

Formal Methods in Software Engineering

Satisfiability Modulo Theories, Spring 2026

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Many-Sorted First-Order Logic

Why Many-Sorted?

In standard (mono-sorted) FOL, *all* variables range over a *single* domain. This is too restrictive for practical reasoning:

- **Programming languages** have *typed* variables: `int x`, `bool b`, `double[] arr`.
- **Databases** distinguish between strings, integers, dates, *etc.*
- **Mathematics** itself uses distinct sorts: \mathbb{N} , \mathbb{R} , matrices, functions, *etc.*

Key insight: *Many-sorted FOL* is a natural generalization of FOL where each variable, constant, and function is associated with a *sort* (type). This is exactly what SMT solvers use internally, and what SMT-LIB expresses.

Note: We have already seen first-order logic in the previous lecture. Here, we generalize it to the *many-sorted* setting, which is the standard framework for SMT.

From Single-Sorted to Many-Sorted FOL

In the previous lectures, we studied *single-sorted* (mono-sorted) FOL:

Single-sorted FOL (Weeks 4–5):

- One universal domain D
- All variables range over D
- $\forall x. P(x)$ means “for all x in D ”
- Sufficient for pure math foundations

Many-sorted FOL (this lecture):

- Multiple domains: $\text{Int}^J, \text{Bool}^J, \text{Array}^J, \dots$
- Variables are *typed*: $x : \text{Int}$
- $\forall x : \sigma. P(x)$ — “for all x in σ^J ”
- Essential for programming & verification

The key extension: many-sorted FOL adds a *type system* to FOL. This is exactly what programming languages do — and why SMT solvers speak many-sorted logic natively.

Many-Sorted Signatures

Definition 1: A *many-sorted signature* $\Sigma = \langle \Sigma^S, \Sigma^F \rangle$ consists of:

- Σ^S — a set of *sorts* (also called *types*), e.g. `Bool`, `Int`, `Real`, `Array`.
- Σ^F — a set of *function symbols*, e.g. `=`, `+`, `<`, `read`, `write`.

Definition 2: Each *function symbol* $f \in \Sigma^F$ has a *rank* — an $(n + 1)$ -tuple of sorts:

$$\text{rank}(f) = \langle \sigma_1, \dots, \sigma_n, \sigma_{n+1} \rangle$$

Intuitively, f takes n arguments of sorts $\sigma_1, \dots, \sigma_n$ and returns a value of sort σ_{n+1} .

- Functions of arity 0 are called *constants*, with $\text{rank}(f) = \langle \sigma \rangle$.
- Functions returning `Bool` are called *predicates*.

Note: Every signature includes a distinguished sort `Bool` with constants \top and \perp , and an *equality predicate* \doteq_σ with $\text{rank}(\doteq_\sigma) = \langle \sigma, \sigma, \text{Bool} \rangle$ for each sort σ .

Signature Examples

Number Theory:

- $\Sigma^S = \{\text{Nat}, \text{Bool}\}$
- $\Sigma^F = \{0, S, <, +, \times, \dot{=}_{\text{Nat}}, \dots\}$
- $\text{rank}(0) = \langle \text{Nat} \rangle$
- $\text{rank}(S) = \langle \text{Nat}, \text{Nat} \rangle$
- $\text{rank}(<) = \langle \text{Nat}, \text{Nat}, \text{Bool} \rangle$
- $\text{rank}(+) = \langle \text{Nat}, \text{Nat}, \text{Nat} \rangle$

Arrays:

- $\Sigma^S = \{\text{Array}, \text{I}, \text{E}, \text{Bool}\}$
- $\Sigma^F = \{\text{read}, \text{write}, \dot{=}_{\text{Array}}, \dots\}$
- $\text{rank}(\text{read}) = \langle \text{Array}, \text{I}, \text{E} \rangle$
- $\text{rank}(\text{write}) = \langle \text{Array}, \text{I}, \text{E}, \text{Array} \rangle$

Example: *Propositional logic* can be viewed as a *one-sorted* theory:

- $\Sigma^S = \{\text{Bool}\}$, $\Sigma^F = \{\neg, \wedge, \vee, p_1, p_2, \dots\}$
- $\text{rank}(p_i) = \langle \text{Bool} \rangle$ (propositional variables are Boolean constants)
- $\text{rank}(\neg) = \langle \text{Bool}, \text{Bool} \rangle$, $\text{rank}(\wedge) = \langle \text{Bool}, \text{Bool}, \text{Bool} \rangle$

Terms and Formulas

A *well-sorted term* of sort σ is built from variables, constants, and function applications that *respect* the rank:

- A variable x of sort σ is a term of sort σ .
- A constant c with $\text{rank}(c) = \langle \sigma \rangle$ is a term of sort σ .
- If f has $\text{rank}(f) = \langle \sigma_1, \dots, \sigma_n, \sigma \rangle$ and t_1, \dots, t_n are terms of sorts $\sigma_1, \dots, \sigma_n$, then $f(t_1, \dots, t_n)$ is a term of sort σ .

Example: In the Number Theory signature with $x : \text{Nat}$:

- $S(0)$ is a well-sorted term of sort Nat ✓
- $S(x) + 0$ is a well-sorted term of sort Nat ✓
- $S(x < 0)$ is *not* well-sorted: $<$ returns Bool , but S expects Nat ✗

A Σ -*formula* is built from *atoms* (terms of sort Bool) using the standard logical connectives ($\neg, \wedge, \vee, \rightarrow, \iff$) and quantifiers ($\forall x : \sigma. \varphi, \exists x : \sigma. \varphi$).

Many-Sorted Interpretations

Definition 3: A *many-sorted interpretation* \mathcal{I} of a signature $\Sigma = \langle \Sigma^S, \Sigma^F \rangle$ assigns:

- To each sort $\sigma \in \Sigma^S$, a non-empty *domain* $\sigma^{\mathcal{I}}$ (e.g. $\text{Int}^{\mathcal{I}} = \mathbb{Z}$, $\text{Bool}^{\mathcal{I}} = \{\text{true}, \text{false}\}$).
- To each function symbol f with $\text{rank}(f) = \langle \sigma_1, \dots, \sigma_n, \sigma \rangle$, a function $f^{\mathcal{I}} : \sigma_1^{\mathcal{I}} \times \dots \times \sigma_n^{\mathcal{I}} \rightarrow \sigma^{\mathcal{I}}$.
- To each variable x of sort σ , a value $x^{\mathcal{I}} \in \sigma^{\mathcal{I}}$.

Example: For the Number Theory signature, one interpretation is:

- $\text{Nat}^{\mathcal{I}} = \mathbb{N}$, $0^{\mathcal{I}} = 0$, $S^{\mathcal{I}}(n) = n + 1$, $+^{\mathcal{I}}(m, n) = m + n$
- $<^{\mathcal{I}}(m, n) = \text{true}$ iff $m < n$ in the usual sense

This is the *intended* (standard) interpretation of natural numbers.

Mono-sorted vs. many-sorted: The only difference is that each variable, constant, and function is now tagged with a sort. Connective semantics (\neg , \wedge , \vee , \forall , \exists) remain exactly the same. Quantifiers now range over the domain of a *specific sort*: $(\forall x : \sigma. \varphi)$ means “for all x in $\sigma^{\mathcal{I}}$ ”.

First-Order Theories

Motivation

Consider the signature $\Sigma = \langle \Sigma^S, \Sigma^F \rangle$ for a fragment of number theory:

- $\Sigma^S = \{\text{Nat}\}$, $\Sigma^F = \{0, 1, +, <\}$
- $\text{rank}(0) = \text{rank}(1) = \langle \text{Nat} \rangle$
- $\text{rank}(+) = \langle \text{Nat}, \text{Nat}, \text{Nat} \rangle$
- $\text{rank}(<) = \langle \text{Nat}, \text{Nat}, \text{Bool} \rangle$

1. Consider the Σ -sentence: $\forall x : \text{Nat}. \neg(x < x)$

- Is it *valid*, that is, true under *all* interpretations?
- No, e.g., if we interpret $<$ as *equals* or *divides*.

2. Consider the Σ -sentence: $\neg \exists x : \text{Nat}. (x < 0)$

- Is it *valid*?
- No, e.g., if we interpret Nat as the set of *all* integers.

3. Consider the Σ -sentence: $\forall x : \text{Nat}. \forall y : \text{Nat}. \forall z : \text{Nat}. (x < y) \wedge (y < z) \rightarrow (x < z)$

- Is it *valid*?
- No, e.g., if we interpret $<$ as the *successor* relation.

Motivation [2]

In practice, we often *do not care* about satisfiability or validity in *general*, but rather with respect to a *limited class* of interpretations.

A practical reason:

- When reasoning in a particular application domain, we typically have *specific* data types/structures in mind (e.g., integers, strings, lists, arrays, finite sets, ...).
- More generally, we are typically *not* interested in *arbitrary* interpretations, but rather in *specific* ones.

Theories formalize this domain-specific reasoning: we talk about satisfiability and validity *with respect to a theory* or “*modulo a theory*”.

A computational reason:

- The validity problem for FOL is *undecidable* in general.
- However, the validity problem for many *restricted* theories, is *decidable*.

First-Order Theories

Hereinafter, we assume that we have an infinite set of variables X .

Definition 4 (Theory): A first-order *theory* \mathcal{T} is a pair¹ $\langle \Sigma, M \rangle$, where

- $\Sigma = \langle \Sigma^S, \Sigma^F \rangle$ is a first-order signature,
- M is a class² of Σ -interpretations over X that is *closed under variable re-assignment*.

Definition 5: M is *closed under variable re-assignment* if every Σ -interpretation that differs from one in M in the way it interprets the variables in X is also in M .

A theory limits the interpretations of Σ -formulas to those from M .

¹Here, we use **bold** style for M to denote that it is *not a single* model, but a *collection* of them.

²*Class* is a generalization of a set.

