COL:750

Foundations of Automatic Verification

Instructor: Priyanka Golia

Course Webpage

https://priyanka-golia.github.io/teaching/COL-750/index.html



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A theory T is a set of sentences closed under implications

Theory = Subject Knowledge + FOL

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No, it is unsatisfiable, $\nvDash T_{\mathbb{N}} \cup F$

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Formulas in different theories \longrightarrow SMT (Linear integer arithmetic, Linear real arithmetic, bit vectors, strings)

If formula is satisfiable, gives an satisfying assignment





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- Order of magnitude faster than previous SAT solvers
- Many real-world problems don't exhibit worst case theoretical performance

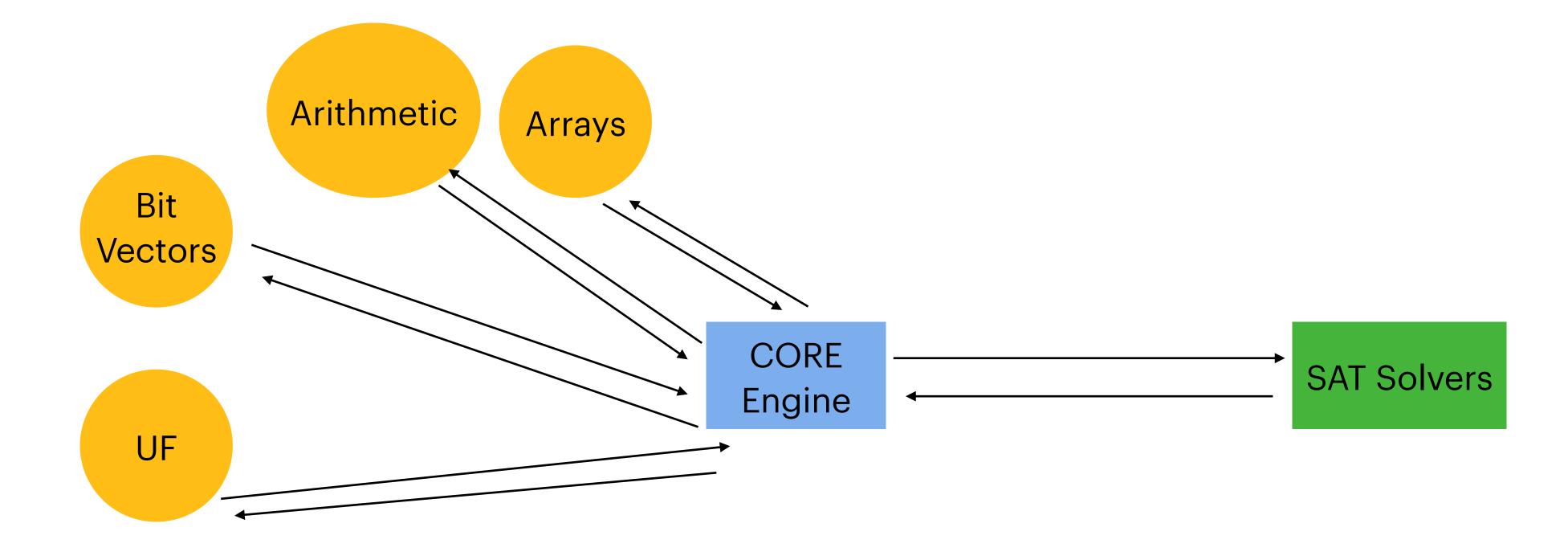
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decidable first-order theories.

