literal



- **Transitions:**

- **decide**  $M \longrightarrow M \ l^d$  if  $l$  undefined in  $M$
- **unit-propagate**  $M \longrightarrow M \ l$  if there is a clause where all literals are false except  $l$ , which is undefined
- **backtrack**  $M \ l^d \ M' \longrightarrow M \ \neg l$  if there is a conflicting clause and  $M'$  has no decision literals
- **fail**  $M \longrightarrow Unsat$  if there is a conflicting clause and no decision literals

# DPLL: example

$$(\neg a \vee \neg b) \wedge (b \vee c) \wedge (\neg a \vee \neg c \vee d) \wedge (b \vee \neg c \vee \neg d) \wedge (a \vee d)$$

$M =$	$\emptyset$	decide
	$a^d$	unit-propagate
	$a^d \neg b$	unit-propagate
	$a^d \neg b c$	unit-propagate
	$a^d \neg b c d$	backtrack
	$\neg a$	unit-propagate
	$\neg a d$	decide
	$\neg a d \neg c^d$	unit-propagate
	$\neg a d \neg c^d b$	SAT!

# DPLL + clause learning

$$(\neg a \vee b) \wedge (\neg c \vee d) \wedge (\neg e \vee \neg f) \wedge (f \vee \neg b \vee \neg e) \wedge (\neg a \vee \neg e)$$

$M =$

$\emptyset$

$a^d$

$a^d b$

$a^d b c^d$

$a^d b c^d d$

$a^d b c^d d e^d$

$a^d b c^d d e^d \neg f$

$a^d b c^d d \neg e$

Bad decision!

decide

unit-propagate

decide

unit-propagate

decide

unit-propagate

backtrack

Wait, but why?



# DPLL + clause learning

$$(\neg a \vee b) \wedge (\neg c \vee d) \wedge (\neg e \vee \neg f) \wedge (f \vee \neg b \vee \neg e) \wedge (\neg a \vee \neg e)$$

$M =$	$\emptyset$	decide
	$a^d$	unit-propagate
	$a^d b$	decide
	$a^d b c^d$	unit-propagate
	$a^d b c^d d$	decide
	$a^d b c^d d e^d$	unit-propagate
	$a^d b c^d d e^d \neg f$	<b>backjump</b>
	$a^d b \neg e$	

# This lecture

1. Demo: how to use Z3 to
  - solve constraints
  - verify programs
  - synthesize programs
2. How do SAT solvers work?
3. How do SMT solvers work?



# Beyond propositional logic

- What if our formula looks like this?

$$(p \wedge \neg q \vee a = f(b - c)) \wedge (g(g(b)) \neq c \vee a - c \leq 7)$$

- talks about integers, functions, sets, lists...
- One idea: bit-blast everything and use SAT
  - can only find solutions within bounds
  - very inefficient, so bounds are small
- Better idea: combine SAT with special **solvers** for **theories**
  - they “natively understand” integers, functions, etc.