Theory Examples

Example: Theory of Real Arithmetic $\mathcal{T}_{\text{RA}} = \langle \Sigma_{\text{RA}}, M_{\text{RA}} \rangle$:

- $\Sigma_{\text{RA}}^S = \{\text{Real}\}$
- $\Sigma_{\text{RA}}^F = \{+, -, \times, \leq\} \cup \{q \mid q \text{ is a decimal numeral}\}$
- All $\mathcal{I} \in M_{\text{RA}}$ interpret **Real** as the set of *real numbers* \mathbb{R} , each q as the *decimal number* that it denotes, and the function symbols in the usual way.

Example: Theory of Ternary Strings $\mathcal{T}_{\text{TS}} = \langle \Sigma_{\text{TS}}, M_{\text{TS}} \rangle$:

- $\Sigma_{\text{TS}}^S = \{\text{String}\}$
- $\Sigma_{\text{TS}}^F = \{\cdot, <\} \cup \{a, b, c\}$
- All $\mathcal{I} \in M_{\text{TS}}$ interpret **String** as the set $\{a, b, c\}^*$ of all finite strings over the characters $\{a, b, c\}$, symbol \cdot as string concatenation (e.g., $a \cdot b = ab$), and $<$ as lexicographic order.

\mathcal{T} -interpretations

Definition 6 (Reduct): Let Σ and Ω be two signatures over variables X , where $\Omega \supseteq \Sigma$, that is, $\Omega^S \supseteq \Sigma^S$ and $\Omega^F \supseteq \Sigma^F$.

Let \mathcal{J} be an Ω -interpretation over X .

The *reduct* \mathcal{J}^Σ of \mathcal{J} to Σ is a Σ -interpretation obtained from \mathcal{J} by restricting it to the symbols in Σ .

Definition 7 (\mathcal{T} -interpretation): Given a theory $\mathcal{T} = \langle \Sigma, \mathcal{M} \rangle$, a *\mathcal{T} -interpretation* is any Ω -interpretation \mathcal{J} for some signature $\Omega \supseteq \Sigma$ such that $\mathcal{J}^\Sigma \in \mathcal{M}$.

Note: This definition allows us to consider the satisfiability in a theory $\mathcal{T} = \langle \Sigma, \mathcal{M} \rangle$ of formulas that contain sorts or function symbols not in Σ . These symbols are usually called *uninterpreted* (in \mathcal{T}).

\mathcal{T} -interpretations [2]

Example: Consider again the theory of real arithmetic $\mathcal{T}_{\text{RA}} = \langle \Sigma_{\text{RA}}, \mathbf{M}_{\text{RA}} \rangle$.

All $\mathcal{J} \in \mathbf{M}_{\text{RA}}$ interpret `Real` as \mathbb{R} and function symbols as usual.

Which of the following interpretations are \mathcal{T}_{RA} -interpretations?

1. $\text{Real}^{\mathcal{J}_1} = \mathbb{Q}$, symbols in Σ_{RA}^F interpreted as usual. ✗
2. $\text{Real}^{\mathcal{J}_2} = \mathbb{R}$, symbols in Σ_{RA}^F interpreted as usual, and $\text{String}^{\mathcal{J}_2} = \{0.5, 1.3\}$. ✓
3. $\text{Real}^{\mathcal{J}_3} = \mathbb{R}$, symbols in Σ_{RA}^F interpreted as usual, and $\log^{\mathcal{J}_3}$ is the successor function. ✓

\mathcal{T} -satisfiability, \mathcal{T} -entailment, \mathcal{T} -validity

Definition 8 (\mathcal{T} -satisfiability): A Σ -formula α is *satisfiable in \mathcal{T}* , or *\mathcal{T} -satisfiable*, if it is satisfied by *some* \mathcal{T} -interpretation \mathcal{I} .

Definition 9 (\mathcal{T} -entailment): A set Γ of formulas *\mathcal{T} -entails* a formula α , if every \mathcal{T} -interpretation that satisfies all formulas in Γ also satisfies α .

Definition 10 (\mathcal{T} -validity): A formula α is *\mathcal{T} -valid*, if it is satisfied by *all* \mathcal{T} -interpretations.

Note: A formula α is *\mathcal{T} -valid* iff $\emptyset \models \alpha$.

Example: Which of the following Σ_{RA} -formulas is satisfiable or valid in \mathcal{T}_{RA} ?

1. $(x_0 + x_1 \leq 0.5) \wedge (x_0 - x_1 \leq 2)$
2. $\forall x_0. (x_0 + x_1 \leq 1.7) \rightarrow (x_1 \leq 1.7 - x_0)$
3. $\forall x_0. \forall x_1. (x_0 + x_1 \leq 1)$

satisfiable, falsifiable

satisfiable, valid

unsatisfiable, falsifiable

FOL vs Theory

For every signature Σ , entailment and validity in “pure” FOL can be seen as entailment and validity in the theory $\mathcal{T}_{\text{FOL}} = \langle \Sigma, \mathbf{M}_{\text{FOL}} \rangle$ where \mathbf{M}_{FOL} is the class of *all possible* Σ -interpretations.

- Pure first-order logic = reasoning over *all* possible interpretations.
- Reasoning modulo a theory = *restricting* interpretations with some domain constraints.
- Theories make automated reasoning *feasible* in many domains.

Axiomatization

Definition 11 (Axiomatic theory): A first-order *axiomatic theory* \mathcal{T} is defined by a signature Σ and a set \mathcal{A} of Σ -sentences, or *axioms*.

Definition 12 (\mathcal{T} -validity in axiomatic theory): An Ω -formula α is *valid* in an axiomatic theory \mathcal{T} if it is entailed by the axioms of \mathcal{T} , that is, every Ω -interpretation \mathcal{I} that satisfies \mathcal{A} also satisfies α .

Note: Axiomatic theories are a *special case* of the general definition (via \mathbf{M}) of theories.

- Given an axiomatic theory \mathcal{T}' defined by Σ and \mathcal{A} , we can define a theory $\mathcal{T} = \langle \Sigma, \mathbf{M} \rangle$ where \mathbf{M} is the class of all Σ -interpretations that satisfy all axioms in \mathcal{A} .
- It is not hard to show that a formula α is valid in \mathcal{T} *iff* it is valid in \mathcal{T}' .

Note: Not all theories are first-order axiomatizable.

Non-Axiomatizable Theories

Note: Not all theories are first-order axiomatizable.

Example: Consider the theory \mathcal{T}_{Nat} of the natural numbers, with signature Σ with $\Sigma^S = \{\text{Nat}\}$, $\Sigma^F = \{0, S, +, <\}$, and $M = \{\mathcal{I}\}$ where $\text{Nat}^{\mathcal{I}} = \mathbb{N}$ and Σ^F is interpreted as usual.

Any set of axioms (for example, *Peano axioms*) for this theory is satisfied by *non-standard models*, e.g., interpretations \mathcal{I}' where $\text{Nat}^{\mathcal{I}'}$ includes other chains of elements besides the natural numbers.

However, these models *falsify* formulas that are *valid* in \mathcal{T}_{Nat} .

For example, “every number is either zero or a successor”: $\forall x. (x \doteq 0) \vee \exists y. (x \doteq S(y))$.

- **true** in the *standard* model, i.e. $\text{Nat}^{\mathcal{I}} = \mathbb{N} = \{0, 1 := S(0), 2 := S(1), \dots\}$.
- **false** in *non-standard* models, e.g., $\text{Nat}^{\mathcal{I}'} = \{0, 1, 2, \dots\} \cup \{\omega, \omega + 1, \dots\}$
 - ▶ Intuitively, ω is “an infinite element”.
 - ▶ The successor function still applies: $S(\omega) = \omega + 1$, $S(\omega + 1) = \omega + 2$, etc.
 - ▶ Even the addition and multiplication still works: $\omega + 3 = S(S(S(\omega)))$, $\omega \times 2 = \omega + \omega$.
 - ▶ But ω is larger than all standard numbers: $\omega > 0, \omega > 1, \dots$

Peano Arithmetic

Definition 13: *Peano arithmetic* \mathcal{T}_{PA} , or *first-order arithmetic*, is the axiomatic theory of natural numbers with signature $\Sigma_{\text{PA}}^F = \{0, S, +, \times, =\}$ and *Peano axioms*:

1. $\forall x. (S(x) \neq 0)$ (zero)
2. $\forall x. \forall y. (S(x) = S(y)) \rightarrow (x = y)$ (successor)
3. $F[0] \wedge (\forall x. F[x] \rightarrow F[x + 1]) \rightarrow \forall x. F[x]$ (induction)
4. $\forall x. (x + 0 = x)$ (plus zero)
5. $\forall x. \forall y. (x + S(y) = S(x + y))$ (plus successor)
6. $\forall x. (x \times 0 = 0)$ (times zero)
7. $\forall x. \forall y. (x \times S(y) = (x \times y) + x)$ (times successor)

Axiom (induction) is the *induction axiom schema*. It stands for an *infinite* set of axioms, one for each Σ_{PA} -formula F with one free variable. The notation $F[\alpha]$ means that F contains α as a sub-formula.

The *intended interpretation* (*standard models*) of \mathcal{T}_{PA} have the domain \mathbb{N} and the usual interpretations of the function symbols as $0_{\mathbb{N}}$, $S_{\mathbb{N}}$, $+_{\mathbb{N}}$, and $\times_{\mathbb{N}}$.

Presburger Arithmetic

Note: Satisfiability and validity in $\mathcal{T}_{\mathcal{P}_A}$ is undecidable. Therefore, we need a more restricted theory of arithmetic that does not include multiplication.

Definition 14: *Presburger arithmetic* $\mathcal{T}_{\mathbb{N}}$ is the axiomatic theory of natural numbers with signature $\Sigma_{\mathbb{N}}^F = \{0, S, +, =\}$ and the *subset* of *Peano axioms*:

1. $\forall x. (S(x) \neq 0)$ (zero)
2. $\forall x. \forall y. (S(x) = S(y)) \rightarrow (x = y)$ (successor)
3. $F[0] \wedge (\forall x. F[x] \rightarrow F[x + 1]) \rightarrow \forall x. F[x]$ (induction)
4. $\forall x. (x + 0 = x)$ (plus zero)
5. $\forall x. \forall y. (x + S(y) = S(x + y))$ (plus successor)

Note: Presburger arithmetic is decidable.