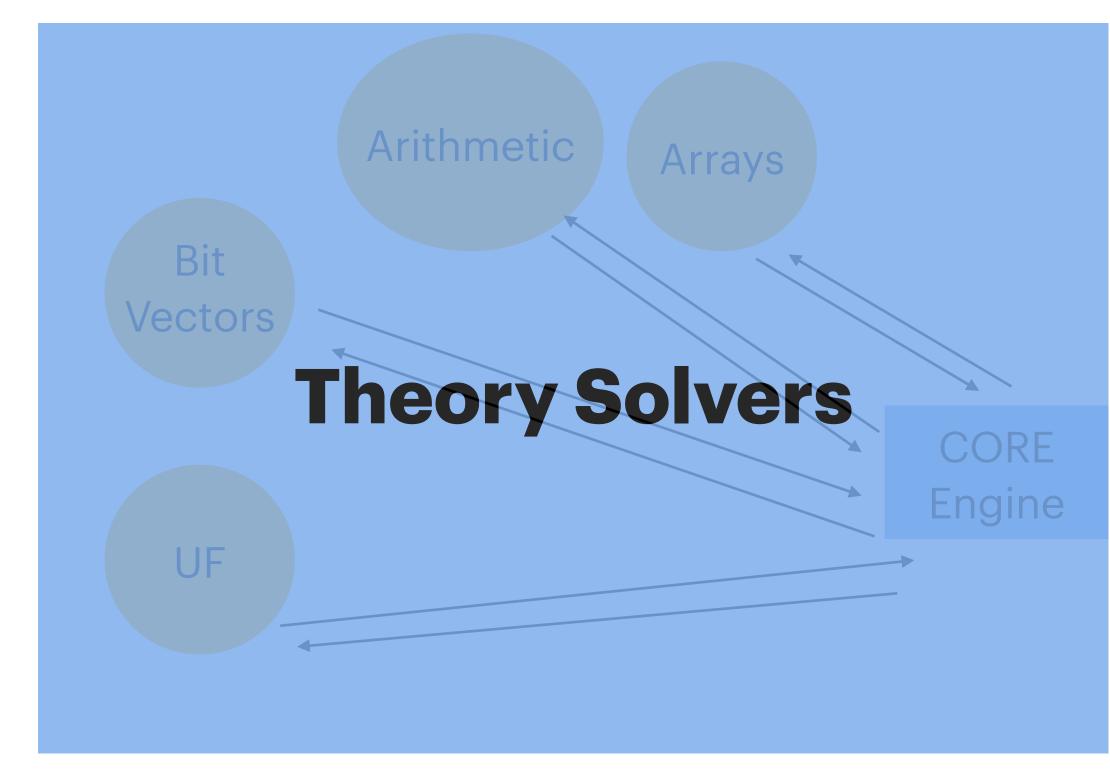
SVC, CVC, Yices solver came to picture — first SMT solver was born!!!

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Alto, 2001, came up with idea of combining SAT solvers with decision procedures for



SMT solvers





SMT solvers

Theory Solvers

Theory Solver: Difference Logic

Difference logic — the satisfiability of a conjunction of arithmetic atoms.

Each atom is of the form $x - y \oplus c$, where x and y are variables, c is a numeric constant, and

The variables can range over either the integers (QF_IDL) or the reals (QF_RDL).

 $\bigoplus \in \{ <, >, \le, \ge, = \}$

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 $x - y > c \quad \equiv y - x < -c$



- A conjunction of literals, all of the form $x y \le c$.
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$$(x - y = 5) \land (z - y \ge 2) \land (z - x > 2) \land (w - x = 2) \land (z - w < 0)$$

Theory Solvers

Linear Arithmetic Solver

Handles inequalities and equalities over integers or real numbers: Techniques: Fourier-Motzkin elimination, Simplex algorithm. Check if $(x + 2y \le 10) \land (x - y \ge 3)$?