# First-order theories

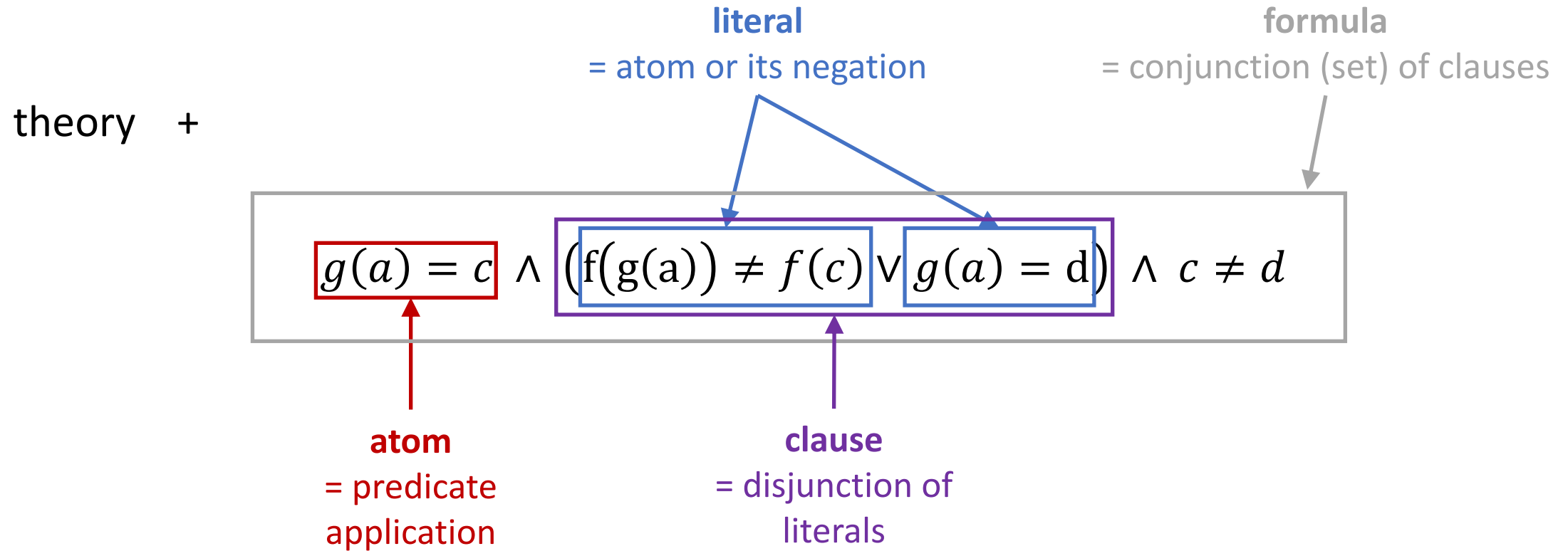
- theory = <function symbols, predicate symbols, axioms>

ground first-order formulas over  
functions and predicates

- Example:** theory of Equality and Uninterpreted Functions

- EUUF = <{f, g, h, ...}, {=}, { $\forall x. x = x$   
 $\forall x y. x = y \Rightarrow y = x$   
 $\forall x y z. x = y \wedge y = z \Rightarrow x = z$   
 $\forall x y. x = y \Rightarrow f(x) = f(y)$   
}>