Completeness of Theories

Definition 15: A Σ -theory \mathcal{T} is *complete* if for every Σ -sentence α , either α or $\neg\alpha$ is valid in \mathcal{T} .

Note: In a complete Σ -theory, every Σ -sentence is either valid or unsatisfiable.

Example: Any theory $\mathcal{T} = \langle \Sigma, \mathbf{M} \rangle$ where all interpretations in \mathbf{M} only differ in how they interpret the variables (e.g., \mathcal{T}_{RA}) is *complete*.

Example: The axiomatic (mono-sorted) theory of *monoids* with $\Sigma^F = \{ \cdot, \varepsilon \}$ and axioms

$$\forall x. \forall y. \forall z. (x \cdot y) \cdot z \doteq x \cdot (y \cdot z) \quad \forall x. (x \cdot \varepsilon \doteq x) \quad \forall x. (\varepsilon \cdot x \doteq x)$$

is *incomplete*. For example, the sentence $\forall x. \forall y. (x \cdot y \doteq y \cdot x)$ is *true* in some monoids (e.g. the addition of integers *is* commutative) but *false* in others (e.g. the concatenation of strings *is not* commutative).

Completeness of Theories [2]

Example: The axiomatic (mono-sorted) theory of *dense linear orders without endpoints* with $\Sigma^F = \{<\}$ and the following axioms is *complete*.

$$\forall x. \forall y. (x < y) \rightarrow \exists z. ((x < z) \wedge (z < y)) \quad (\text{dense})$$

$$\forall x. \forall y. ((x < y) \vee (y < x) \vee (x = y)) \quad (\text{linear})$$

$$\forall x. \neg(x < x) \quad \forall x. \forall y. \forall z. ((x < y) \wedge (y < z) \rightarrow (x < z)) \quad (\text{orders})$$

$$\forall x. \exists y. (y < x) \quad \forall x. \exists y. (x < y) \quad (\text{without endpoints})$$

Decidability and Fragments

Recall that a set A is *decidable* if there exists a *terminating* procedure that, given an input element a , returns (after *finite* time) either “yes” if $a \in A$ or “no” if $a \notin A$.

Definition 16: A theory $\mathcal{T} = \langle \Sigma, M \rangle$ is *decidable* if the set of all \mathcal{T} -valid Σ -formulas is decidable.

Definition 17: A *fragment* of \mathcal{T} is a *syntactically-restricted subset* of \mathcal{T} -valid Σ -formulas.

Example: The *quantifier-free* fragment of \mathcal{T} is the set of all \mathcal{T} -valid Σ -formulas *without quantifiers*. The *linear* fragment of \mathcal{T}_{RA} is the set of all \mathcal{T} -valid Σ_{RA} -formulas *without multiplication* (\times).

Axiomatizability

Definition 18: A theory $\mathcal{T} = \langle \Sigma, M \rangle$ is *recursively axiomatizable* if M is the class of all interpretations satisfying a *decidable set* of first-order axioms \mathcal{A} .

Theorem 1 (Lemma): Every recursively axiomatizable theory \mathcal{T} admits a procedure $E_{\mathcal{T}}$ that *enumerates* all \mathcal{T} -valid formulas.

Theorem 2: For every *complete* and *recursively axiomatizable* theory \mathcal{T} , validity in \mathcal{T} is decidable.

Proof: Given a formula α , use $E_{\mathcal{T}}$ to enumerate all valid formulas. Since \mathcal{T} is complete, either α or $\neg\alpha$ will eventually (after *finite* time) be produced by $E_{\mathcal{T}}$. □

Introduction to SMT

Common Theories in SMT

Satisfiability Modulo Theories (SMT) traditionally focuses on theories with *decidable quantifier-free fragments*.

SMT is concerned with (un)satisfiability, but recall that a formula α is \mathcal{T} -*valid* iff $\neg\alpha$ is \mathcal{T} -*unsatisfiable*.

Checking the (un)satisfiability of quantifier-free formulas in main background theories *efficiently* has a large number of applications in:

- hardware and software verification
- model checking
- symbolic execution
- compiler validation
- type checking
- planning and scheduling
- software synthesis
- cyber-security
- verifiable machine learning
- analysis of biological systems

Further, we are going to study:

- A few of those *theories* and their *decision procedures*.
- *Proof systems* to reason *modulo theories* automatically.

From Quantifier-Free Formulas to Conjunctions of Literals

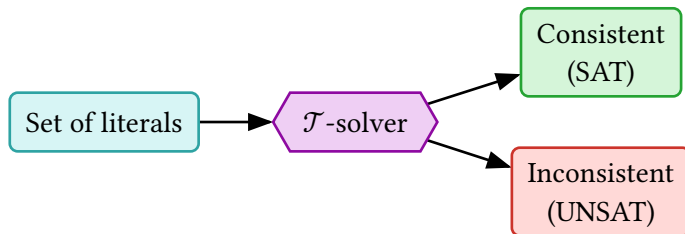
Theorem 3: The satisfiability of *quantifier-free* formulas in a theory \mathcal{T} is *decidable* iff the satisfiability in \mathcal{T} of *conjunctions of literals* is decidable.

Here, *literal* is an atom or its negation. For example: $(a \doteq b)$, $\neg(a + 1 < b)$, $(f(b) \doteq g(f(a)))$.

Proof: A quantifier-free formula can be transformed into disjunctive normal form (DNF), and its satisfiability reduces to checking satisfiability of conjunctions of literals. Conversely, a conjunction of literals is a special case of a quantifier-free formula. Thus, the two satisfiability problems are equivalent. \square

Theory Solvers

Definition 19 (\mathcal{T} -solver): A *theory solver*, or \mathcal{T} -*solver*, is a specialized decision procedure for the satisfiability of conjunctions of literals in a theory \mathcal{T} .



Key insight: A \mathcal{T} -solver is the bridge between SAT and theory reasoning. It answers: “Is this set of theory literals consistent in this domain?”

Theory of Uninterpreted Functions

Definition 20: Given a signature Σ , the most general theory consists of the class of *all* Σ -interpretations. In fact, this is a *family* of theories parameterized by the signature Σ .

It is known as the theory of *equality with uninterpreted functions* \mathcal{T}_{EUF} , or the *empty theory*, since it contains no *sentences*.

Example: $(a \doteq b) \wedge (f(a) \doteq b) \wedge \neg(g(a) \doteq g(f(a)))$ Is this formula satisfiable in \mathcal{T}_{EUF} ?

Both validity and satisfiability are undecidable in \mathcal{T}_{EUF} .

- Validity in \mathcal{T}_{EUF} is *semi-decidable* — this is just a validity in FOL.
- Since a formula α is \mathcal{T} -satisfiable iff $\neg\alpha$ is not \mathcal{T} -valid, \mathcal{T}_{EUF} -satisfiability is *co-recognizable*.

However, the satisfiability of *conjunctions of \mathcal{T}_{EUF} -literals* is *decidable*, in polynomial time, using the *congruence closure* algorithm.

Theory of Real Arithmetic

Definition 21: The theory of *real arithmetic* \mathcal{T}_{RA} is a theory of inequalities over the real numbers.

- $\Sigma^S = \{\text{Real}\}$
- $\Sigma^F = \{+, -, \times, <\} \cup \{q \mid q \text{ is a decimal numeral}\}$
- \mathcal{M} is the class of interpretations that interpret Real as the set of *real numbers* \mathbb{R} , and the function symbols in the usual way.

Satisfiability in the full \mathcal{T}_{RA} is *decidable* (in worst-case doubly-exponential time).

Restricted fragments of \mathcal{T}_{RA} can be decided more efficiently.

Example: Quantifier-free linear real arithmetic (QF_LRA) is the theory of *linear* inequalities over the reals, where \times can only be used in the form of *multiplication by constants* (decimal numerals).

The satisfiability of conjunctions of literals in QF_LRA is *decidable* in *polynomial time*.

Theory of Integer Arithmetic

Definition 22: The theory of *integer arithmetic* \mathcal{T}_{IA} is a theory of inequalities over the integers.

- $\Sigma^S = \{\text{Int}\}$
- $\Sigma^F = \{+, -, \times, <\} \cup \{n \mid n \text{ is an integer numeral}\}$
- \mathcal{M} is the class of interpretations that interpret Int as the set of *integers* \mathbb{Z} , and the function symbols in the usual way.

Satisfiability in \mathcal{T}_{IA} is *not even semi-decidable*!

Satisfiability of quantifier-free Σ -formulas in \mathcal{T}_{IA} is *undecidable* as well.

Linear integer arithmetic (LIA, also known as *Presburger arithmetic*) is decidable, but not efficiently (in worst-case triply-exponential time). Its quantifier-free fragment (QF_LIA) is NP-complete.

Theory of Arrays with Extensionality

Definition 23: The theory of *arrays* \mathcal{T}_{AX} is useful for modelling RAM or array data structures.

- $\Sigma^S = \{A, I, E\}$ (arrays, indices, elements)
- $\Sigma^F = \{\text{read}, \text{write}\}$, where $\text{rank}(\text{read}) = \langle A, I, E \rangle$ and $\text{rank}(\text{write}) = \langle A, I, E, A \rangle$

Let a be a variable of sort A , variable i of sort I , and variable v of sort E .

- $\text{read}(a, i)$ denotes the value stored in array a at index i .
- $\text{write}(a, i, v)$ denotes the array that stores value v at index i and is otherwise identical to a .

Example: $\text{read}(\text{write}(a, i, v), i) \doteq_E v$

- Is this formula *intuitively* valid/satisfiable/unsatisfiable in \mathcal{T}_A ?

Example: $\forall i. (\text{read}(a, i) \doteq_E \text{read}(a', i)) \rightarrow (a \doteq_A a')$

- Is this formula *intuitively* valid/satisfiable/unsatisfiable in \mathcal{T}_A ?

Theory of Arrays with Extensionality [2]

Definition 24: The theory of arrays $\mathcal{T}_{AX} = \langle \Sigma, M \rangle$ is finitely axiomatizable.