Bit-Vector Solver

Deals with fixed-width integers and bitwise operations:

reasoning

Check if x > 4 = 0x0A

- Techniques: Bit-blasting (reducing bit-vector problems to SAT), word-level

Theory Solvers

Theory Propagation

Deducing new constraints or facts based on existing ones. For example, in linear integer arithmetic:

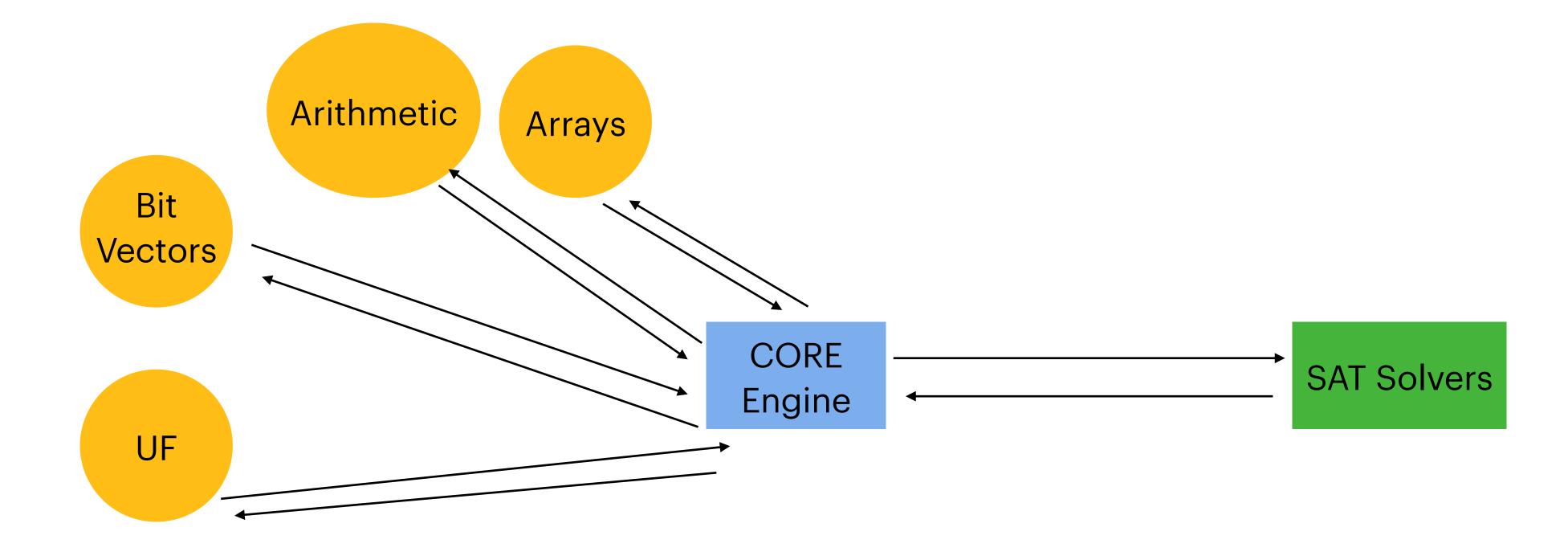
given $(x \ge 5) \land (y = x + 2)$, we can deduce y > = 7

Theory Consistency Checking

Check if a set of constraints is consistent within the theory.

If not, it provides a conflict (a minimal subset of constraints that are unsatisfiable)





Two main approaches:

- "Eager" approach 1.
 - Translate into an equisatisfiable propositional formula 1. 2. Feed it to any SAT solver
- 2. "Lazy" approach
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Cvc5, z3, MathSAT, OpenSMT

 $(g(a) = c) \land (f(g(a)) \neq f(a))$

$$(c) \lor g(a) = d) \land (c \neq d)$$

$(g(a) = c) \land (f(g(a)) \neq f(c) \lor g(a) = d) \land (c \neq d)$

 p_1



$$p_3 \qquad \neg p_4$$

$(g(a) = c) \land (f(g(a)) \neq f(a))$



Send $(p_1 \land (\neg p_2 \lor p_3) \land \neg p_4)$ to a SAT solver. SAT solver returns $\sigma = \{p_1 \mapsto 1, p_2 \mapsto 0, p_3 \mapsto 0, p_4 \mapsto 0\}$

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Theory: Equality with Uninterpreted Functions