# The SMT problem



# Theories for our purpose

- theory = <function symbols, predicate symbols, ~~axioms~~>

can decide consistency of  
conjunctions of literals

→ solve  
r

$$f(a) = c$$

$$f(b) \neq d$$

$$c = d$$

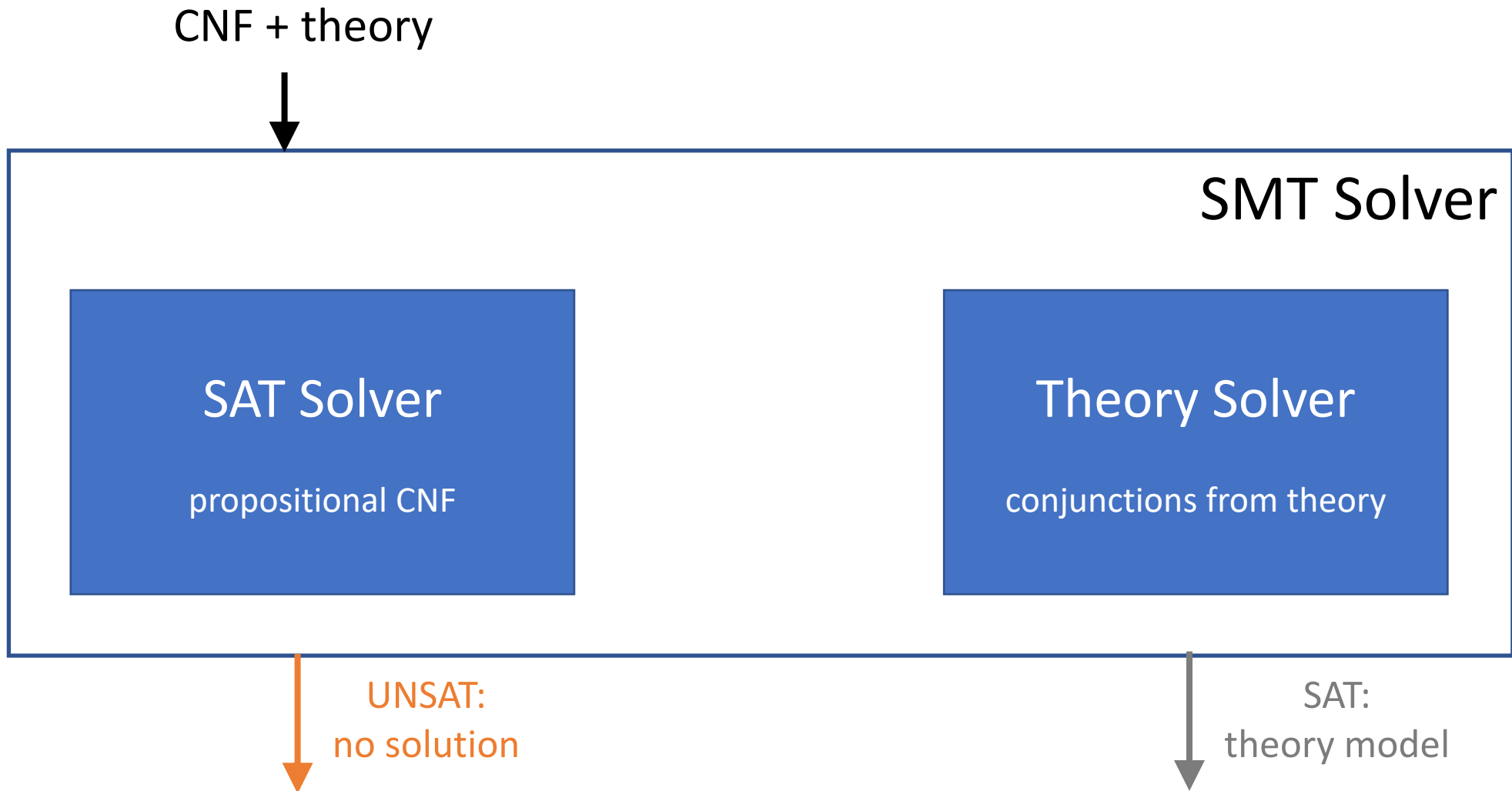
$$a = b$$

EUF solver



Inconsistent!

# DPLL(T) architecture



# Basic DPLL(T)

$$\boxed{g(a) = c} \wedge (\boxed{f(g(a)) \neq f(c)} \vee \boxed{g(a) = d}) \wedge \boxed{c \neq d}$$

abstract atoms to  
propositional variables

$$p \wedge (\neg q \vee r) \wedge \neg s$$

$$p \wedge (\neg q \vee r) \wedge \neg s \xrightarrow{\text{SAT solver}} p \neg q \neg s$$

Inconsistent!

$$\xleftarrow{\text{EUF solver}} \boxed{g(a) = c} \quad \boxed{f(g(a)) \neq f(c)} \quad c \neq d$$

$$p \wedge (\neg q \vee r) \wedge \neg s \wedge (\neg p \vee q) \xrightarrow{\text{SAT solver}} p q r \neg s$$