M is the class of interpretations that satisfy the following axioms:

1. $\forall a. \forall i. \forall v. (\text{read}(\text{write}(a, i, v), i) \doteq_E v)$
2. $\forall a. \forall i. \forall j. \forall v. \neg(i \doteq_I j) \rightarrow (\text{read}(\text{write}(a, i, v), j) \doteq_E \text{read}(a, j))$
3. $\forall a. \forall b. (\forall i. (\text{read}(a, i) \doteq_E \text{read}(b, i))) \rightarrow (a \doteq_A b)$

Note: The last axiom is called *extensionality* axiom. It states that two arrays are equal if they have the same values at all indices. It can be omitted to obtain a theory of arrays *without extensionality* \mathcal{T}_A .

Validity and satisfiability in \mathcal{T}_{AX} is *undecidable*.

There are several *decidable fragments* of \mathcal{T}_A .

Survey of Decidability and Complexity

Theory	Description	Full	QF	Full complexity	QFC complexity
PL	Propositional Logic	—	yes	NP-complete	$\Theta(n)$
\mathcal{T}_{EUF}	Equality	no	yes	undecidable	$\mathcal{O}(n \log n)$
\mathcal{T}_{PA}	Peano Arithmetic	no	no	undecidable	undecidable
$\mathcal{T}_{\mathbb{N}}$	Presburger Arithmetic	yes	yes	$\Omega(2^{2^n}), \mathcal{O}(2^{2^{kn}})$	NP-complete
$\mathcal{T}_{\mathbb{Z}}$	Linear Integers (LIA)	yes	yes	$\Omega(2^{2^n}), \mathcal{O}(2^{2^{kn}})$	NP-complete
$\mathcal{T}_{\mathbb{R}}$	Reals	yes	yes	$\mathcal{O}(2^{2^{kn}})$	$\mathcal{O}(2^{2^{kn}})$
$\mathcal{T}_{\mathbb{Q}}$	Linear Rationals	yes	yes	$\Omega(2^n), \mathcal{O}(2^{2^{kn}})$	PTIME
\mathcal{T}_{RDS}	Recursive Data Structures	no	yes	undecidable	$\mathcal{O}(n \log n)$
$\mathcal{T}_{\text{ARDS}}$	Acyclic RDS	yes	yes	not elementary recursive	$\Theta(n)$
\mathcal{T}_{A}	Arrays	no	yes	undecidable	NP-complete
\mathcal{T}_{AX}	Arrays with Extensionality	no	yes	undecidable	NP-complete

Survey of Decidability and Complexity [2]

Legend for the table:

- “**Full**” denotes the decidability of a complete theory *with* quantifiers.
- “**QF**” denotes the decidability of a *quantifier-free* theory.
- “**Full complexity**” denotes the complexity of the satisfiability in a complete theory *with quantifiers*.
- “**QFC complexity**” denotes the complexity of the satisfiability in a *quantifier-free conjunctive* fragment.
- For complexities, n is the size of the input formula, k is some positive integer.
- “*Not elementary recursive*” means the runtime cannot be bounded by a fixed-height stack of exponentials.

Roadmap

In the following sections, we will study *theory solvers* for several important theories:

1. **Difference Logic** (DL) — a very restricted fragment of integer arithmetic; solved via *graph-based* cycle detection.
2. **Equality with Uninterpreted Functions** (\mathcal{T}_{EUF}) — the “empty” theory; solved via *congruence closure*.
3. **Arrays** (\mathcal{T}_{AX}) — models memory/arrays; extends the EUF proof system.
4. **Linear Real Arithmetic** (\mathcal{T}_{RA}) — linear inequalities over \mathbb{R} ; solved via the *Simplex* algorithm.

We will then see how to *combine* theory solvers (Nelson-Oppen method), build a complete *SMT solver* (CDCL(\mathcal{T}) architecture), and use the *SMT-LIB* standard language with the Z3 solver.

Difference Logic

Difference Logic

Definition 25: *Difference logic* (DL) is a fragment of linear integer arithmetic consisting of conjunctions of literals of the very restricted form:

$$x - y \bowtie c$$

where x and y are integer variables, c is a numeral, and $\bowtie \in \{=, <, \leq, >, \geq\}$.

A solver for difference logic consists of three steps:

1. Literals normalization.
2. Conversion to a graph.
3. Cycle detection.

Decision Procedure for DL

Step 1: Rewrite each literal using \leq by applying the following rules:

1. $(x - y = c) \rightarrow (x - y \leq c) \wedge (x - y \geq c)$
2. $(x - y \geq c) \rightarrow (y - x \leq -c)$
3. $(x - y > c) \rightarrow (y - x < -c)$
4. $(x - y < c) \rightarrow (x - y \leq c - 1)$

Step 2: Construct a weighted directed graph G with a vertex for each variable and an edge $x \xrightarrow{c} y$ for each literal $(x - y \leq c)$.

Step 3: Check for *negative cycles* in G .

- Use, for example, the Bellman-Ford algorithm.
- If G contains a negative cycle, the set of literals is *inconsistent* (UNSAT).
- Otherwise, the set of literals is *consistent* (SAT).

Difference Logic Example

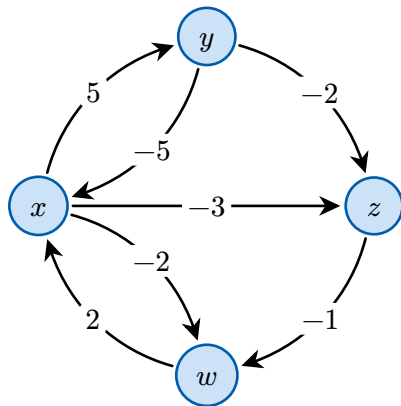
Consider the following set of difference logic literals:

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

Normalize the literals:

- $(x - y = 5) \implies (x - y \leq 5) \wedge (y - x \leq -5)$
- $(z - y \geq 2) \implies (y - z \leq -2)$
- $(z - x > 2) \implies (x - z \leq -3)$
- $(w - x = 2) \implies (w - x \leq 2) \wedge (x - w \leq -2)$
- $(z - w < 0) \implies (z - w \leq -1)$

UNSAT because of the negative cycle: $x \xrightarrow{-3} z \xrightarrow{-1} w \xrightarrow{2} x$.



Equality

Theory of Equality with Uninterpreted Functions

Definition 26: The theory of equality with uninterpreted functions \mathcal{T}_{EUF} is defined by the signature $\Sigma^F = \{\dot{=}, f, g, h, \dots\}$ (*interpreted* equality and *uninterpreted* functions) and the following axioms:

1. $\forall x. x \dot{=} x$ (reflexivity)
2. $\forall x. \forall y. (x \dot{=} y) \rightarrow (y \dot{=} x)$ (symmetry)
3. $\forall x. \forall y. \forall z. (x \dot{=} y) \wedge (y \dot{=} z) \rightarrow (x \dot{=} z)$ (transitivity)
4. $\forall x. \forall y. (x = y) \rightarrow (f(x) \dot{=} f(y))$ (function congruence)

Flattening

Definition 27: A literal is *flat* if it is of the form:

- $x \doteq y$
- $\neg(x \doteq y)$
- $x \doteq f(z)$

where x and y are variables, f is a function symbol, and z is a tuple of 0 or more variables.

Note: Any set of literals can be converted to an equisatisfiable set of *flat* literals by introducing *new* variables and equating non-equational atoms to *true*.

Example: Consider the set of literals: $\{x + y > 0, y \doteq f(g(z))\}$.

We can convert it to an equisatisfiable set of flat literals by introducing fresh variables v_i :

$$\{ v_1 \doteq v_2 > v_3, \quad v_1 \doteq \text{true}, \quad v_2 \doteq x + y, \quad v_3 \doteq 0, \quad y \doteq f(v_4), \quad v_4 \doteq g(z) \}$$

Hereinafter, we will assume that all literals are *flat*.

Notation and Assumptions

- We abbreviate $\neg(s \doteq t)$ with $s \not\doteq t$.
- For tuples $\mathbf{u} = \langle u_1, \dots, u_n \rangle$ and $\mathbf{v} = \langle v_1, \dots, v_n \rangle$, we abbreviate $(u_1 \doteq v_1) \wedge \dots \wedge (u_n \doteq v_n)$ with $\mathbf{u} = \mathbf{v}$.
- Γ is used to refer to the “current” proof state in rule premises.
- $\Gamma, s \doteq t$ is an abbreviation for $\Gamma \cup \{s \doteq t\}$.
- If applying a rule R does not change Γ , then R *is not applicable* to Γ , that is, Γ is *irreducible* w.r.t. R .

Satisfiability Proof System for QF_UF

Let QF_UF be the quantifier-free fragment of FOL over some signature Σ .

Below is a simple *satisfiability proof system* R_{UF} for QF_UF:

$$\mathbf{REFL} \frac{x \text{ occurs in } \Gamma}{\Gamma := \Gamma, x \doteq x}$$

$$\mathbf{SYMM} \frac{x \not\doteq y \in \Gamma}{\Gamma := \Gamma, y \doteq x}$$

$$\mathbf{TRANS} \frac{x \not\doteq y \in \Gamma \quad y \doteq z \in \Gamma}{\Gamma := \Gamma, x \doteq z}$$

$$\mathbf{CONG} \frac{x \doteq f(\mathbf{u}) \in \Gamma \quad y \doteq f(\mathbf{v}) \in \Gamma \quad \mathbf{u} = \mathbf{v} \in \Gamma}{\Gamma := \Gamma, x \doteq y}$$

$$\mathbf{CONTR} \frac{x \doteq y \in \Gamma \quad x \not\doteq y \in \Gamma}{\text{UNSAT}}$$

$$\mathbf{SAT} \frac{\text{No other rules apply}}{\text{SAT}}$$

Is R_{UF} *sound*?

Is R_{UF} *terminating*?