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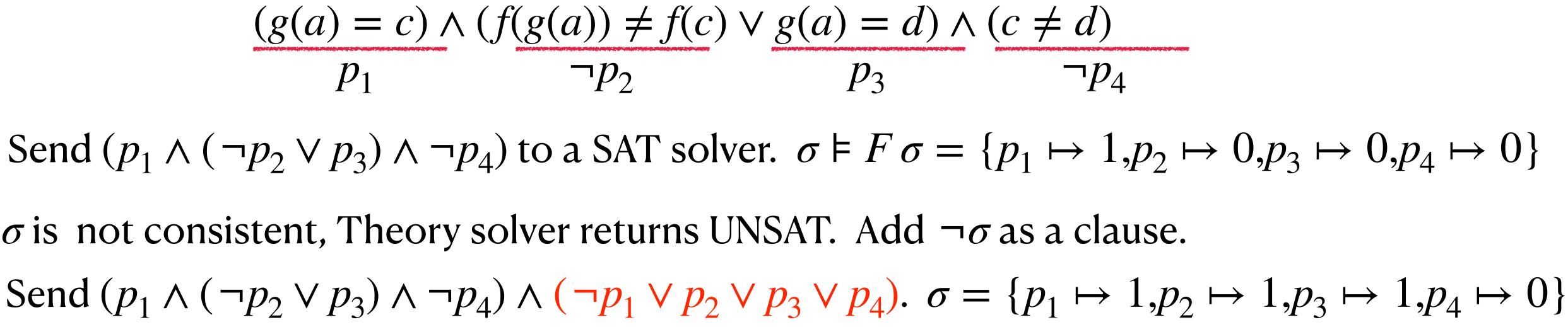
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$$p_3 \mapsto 0, p_4 \mapsto 0$$

SMT solving — Lazy Approach $(g(a) = c) \land (f(g(a)) \neq f(c) \lor g(a) = d) \land (c \neq d)$