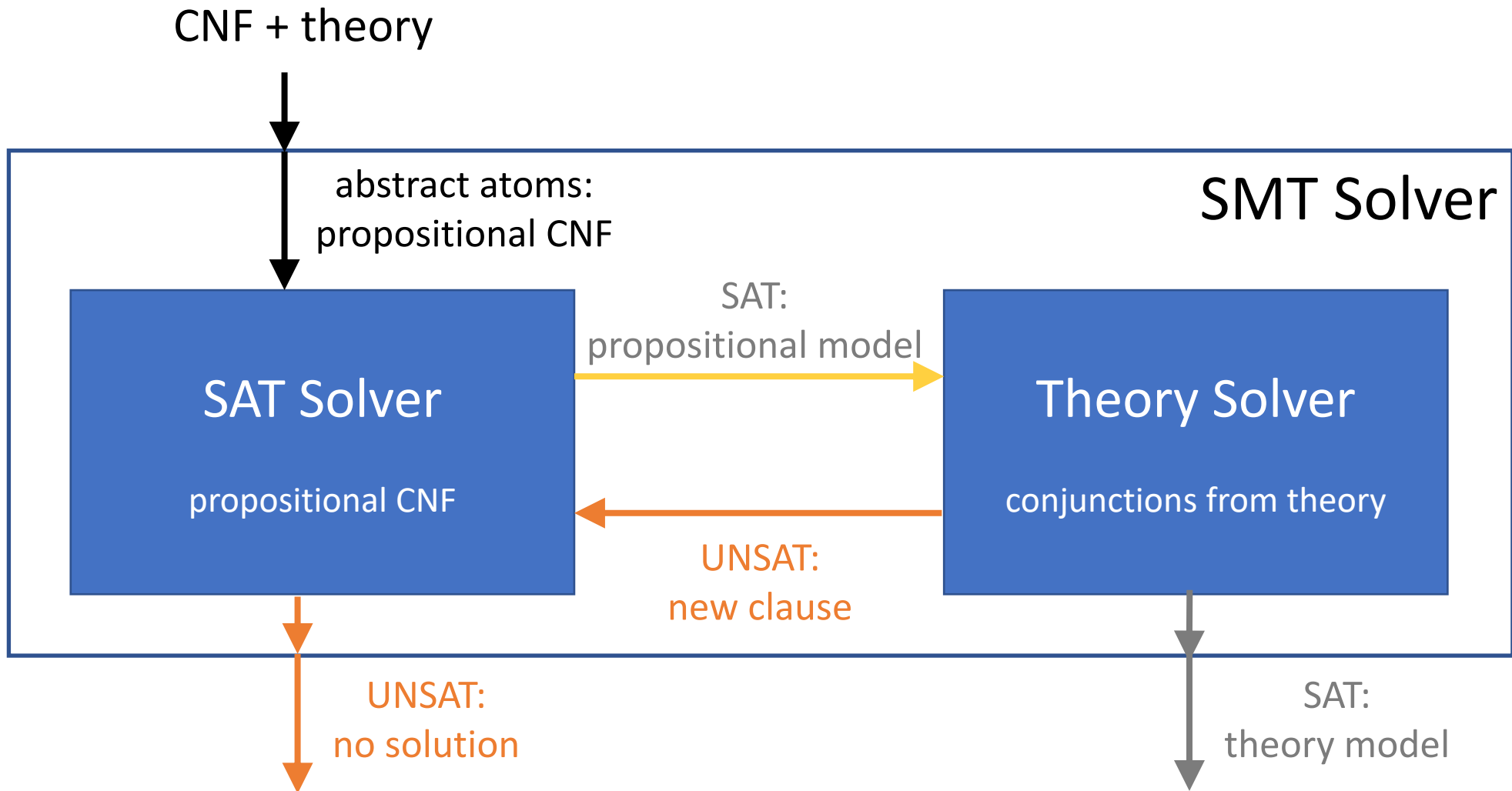
Inconsistent!

$$\xleftarrow{\text{EUF solver}} \boxed{g(a) = c} \quad f(g(a)) = f(c) \quad \boxed{g(a) = d} \quad \boxed{c \neq d}$$

$$p \wedge (\neg q \vee r) \wedge \neg s \wedge (\neg p \vee q) \wedge (\neg p \wedge \neg r \wedge s) \xrightarrow{\text{SAT solver}} \text{Unsat}$$



# DPLL(T) architecture



# Popular theories

- Equality and Uninterpreted Functions
- EUF =  $\langle \{\mathbf{f}, \mathbf{g}, \mathbf{h}, \dots\}, \{=\}, \text{axioms of equality \& congruence} \rangle$
- Linear Integer Arithmetic
- LIA =  $\langle \{\mathbf{0}, \mathbf{1}, \dots, +, -\}, \{=, \leq\}, \text{axioms of arithmetic} \rangle$
- Arrays

Arrays =  $\langle \{\mathbf{sel}, \mathbf{store}\}, \{=\}, \forall a \ i \ v. \mathbf{sel}(\mathbf{store}(a, i, v), i) = v$   
 $\forall a \ i \ j \ v. i \neq j \Rightarrow \mathbf{sel}(\mathbf{store}(a, i, v), j) = \mathbf{sel}(a, j) \ \rangle$

Theories can be combined!

Nelson-Oppen combination

# Why do we care?

- If we can encode a synthesis problem as SAT/SMT, we can use solvers to do the search for us
- Get some inspiration from how solvers search
  - Unit propagation similar to top-down propagation (pruning through inference of consequences of a guess)
  - Backjumping / clause learning?
    - Feng, Martins, Bastani, Dillig: [Program synthesis using conflict-driven learning](#). PLDI'18
  - Coarse-grained reasoning and gradual refinement like in DPLL(T)?
    - Wang, Dillig, Singh: [Program synthesis using abstraction refinement](#). POPL'18