Example Derivation in R_{UF}

Example: Determine the satisfiability of the following set of literals: $a \doteq f(f(a))$, $a \doteq f(f(f(a)))$, $g(a, f(a)) \not\doteq g(f(a), a)$. Flatten the literals and construct the following proof:

$$\begin{array}{l} \frac{a \doteq f(a_1), a_1 \doteq f(a), a \doteq f(a_2), a_2 \doteq f(a_1), a_3 \not\doteq a_4, a_3 \doteq g(a, a_1), a_4 \doteq g(a_1, a)}{\text{REFL}} \\ \frac{a_1 \doteq a_1}{\text{CONG applied to } a \doteq f(a_1), a_2 \doteq f(a_1), a_1 \doteq a_1} \\ \frac{a \doteq a_2}{\text{CONG applied to } a_1 \doteq f(a), a \doteq f(a_2), a \doteq a_2} \\ \frac{a_1 \doteq a}{\text{SYMM}} \\ \frac{a \doteq a_1}{\text{CONG applied to } a_3 \doteq g(a, a_1), a_4 \doteq g(a_1, a), a \doteq a_1, a_1 \doteq a} \\ \frac{a_3 \doteq a_4}{\text{CONTR applied to } a_3 \doteq a_4, a_3 \not\doteq a_4} \\ \text{UNSAT} \end{array}$$

Soundness of R_{UF}

Theorem 4 (Refutation soundness): A literal set Γ_0 is unsatisfiable if R_{UF} derives UNSAT from it.

Proof: All rules except SAT are satisfiability-preserving.

If a derivation from Γ_0 ends with UNSAT, then Γ_0 must be unsatisfiable. □

Theorem 5 (Solution soundness): A literal set Γ_0 is satisfiable if R_{UF} derives SAT from it.

Proof: Let Γ be a proof state to which SAT applies. From Γ , we can construct an interpretation \mathcal{J} that satisfies Γ_0 . Let $s \sim t$ iff $(s \doteq t) \in \Gamma$. One can show that \sim is an equivalence relation.

Let the domain of \mathcal{J} be the equivalence classes E_1, \dots, E_k of \sim .

- For every variable or a constant t , let $t^{\mathcal{J}} = E_i$ if $t \in E_i$ for some i . Otherwise, let $t^{\mathcal{J}} = E_1$.
- For every unary function symbol f , and equivalence class E_i , let $f^{\mathcal{J}}$ be such that $f^{\mathcal{J}}(E_i) = E_j$ if $f(t) \in E_j$ for some $t \in E_i$. Otherwise, let $f^{\mathcal{J}}(E_i) = E_1$. Define $f^{\mathcal{J}}$ for non-unary f similarly.

We can show that $\mathcal{J} \models \Gamma$. This means that \mathcal{J} models Γ_0 as well since $\Gamma_0 \subseteq \Gamma$. □

Termination in R_{UF}

Theorem 6: Every derivation strategy for R_{UF} terminates.

Proof: R_{UF} adds to the current state Γ only equalities between variables of Γ_0 .

So, at some point it will run out of new equalities to add.

□

Completeness of R_{UF}

Theorem 7 (Refutation completeness): Every derivation strategy applied to an unsatisfiable state Γ_0 ends with UNSAT.

Proof: Let Γ_0 be an unsatisfiable state. Suppose there was a derivation from Γ_0 that did not end with UNSAT. Then, by the termination theorem, it would have to end with SAT. But then R_{UF} would be not be solution sound. \square

Theorem 8 (Solution completeness): Every derivation strategy applied to a satisfiable state Γ_0 ends with SAT.

Proof: Let Γ_0 be a satisfiable state. Suppose there was a derivation from Γ_0 that did not end with SAT. Then, by the termination theorem, it would have to end with UNSAT. But then R_{UF} would be not be refutation sound. \square

Arrays

Theory of Arrays

Definition 28: The theory of *arrays* \mathcal{T}_{AX} is defined by the signature $\Sigma^S = \{A, I, E\}$ (arrays, indices, elements), $\Sigma^F = \{\text{read}, \text{write}\}$ and the following axioms:

1. $\forall a. \forall i. \forall v. (\text{read}(\text{write}(a, i, v), i) \doteq_E v)$
2. $\forall a. \forall i. \forall j. \forall v. \neg(i \doteq_I j) \rightarrow (\text{read}(\text{write}(a, i, v), j) \doteq_E \text{read}(a, j))$
3. $\forall a. \forall b. (\forall i. (\text{read}(a, i) \doteq_E \text{read}(b, i))) \rightarrow (a \doteq_A b)$

Example

```
void ReadBlock(int data[], int x, int len) {  
    int i = 0;  
    int next = data[0];  
    for (; i < next && i < len; i = i + 1) {  
        if (data[i] == x)  
            break;  
        else  
            Process(data[i]);  
    }  
    assert(i < len);  
}
```

One pass through this code can be translated into the following \mathcal{T}_A formula:

$$(i \doteq 0) \wedge (\text{next} \doteq \text{read}(\text{data}, 0)) \wedge (i < \text{next}) \wedge \\ \wedge (i < \text{len}) \wedge (\text{read}(\text{data}, i) \doteq x) \wedge \neg(i < \text{len})$$

Satisfiability Proof System for QF_AX

The satisfiability proof system R_{AX} for \mathcal{T}_{AX} *extends* the proof system R_{UF} for \mathcal{T}_{UF} with the following rules:

$$\begin{array}{c} \mathbf{RINTRO1} \frac{b \doteq \text{write}(a, i, v) \in \Gamma}{\Gamma := \Gamma, v \doteq \text{read}(b, i)} \\[2ex] \mathbf{RINTRO2} \frac{b \doteq \text{write}(a, i, v) \in \Gamma \quad u \doteq \text{read}(x, j) \in \Gamma \quad x \in \{a, b\}}{\Gamma := \Gamma, i \doteq j \quad \Gamma := \Gamma, i \neq j, u \doteq \text{read}(a, j), u \doteq \text{read}(b, j)} \\[2ex] \mathbf{EXT} \frac{a \neq b \in \Gamma \quad a \text{ and } b \text{ are arrays}}{\Gamma := \Gamma, u \neq v, u \doteq \text{read}(a, k), v \doteq \text{read}(b, k)} \end{array}$$

- **RINTRO1**: After writing v at index i , the reading at the same index i gives us back the value v .
- **RINTRO2**: After writing v in a at index i , the reading from a or b at index j *splits* in two cases: (1) i equals j , (2) a and b have the same value u at position j .
- **EXT**: If two arrays a and b are distinct, they must differ at some index k .

Example Derivation in R_{AX}

Example: Determine the satisfiability of $\{\text{write}(a_1, i, \text{read}(a_1, i)) \doteq \text{write}(a_2, i, \text{read}(a_2, i)), a_1 \not\doteq a_2\}$.

First, flatten the literals:

$$\begin{aligned} & \{\text{write}(a_1, i, \text{read}(a_1, i)) \doteq \text{write}(a_2, i, \text{read}(a_2, i))\} \rightarrow \\ & \rightarrow \{a'_1 \doteq a'_2, a'_1 \doteq \text{write}(a_1, i, \text{read}(a_2, i)), a'_2 \doteq \text{write}(a_2, i, \text{read}(a_1, i)), a_1 \not\doteq a_2\} \rightarrow \\ & \rightarrow \{a'_1 \doteq a'_2, a'_1 \doteq \text{write}(a_1, i, v_2), v_2 \doteq \text{read}(a_2, i), a'_2 \doteq \text{write}(a_2, i, v_1), v_1 \doteq \text{read}(a_1, i), a_1 \not\doteq a_2\} \end{aligned}$$

Example Derivation in R_{AX} [2]

1. $a'_1 \doteq a'_2, a'_1 \doteq \text{write}(a_1, i, v_2), v_2 \doteq \text{read}(a_2, i), a'_2 \doteq \text{write}(a_2, i, v_1), v_1 \doteq \text{read}(a_1, i), a_1 \not\doteq a_2$
2. (by REFL) $a_1 \doteq a_1$
3. (by REFL) $a_2 \doteq a_2$
4. (by EXT) $u_1 \not\doteq u_2, u_1 \doteq \text{read}(a_1, n), u_2 \doteq \text{read}(a_2, n)$
5. (by RINTRO2) split

-
6. $i \doteq n$
 7. (by CONG) $v_1 \doteq u_1$
 8. (by SYMM) $u_1 \doteq v_1$
 9. (by CONG) $v_2 \doteq u_2$
 10. (by RINTRO1) $v_2 \doteq \text{read}(a'_1, i)$
 11. (by RINTRO1) $v_1 \doteq \text{read}(a'_2, i)$
 12. (by REFL) $i \doteq i$
 13. (by CONG) $v_1 \doteq v_2$
 14. (by TRANS) $u_1 \doteq u_2$
 15. (by CONTR) UNSAT

-
6. $i \not\doteq n, u_1 \doteq \text{read}(a'_1, n)$
 7. (by RINTRO2) split

-
8. $i \doteq n$
 9. (by CONTR) UNSAT

-
8. $i \not\doteq n, u_2 \doteq \text{read}(a'_2, n)$
 9. (by REFL) $n \doteq n$
 10. (by CONG) $u_1 \doteq u_2$
 11. (by CONTR) UNSAT

Arithmetic

Theory of Real Arithmetic

Definition 29: The theory of *real arithmetic* \mathcal{T}_{RA} is defined by the signature $\Sigma_{\text{RA}}^S = \{\text{Real}\}$, $\Sigma_{\text{RA}}^F = \{+, -, \times, \leq\} \cup \{q \mid q \text{ is a decimal numeral}\}$ and the class of interpretations \mathcal{M}_{RA} that interpret Real as the set of *real numbers* \mathbb{R} , and the function symbols in the usual way.

Quantifier-free linear real arithmetic (QF_LRA) is the theory of *linear inequalities* over the reals, where \times can only be used in the form of *multiplication by constants (decimal numerals)*.

Linear Programming

Definition 30: A *linear program* (LP) consists of:

1. An $m \times n$ matrix \mathbf{A} , the *constraint matrix*.
2. An m -dimensional vector \mathbf{b} .
3. An n -dimensional vector \mathbf{c} , the *objective function*.

Let \mathbf{x} be a vector of n variables.

Goal: Find a solution \mathbf{x} that *maximizes* $\mathbf{c}^T \mathbf{x}$ subject to the linear constraints $\mathbf{A}\mathbf{x} \leq \mathbf{b}$ (and³ $\mathbf{x} \geq \mathbf{0}$).