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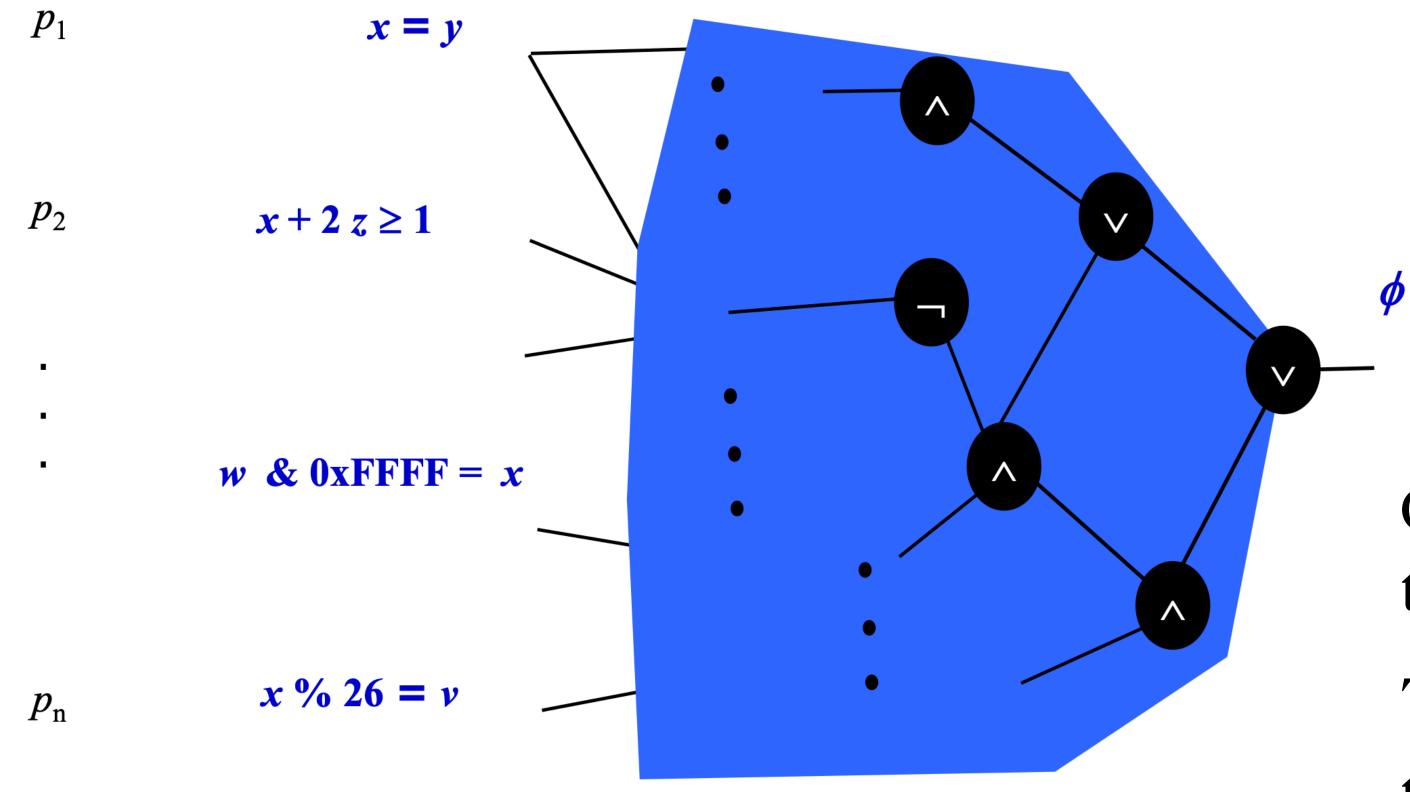
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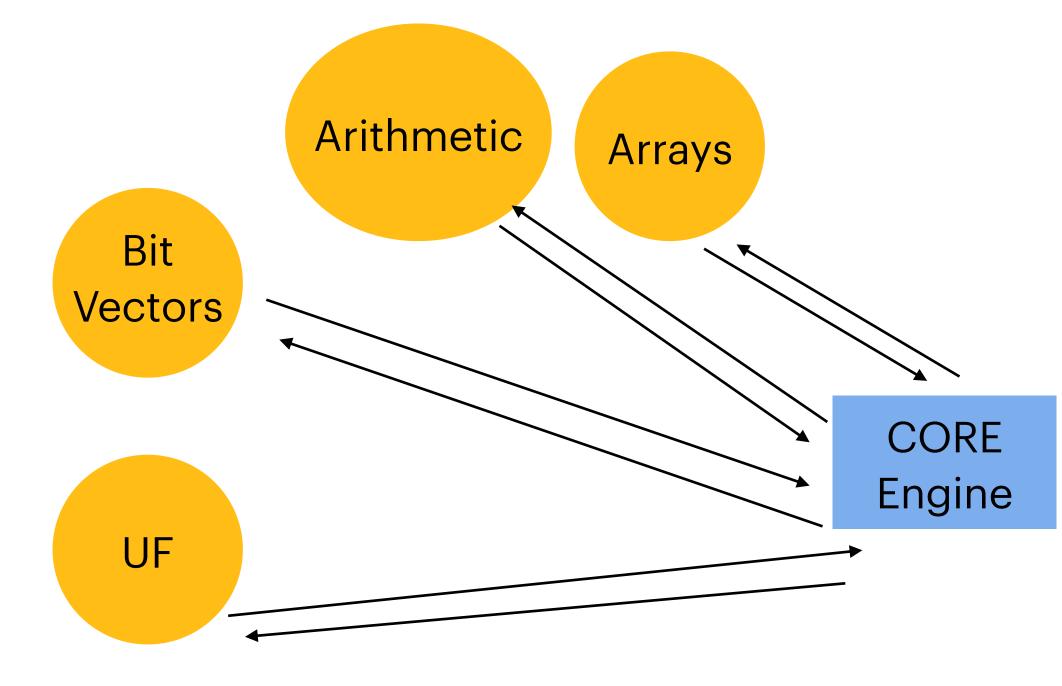
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- Backtrack to a point where M was still T-Satisfiable, use this to pass more explanation to SAT solver.

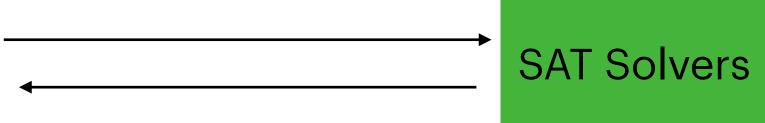


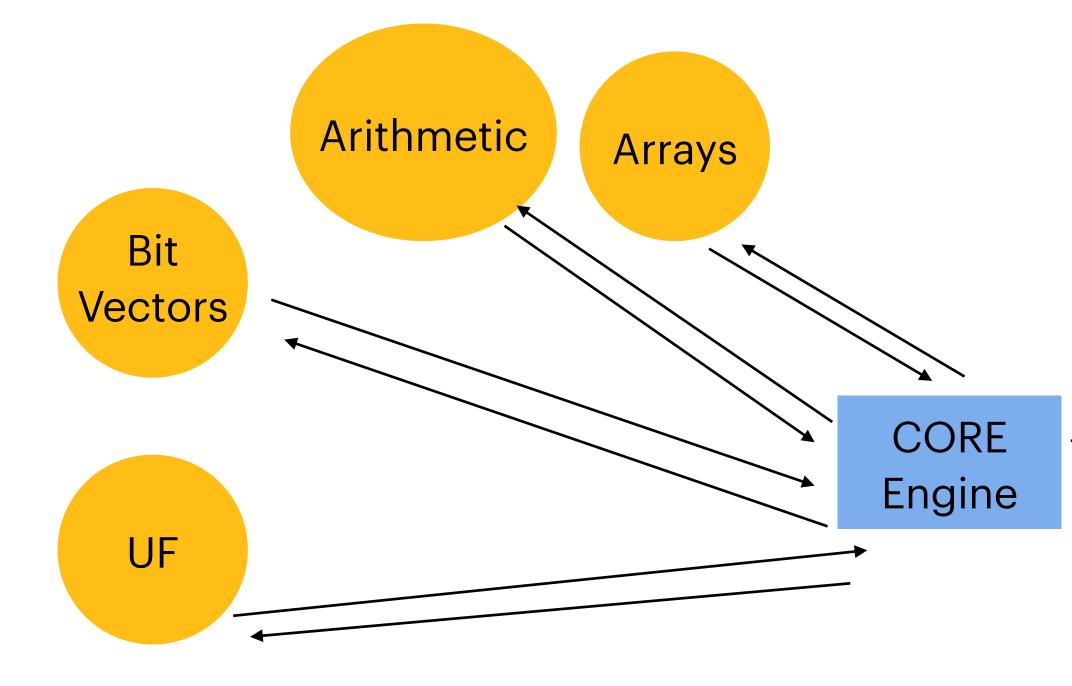
Can have combinations of theories!

Task is to find an assignment to $Vars(\phi)$ such that ϕ is satisfiable!