Note: All **bold**-styled symbols denote *vectors* or *matrices*, e.g., \mathbf{x} , \mathbf{A} , $\mathbf{0}$.

³The constraint $\mathbf{x} \geq \mathbf{0}$ is introduced when LP is expressed in *standard form*, explained later in these slides.

Example and Terminology

Example: Maximize $2x_2 - x_1$ subject to:

$$x_1 + x_2 \leq 3$$

$$2x_1 - x_2 \leq -5$$

Here, $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, $\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix}$, $\mathbf{b} = \begin{bmatrix} 3 \\ -5 \end{bmatrix}$, $\mathbf{c} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$.

Find \mathbf{x} that maximizes $\mathbf{c}^T \mathbf{x}$ subject to $\mathbf{Ax} \leq \mathbf{b}$.

Definition 31: An assignment of \mathbf{x} is a *feasible solution* if it satisfies $\mathbf{Ax} \leq \mathbf{b}$.

- Is $\mathbf{x} = \langle 0, 0 \rangle$ a feasible solution? ✗
- Is $\mathbf{x} = \langle -2, 1 \rangle$ a feasible solution? ✓

Definition 32: For a given assignment \mathbf{x} , the value $\mathbf{c}^T \mathbf{x}$ is the *objective value*, or *cost*, of \mathbf{x} .

- What is the objective value of $\mathbf{x} = \langle -2, 1 \rangle$?

Example and Terminology [2]

Definition 33: An *optimal solution* is a feasible solution with a *maximal* objective value among all feasible solutions.

Definition 34: If a linear program has no feasible solutions, it is *infeasible*.

Definition 35: The linear program is *unbounded* if the objective value of the optimal solution is ∞ .

Geometric Interpretation

Definition 36: A *polytope* is a generalization of 3-dimensional polyhedra to higher dimensions.

Definition 37: A polytope P is *convex* if every point on the line segment connecting any two points in P is also within P .

Formally, for all $a, b \in \mathbb{R}^n \cap P$, and for all $\lambda \in [0; 1]$, it holds that $\lambda a + (1 - \lambda)b \in P$.

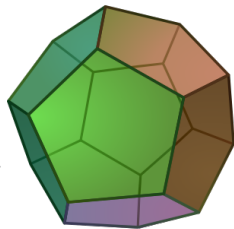
Note: For an $m \times n$ constraint matrix A , the set of points $P = \{x \mid Ax \leq b\}$ forms a *convex polytope* in n -dimensional space.

LP goal: find a point x *inside the polytope* that maximizes $c^T x$ for a given c .

Note: LP is *infeasible* iff the polytope is *empty*.

Note: LP is *unbounded* iff the polytope is *open* in the direction of the objective function.

Note: The *optimal solution* for a bounded LP lies on a *vertex* of the polytope.



Satisfiability as Linear Programming

Our goal is to use LP to check the satisfiability of *sets of linear \mathcal{T}_{RA} -literals*.

Step 1: Convert equalities to inequalities.

- A linear \mathcal{T}_{RA} -equality can be written to have the form $\mathbf{a}^T \mathbf{x} = \mathbf{b}$.
- We rewrite this further as $\mathbf{a}^T \mathbf{x} \geq \mathbf{b}$ and $\mathbf{a}^T \mathbf{x} \leq \mathbf{b}$.
- And finally to $-\mathbf{a}^T \mathbf{x} \leq -\mathbf{b}$ and $\mathbf{a}^T \mathbf{x} \leq \mathbf{b}$.

Step 2: Handle inequalities.

- A \mathcal{T}_{RA} -literal of the form $\mathbf{a}^T \mathbf{x} \leq \mathbf{b}$ is already in the desired form.
- A \mathcal{T}_{RA} -literal of the form $\neg(\mathbf{a}^T \mathbf{x} \leq \mathbf{b})$ is transformed as follows:

$$\neg(\mathbf{a}^T \mathbf{x} \leq \mathbf{b}) \rightarrow (\mathbf{a}^T \mathbf{x} > \mathbf{b}) \rightarrow (-\mathbf{a}^T \mathbf{x} < -\mathbf{b}) \rightarrow (-\mathbf{a}^T \mathbf{x} + y \leq -\mathbf{b}), (y > 0)$$

where y is a fresh variable used for all negated inequalities.

Example: $\neg(2x_1 - x_2 \leq 3)$ rewrites to $-2x_1 + x_2 + y \leq -3, y > 0$

- If there are no negated inequalities, add the inequality $y \leq 1$, where y is a fresh variable.

Satisfiability as Linear Programming [2]

- In either case, we end up with a set of the form $\mathbf{a}^T \mathbf{x} \leq \mathbf{b} \cup \{y > 0\}$

Step 3: Check the satisfiability of $\mathbf{a}^T \mathbf{x} \leq \mathbf{b} \cup \{y > 0\}$.

Encode it as LP: maximize y subject to $\mathbf{a}^T \mathbf{x} \leq \mathbf{b}$.

The final system is *satisfiable* iff the *optimal value* for y is *positive*.

LP Solving Methods and Standard Form

- *Simplex* (Dantzig, 1947) — exponential time $\mathcal{O}(2^n)$, but very efficient in practice
- *Ellipsoid* (Khachiyan, 1979) — polynomial time $\mathcal{O}(n^6)$
- *Projective* (Karmarkar, 1984) — polynomial time $\mathcal{O}(n^{3.5})$

Any LP can be transformed to *standard form*:

$$\begin{aligned} & \text{maximize } \sum_{j=1}^n c_j x_j \\ & \text{such that } \sum_{j=1}^m a_{ij} x_j \leq b_i \text{ for } i = 1, \dots, m \\ & \quad x_j \geq 0 \text{ for } j = 1, \dots, n \end{aligned}$$

Example: Next, we are going to use the following running example LP:

LP Solving Methods and Standard Form [2]

$$\begin{array}{ll}\text{maximize} & 5x_1 + 4x_2 + 3x_3 \\ \text{such that} & \begin{cases} 2x_1 + 3x_2 + x_3 \leq 5 \\ 4x_1 + x_2 + 2x_3 \leq 11 \\ 3x_1 + 4x_2 + 2x_3 \leq 8 \\ x_1, x_2, x_3 \geq 0 \end{cases}\end{array}$$

Slack Variables

- Observe the first inequality: $2x_1 + 3x_2 + x_3 \leq 5$
- Define a *new variable* to represent the *slack*:

$$x_4 = 5 - 2x_1 - 3x_2 - x_3, \quad x_4 \geq 0$$

- Do this for each constraint, so that everything becomes *equalities*.
- Define a new variable to represent the *objective value*: $z = 5x_1 + 4x_2 + 3x_3$

$$\begin{array}{ll} \max 5x_1 + 4x_2 + 3x_3 & \\ \text{s.t. } \begin{cases} 2x_1 + 3x_2 + x_3 \leq 5 \\ 4x_1 + x_2 + 2x_3 \leq 11 \\ 3x_1 + 4x_2 + 2x_3 \leq 8 \\ x_1, x_2, x_3 \geq 0 \end{cases} & \longrightarrow \begin{array}{ll} \max z & \\ \text{s.t. } \begin{cases} x_4 = 5 - 2x_1 - 3x_2 - x_3 \\ x_5 = 11 - 4x_1 - x_2 - 2x_3 \\ x_6 = 8 - 3x_1 - 4x_2 - 2x_3 \\ z = 5x_1 + 4x_2 + 3x_3 \\ x_1, x_2, x_3, x_4, x_5, x_6 \geq 0 \end{cases} \end{array} \end{array}$$

Note: Optimal solution remains optimal for the new problem.

The Simplex Strategy

- Start with a feasible solution.
 - For our example, assign 0 to all variables.
 $x_1 \mapsto 0, x_2 \mapsto 0, x_3 \mapsto 0$
 - Assign the introduced variables their computed values.
 $x_4 \mapsto 5, x_5 \mapsto 11, x_6 \mapsto 8, z \mapsto 0$
- Iteratively improve the objective value.
 - Go from x to x' only if $z(x) \leq z(x')$.

What can we improve here?

One option is to make x_1 larger, leaving x_2 and x_3 unchanged:

- $x_1 = 1 \rightarrow x_4 = 3, x_5 = 7, x_6 = 1, z = 5$ ✓
- $x_1 = 2 \rightarrow x_4 = 1, x_5 = 3, x_6 = 2, z = 10$ ✓
- $x_1 = 3 \rightarrow x_4 = -1, \dots$ ✗ *no longer feasible!*

$$\begin{cases} x_4 = 5 - 2x_1 - 3x_2 - x_3 \\ x_5 = 11 - 4x_1 - x_2 - 2x_3 \\ x_6 = 8 - 3x_1 - 4x_2 - 2x_3 \\ z = 5x_1 + 4x_2 + 3x_3 \end{cases}$$

The Simplex Strategy [2]

We can't increase x_1 *too much*. Let's increase it as much as possible, *without compromising feasibility*.

$$\begin{array}{l} x_1 \mapsto 0, x_2 \mapsto 0, x_3 \mapsto 0 \\ \left\{ \begin{array}{l} x_4 = 5 - 2x_1 - 3x_2 - x_3 \\ x_5 = 11 - 4x_1 - x_2 - 2x_3 \\ x_6 = 8 - 3x_1 - 4x_2 - 2x_3 \\ z = 5x_1 + 4x_2 + 3x_3 \end{array} \right. \rightarrow \left\{ \begin{array}{l} x_1 \leq \frac{5}{2} \\ x_1 \leq \frac{11}{4} \\ x_1 \leq \frac{8}{3} \end{array} \right. \end{array}$$

Select the *tightest bound*, $x_1 \leq \frac{5}{2}$.

- New assignment: $x_1 \mapsto \frac{5}{2}, x_2 \mapsto x_3 \mapsto x_4 \mapsto 0, x_5 \mapsto 1, x_6 \mapsto \frac{1}{2}, z \mapsto \frac{25}{2}$.