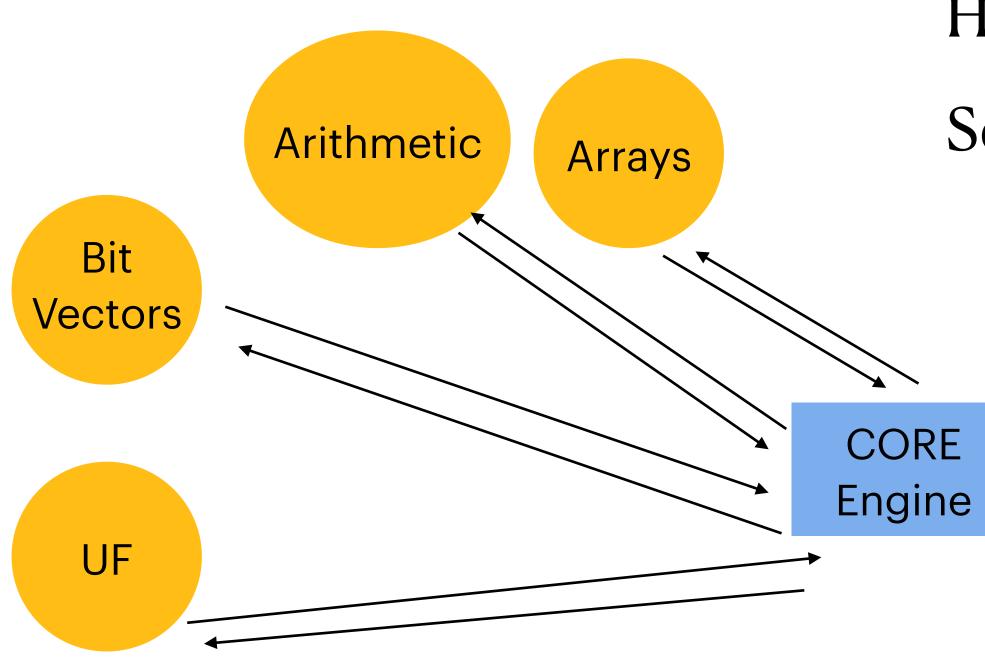




Only sees Boolean Skeleton of the problem!

Builds partial model by assigning truth values to literals

Sends these literals to the core as assertions



Sends each assertions to the appropriate theory

Handles theory combinations

Sends deduced literals to other theories/SAT solver

SAT Solvers

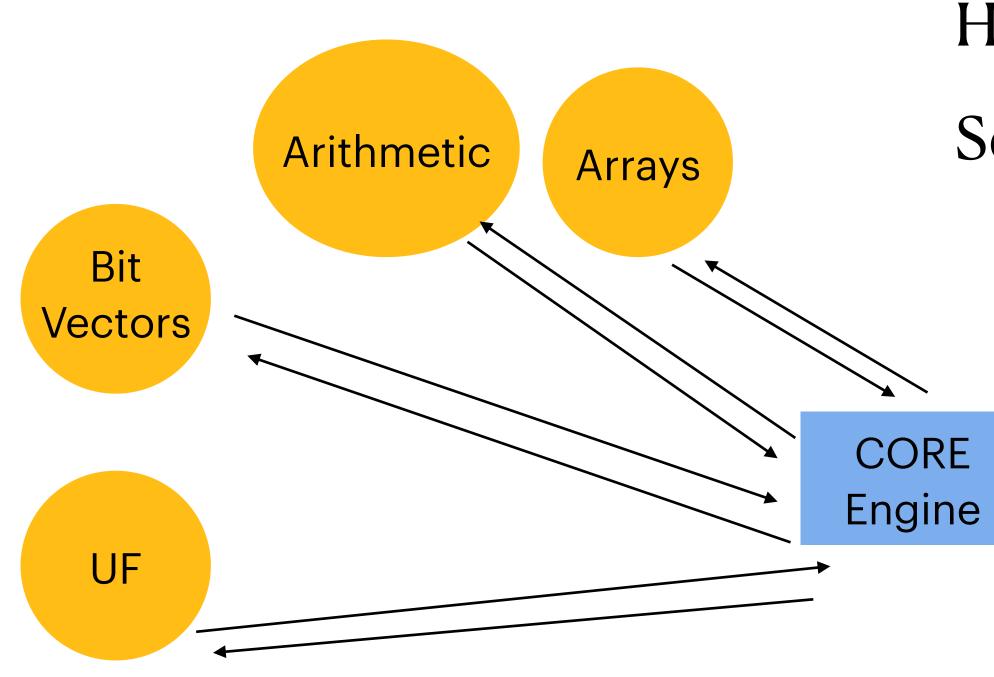
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Theory Solvers! Decide T-satisfiability of a conjunction of literals.

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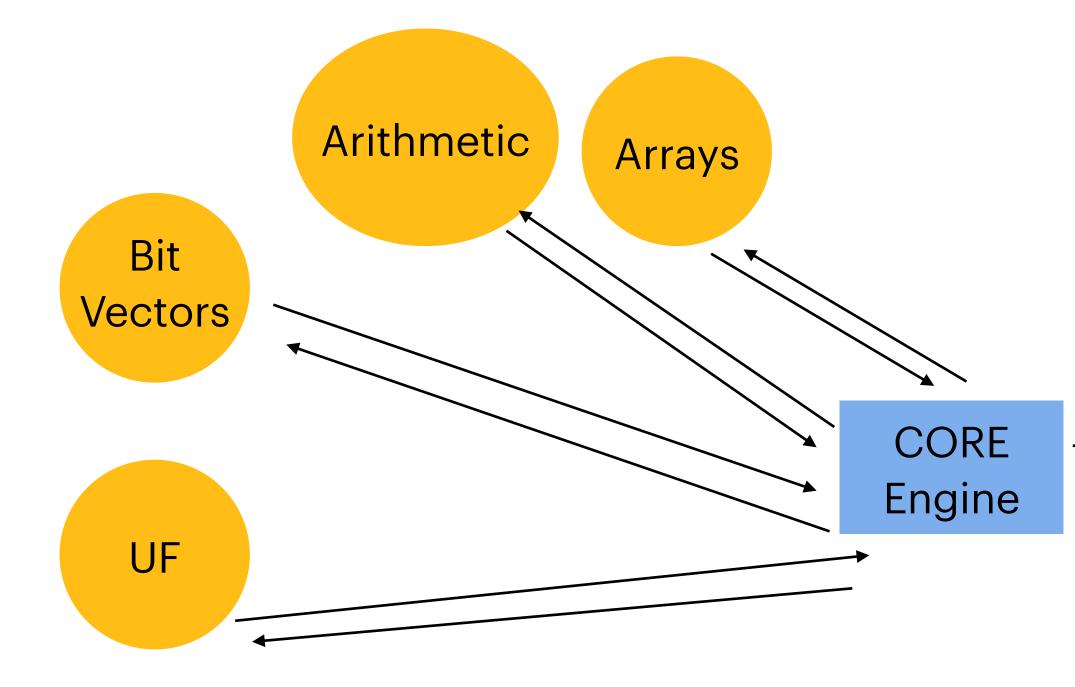
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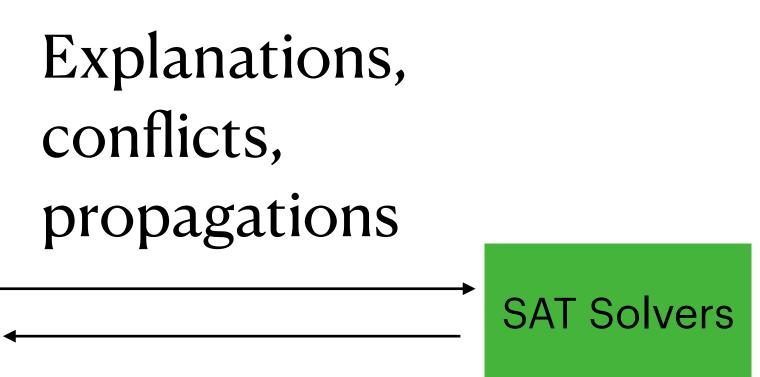
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Assertions

SMT solvers

From SAT & SMT to Temporal Logic

SAT: Checks whether a propositional formula is satisfiable. SMT: Extends SAT with richer theories (e.g., arithmetic, arrays). But What About Time?

SAT/SMT/FOL verify properties in static systems. Many real-world systems evolve over time (e.g., software, robots, protocols).

Can we express this in SAT or FOL?

- "A robot should always eventually return to its charging station." "A user who enters a correct password will eventually get access." "How can we verify that a system never reaches an error state?"

From SAT & SMT to Temporal Logic

Classical logic (SAT/SMT) = Static Reasoning

Temporal logic = Reasoning over time

Linear Temporal Logic (LTL) Assumes a single timeline (one possible sequence of events).

Next Class: Linear Temporal Logic (LTL)

Course Webpage



Thanks!