Now *pivot*: since x_1 became positive and x_4 became 0, swap them by isolating x_1 from the equation for x_4 , then eliminating x_1 from all other equations:

The Simplex Strategy [3]

$$\left\{ \begin{array}{l} x_4 = 5 - 2x_1 - 3x_2 - x_3 \\ x_5 = 11 - 4x_1 - x_2 - 2x_3 \\ x_6 = 8 - 3x_1 - 4x_2 - 2x_3 \\ z = 5x_1 + 4x_2 + 3x_3 \end{array} \right. \longrightarrow \left\{ \begin{array}{l} x_1 = \frac{5}{2} - \frac{3}{2}x_2 - \frac{1}{2}x_3 - \frac{1}{2}x_4 \\ x_5 = 1 + 5x_2 + \quad \quad \quad + 2x_4 \\ x_6 = \frac{1}{2} + \frac{1}{2}x_2 - \frac{1}{2}x_3 + \frac{3}{2}x_4 \\ z = \frac{25}{2} - \frac{7}{2}x_2 + \frac{1}{2}x_3 - \frac{5}{2}x_4 \end{array} \right.$$

The Simplex Strategy [4]

How can we improve z further?

- **Option 1:** decrease x_2 or x_4 , but we can't since $x_2, x_4 \geq 0$.
- **Option 2:** increase x_3 . *By how much?*

x_3 's bounds: $x_3 \leq 5, x_3 \leq \infty, x_3 \leq 1$.

We increase x_3 to its tightest bound 1.

- New assignment: $x_1 \mapsto 2, x_2 \mapsto 0, x_3 \mapsto 1, x_4 \mapsto 0, x_5 \mapsto 0, x_6 \mapsto 0$.
- This gives $z = 13$, which is again an improvement.

As before, we switch x_6 and x_3 , and *eliminate* x_3 from the right-hand-side:

$$\begin{cases} x_1 = \frac{5}{2} - \frac{3}{2}x_2 - \frac{1}{2}x_3 - \frac{1}{2}x_4 \\ x_5 = 1 + 5x_2 + \quad \quad + 2x_4 \\ x_6 = \frac{1}{2} + \frac{1}{2}x_2 - \frac{1}{2}x_3 + \frac{3}{2}x_4 \\ z = \frac{25}{2} - \frac{7}{2}x_2 + \frac{1}{2}x_3 - \frac{5}{2}x_4 \end{cases} \longrightarrow \begin{cases} x_1 = 2 - 2x_2 - 2x_4 + x_6 \\ x_5 = 1 + 5x_2 + 2x_4 \\ x_3 = 1 + x_2 + 3x_4 - 2x_6 \\ z = 13 - 3x_2 - x_4 - x_6 \end{cases}$$

$$x_1 \mapsto \frac{5}{2}, x_2 \mapsto 0, x_3 \mapsto 0, x_4 \mapsto 0$$

$$\begin{cases} x_1 = \frac{5}{2} - \frac{3}{2}x_2 - \frac{1}{2}x_3 - \frac{1}{2}x_4 \\ x_5 = 1 + 5x_2 + \quad \quad + 2x_4 \\ x_6 = \frac{1}{2} + \frac{1}{2}x_2 - \frac{1}{2}x_3 + \frac{3}{2}x_4 \\ z = \frac{25}{2} - \frac{7}{2}x_2 + \frac{1}{2}x_3 - \frac{5}{2}x_4 \end{cases}$$

The Simplex Strategy [5]

Can we improve z again?

- No, because $x_2, x_4, x_6 \geq 0$, and all *appear with negative signs* in the objective function.

So, we are done, and the optimal value of z is 13.

Optimal solution: $x_1 \mapsto 2, x_2 \mapsto 0, x_3 \mapsto 1$, with $z = 13$.

$$x_1 \mapsto 2, x_2 \mapsto 0, x_3 \mapsto 1,$$

$$x_4 \mapsto 0, x_6 \mapsto 0$$

$$\begin{cases} x_1 = 2 - 2x_2 - 2x_4 + x_6 \\ x_5 = 1 + 5x_2 + 2x_4 \\ x_3 = 1 + x_2 + 3x_4 - 2x_6 \\ z = 13 - 3x_2 - x_4 - x_6 \end{cases}$$

The Simplex Algorithm

1. Introduce *slack variables* x_{n+1}, \dots, x_{n+m} for each constraint.
2. Start with initial *feasible* solution (commonly, $x_j \mapsto 0$ for all original variables).
3. While some coefficients in the objective function are *positive*:
 - Pick a variable x_j with positive coefficient (*entering variable*).
 - Compute the tightest bound: $\min_i \left(\frac{b_i}{a_{ij}} \right)$ for $a_{ij} > 0$ (*leaving variable*).
 - *Pivot*: swap entering/leaving variables in the equation system.
4. When all coefficients are non-positive, the current solution is *optimal*. Stop.
5. Go to 4.

CDCL(\mathcal{T})

CDCL(\mathcal{T}) Architecture

$$\text{CDCL}(\mathcal{T}) = \text{CDCL}(X) + \mathcal{T}\text{-solver}$$

CDCL(X):

- Very *similar to a SAT solver*, enumerates Boolean models.
- Not allowed: pure literal rule (and other SAT specific heuristics).
- Required: incremental addition of clauses.
- Desirable: partial model detection.

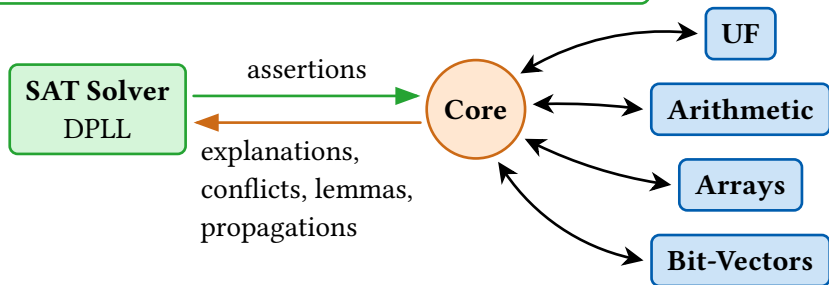
\mathcal{T} -solver:

- Checks the \mathcal{T} -satisfiability of conjunctions of literals.
- Computes *theory propagations*.
- Produces *explanations* of \mathcal{T} -unsatisfiability/propagation.
- Must be *incremental* and *backtrackable*.

Typical SMT Solver Architecture

SAT Solver:

- Only sees *Boolean skeleton* of a problem.
- Builds *partial model* by assigning truth values to literals
- Sends these literals to the core as *assertions*



Core:

- Sends each assertion to the appropriate theory
- Sends deduced literals to other theories/SAT solver
- Handles *theory combination*

Theory Solvers:

- Check \mathcal{T} -satisfiability of sets of theory literals
- Incremental
- Backtrackable
- Conflict generation
- Theory propagation

Theory Propagation

The \mathcal{T} -solver does more than just check consistency — it actively *propagates* information back to the SAT solver:

- **Conflict detection:** If the current set of theory literals is \mathcal{T} -unsatisfiable, report a *conflict clause* (a subset of literals explaining the inconsistency).
- **Theory propagation:** If a literal ℓ is \mathcal{T} -entailed by the current assertions, *propagate* ℓ to the SAT solver (avoiding unnecessary case splits).
- **Lemma learning:** The \mathcal{T} -solver may derive useful *theory lemmas* (clauses that are \mathcal{T} -valid) and add them to the clause database.

Example: If the \mathcal{T}_{RA} -solver sees $x \leq 3$ and $x \geq 5$, it immediately reports a conflict without waiting for the SAT solver to discover the contradiction.

Why incremental and backtrackable? The SAT solver frequently backtracks, so the \mathcal{T} -solver must efficiently *undo* assertions. Typical implementations use a *stack-based* approach: push assertions on decisions, pop on backtrack.

CDCL(\mathcal{T}) Example

Consider the formula $\varphi = (x \doteq y \vee y \doteq z) \wedge (f(x) \not\doteq f(y) \vee y \doteq z) \wedge (x \doteq y \vee f(y) \not\doteq f(z))$ in \mathcal{T}_{EUF} .

Step 1: Boolean abstraction. Replace each theory atom with a Boolean variable:

- $p_1 := (x \doteq y)$, $p_2 := (y \doteq z)$, $p_3 := (f(x) \doteq f(y))$, $p_4 := (f(y) \doteq f(z))$
- Boolean skeleton: $(p_1 \vee p_2) \wedge (\neg p_3 \vee p_2) \wedge (p_1 \vee \neg p_4)$

Step 2: CDCL decides $p_1 = \text{true}$, $p_2 = \text{false}$.

- SAT solver propagates: $\neg p_3$ (from clause 2) and $\neg p_4$ (from clause 3, since $p_2 = \text{false}$).

Step 3: \mathcal{T} -solver checks: $\{x \doteq y, \neg(y \doteq z), \neg(f(x) \doteq f(y)), \neg(f(y) \doteq f(z))\}$.

- From $x \doteq y$, by congruence: $f(x) \doteq f(y)$. But we have $\neg(f(x) \doteq f(y))$ – **conflict!**
- Conflict clause: $\neg(x \doteq y) \vee f(x) \doteq f(y)$, i.e. $\neg p_1 \vee p_3$.

Step 4: CDCL learns the clause $\neg p_1 \vee p_3$, backtracks, and eventually finds SAT with $p_2 = \text{true}$.

Combining Theories

Motivation: Mixed Formulas

In practice, formulas often involve *multiple* theories simultaneously.

Example:

$$f(x) - f(y) \geq 1 \wedge (x \doteq y)$$

This mixes \mathcal{T}_{EUF} (uninterpreted f , equality \doteq) and \mathcal{T}_{RA} (arithmetic $-, \geq, 1$).

Example:

$$\text{read}(\text{write}(a, i, v), j) + 1 \leq x \wedge i \doteq j$$

This mixes \mathcal{T}_{AX} (arrays) and \mathcal{T}_{RA} (arithmetic).

A single-theory solver cannot handle such formulas. We need a *combination method* that orchestrates multiple theory solvers.

The Nelson-Oppen Method

The *Nelson-Oppen* (N-O) method combines decision procedures for *signature-disjoint*, *stably infinite* theories.

Definition 38: Two theories \mathcal{T}_1 and \mathcal{T}_2 are *signature-disjoint* if $\Sigma_1^F \cap \Sigma_2^F = \{=\}$ — the only shared symbol is equality.

Definition 39: A theory \mathcal{T} is *stably infinite* if every \mathcal{T} -satisfiable quantifier-free formula is satisfiable in a model with an *infinite* domain.

Note: Most commonly used SMT theories (\mathcal{T}_{EUF} , \mathcal{T}_{RA} , \mathcal{T}_{IA} , \mathcal{T}_{AX}) are stably infinite and pairwise signature-disjoint (when restricted to their respective sorts).

Nelson-Oppen: Purification

Step 1: Purification. Separate a mixed formula into *pure* conjuncts, one per theory.

Method: For any term t from theory \mathcal{T}_i that appears as an argument in a literal of theory \mathcal{T}_j (with $i \neq j$), introduce a *fresh shared variable* v , add $v \doteq t$ to \mathcal{T}_i 's literals, and replace t by v in \mathcal{T}_j 's literals.

Example: Purify $f(x) - f(y) \geq 1 \wedge x \doteq y$:

Introduce $v_1 \doteq f(x)$ and $v_2 \doteq f(y)$:

- $\mathcal{T}_{\text{EUF}}: \{v_1 \doteq f(x), v_2 \doteq f(y), x \doteq y\}$
- $\mathcal{T}_{\text{RA}}: \{v_1 - v_2 \geq 1\}$
- *Shared variables:* v_1, v_2 (appear in both “pure” sets)

Nelson-Oppen: Equality Propagation

Step 2: Equality propagation. The two solvers exchange *equalities and disequalities* between shared variables until a *fixed point* or a *conflict* is reached.

1. Check each pure set with its own \mathcal{T} -solver.
2. If either solver reports UNSAT \Rightarrow the combined formula is UNSAT.
3. If the \mathcal{T}_i -solver deduces a new equality $v \doteq w$ between shared variables, *propagate* it to the other solver.
4. Repeat until no new equalities are deduced (*fixed point*).
5. If both solvers report SAT at the fixed point \Rightarrow the combined formula is SAT.

Key insight: For *convex* theories (including \mathcal{T}_{EUF} and \mathcal{T}_{RA}), the method only needs to propagate *equalities* — no disjunctions. This makes the procedure *deterministic* and efficient.

Definition 40: A theory \mathcal{T} is *convex* if whenever $\mathcal{T} \models (\ell_1 \wedge \dots \wedge \ell_n) \rightarrow (x_1 \doteq y_1 \vee \dots \vee x_k \doteq y_k)$, then $\mathcal{T} \models (\ell_1 \wedge \dots \wedge \ell_n) \rightarrow (x_i \doteq y_i)$ for some i .

Nelson-Oppen: Worked Example

Formula: $f(x) - f(y) \geq 1 \wedge x \doteq y$

After purification with shared variables v_1, v_2 :

- \mathcal{T}_{EUF} : $\{v_1 \doteq f(x), v_2 \doteq f(y), x \doteq y\}$
- \mathcal{T}_{RA} : $\{v_1 - v_2 \geq 1\}$

Round 1:

- \mathcal{T}_{EUF} -solver: From $x \doteq y$ and congruence, deduces $f(x) \doteq f(y)$, hence $v_1 \doteq v_2$.
- *Propagate* $v_1 \doteq v_2$ to \mathcal{T}_{RA} .

Round 2:

- \mathcal{T}_{RA} -solver: Now has $\{v_1 - v_2 \geq 1, v_1 \doteq v_2\}$.
 - $v_1 \doteq v_2$ implies $v_1 - v_2 = 0$, contradicting $v_1 - v_2 \geq 1$.
 - **UNSAT!**

Conclusion: The original mixed formula is unsatisfiable.

SMT-LIB and Z3

SMT-LIB: The Standard Language

SMT-LIB (v2) is a standardized *input language* for SMT solvers. It defines:

- A set of *logics* (e.g. QF_UF, QF_LIA, QF_LRA, QF_AUFLIA, QF_BV, ALL) specifying which theories and quantifiers are allowed.
- A *command language* for interacting with solvers.
- Standard *theory declarations* shared across all solvers.

Core commands:

Command	Description
(set-logic QF_LIA)	Declare the logic
(declare-sort S 0)	Declare an uninterpreted sort with arity 0
(declare-fun f (Int Int) Bool)	Declare a function symbol with rank
(define-fun g ((x Int)) Int ...)	Define a function (macro)
(assert (< x 5))	Assert a formula
(check-sat)	Check satisfiability

SMT-LIB: The Standard Language [2]

Command	Description
(get-model)	Retrieve a satisfying assignment (if SAT)
(push 1) / (pop 1)	Save/restore assertion stack
(exit)	Terminate the session

SMT-LIB: QF_UF Example

Encoding the EUF problem: $a \doteq b \wedge f(a) \doteq b \wedge \neg(g(a) \doteq g(f(a)))$.

```
(set-logic QF_UF)
(declare-sort U 0)

(declare-fun a () U)
(declare-fun b () U)
(declare-fun f (U) U)
(declare-fun g (U) U)

(assert (= a b))
(assert (= (f a) b))
(assert (not (= (g a) (g (f a)))))

(check-sat)    ; Expected: unsat
(exit)
```

Note: In SMT-LIB, = is the built-in equality. There is no need to declare it. Variables are *declared* as zero-arity functions: (declare-fun a () U) means “*a* is a *constant* of sort *U*”.

SMT-LIB: QF_LIA Example

Is there an integer solution to $x + 2y \geq 5$, $x - y \leq 1$, $x \geq 0$, $y \geq 0$?

```
(set-logic QF_LIA)
```

```
(declare-fun x () Int)
```

```
(declare-fun y () Int)
```

```
(assert (>= (+ x (* 2 y)) 5))
```

```
(assert (<= (- x y) 1))
```

```
(assert (>= x 0))
```

```
(assert (>= y 0))
```

```
(check-sat)      ; Expected: sat
```

```
(get-model)      ; e.g., x = 1, y = 2
```

```
(exit)
```

SMT-LIB: QF_LIA Example [2]

Logic naming convention: QF_ = quantifier-free. L = linear, N = non-linear. I = integers, R = reals. A = arrays. UF = uninterpreted functions. BV = bit-vectors. So QF_AUFLIA = quantifier-free arrays + UF + linear integer arithmetic.

Z3: An SMT Solver

Z3 (Microsoft Research) is one of the most widely used SMT solvers.

- Supports *all major theories*: EUF, LIA, LRA, arrays, bit-vectors, strings, datatypes, ...
- Accepts *SMT-LIB* input and also has *Python*, *C/C++*, and *Java* APIs.
- Used in: Dafny, Boogie, KLEE, Rosette, angr, many other FM tools.

Example: Run the QF_LIA example with Z3 from the command line:

```
z3 example.smt2
```

Output:

```
sat
(model
  (define-fun x () Int 1)
  (define-fun y () Int 2))
```

Z3 Python API: z3py

The *z3py* library provides a Pythonic interface to Z3:

```
from z3 import *

x, y = Ints('x y')
s = Solver()
s.add(x + 2*y >= 5)
s.add(x - y <= 1)
s.add(x >= 0, y >= 0)

if s.check() == sat:
    m = s.model()
    print(f"x = {m[x]}, y = {m[y]}")
```

Common z3py types and constructors:

- `Bool('b')`, `Int('x')`, `Real('r')` — declare sorted variables
- `Function('f', IntSort(), IntSort())` — uninterpreted function $f : \text{Int} \rightarrow \text{Int}$
- `Array('a', IntSort(), IntSort())` — array variable
- `Solver()`, `.add(...)`, `.check()`, `.model()` — solver interaction

Z3 Python API: z3py [2]

- `And(...)`, `Or(...)`, `Not(...)`, `Implies(a, b)` — logical connectives

Z3 Practical Examples

Array swap verification:

```
from z3 import *
a = Array('a', IntSort(), IntSort())
i, j = Ints('i j')
# swap a[i] and a[j]
b = Store(Store(a, i, Select(a, j)), j, Select(a, i))
# verify that b[i] == a[j] and b[j] == a[i]
s = Solver()
s.add(Not(And(Select(b, i) == Select(a, j),
              Select(b, j) == Select(a, i))))
print(s.check()) # unsat => swap is correct
```

Simple scheduling:

```
from z3 import *
A, B, C = Ints('A B C') # start times
s = Solver()
s.add(A >= 0, B >= 0, C >= 0) # non-negative start
s.add(A + 3 <= B)             # A finishes before B starts
```

Z3 Practical Examples [2]

```
s.add(B + 2 <= C)           # B finishes before C starts
s.add(C + 1 <= 8)           # C finishes by deadline 8
if s.check() == sat:
    print(s.model())        # e.g., A=0, B=3, C=5
```

Advanced Applications of SMT

Symbolic Execution

Symbolic execution explores program paths using *symbolic* rather than concrete inputs, building *path conditions* as SMT formulas.

Example:

```
def abs(x: int) -> int:  
    if x >= 0: return x  
    else: return -x
```

- Path 1: condition $x \geq 0$, result = x .
- Path 2: condition $x < 0$, result = $-x$.
- To find an input where $\text{abs}(x) < 0$: query SMT for $(x \geq 0 \wedge x < 0) \vee (x < 0 \wedge -x < 0)$ – UNSAT!

Tools: KLEE (LLVM), SAGE (Microsoft), angr (binary analysis). All rely on SMT solvers (typically Z3) as the constraint-solving backend.

CEGAR and CEGIS

Counterexample-Guided Abstraction Refinement (CEGAR):

1. Start with a *coarse* abstraction of the program.
2. Check if the abstraction satisfies the property (using an SMT solver).
3. If a counterexample is found, check if it is *spurious* (infeasible in the concrete program).
4. If spurious, *refine* the abstraction and repeat.

Counterexample-Guided Inductive Synthesis (CEGIS):

1. Propose a candidate program (from a template).
2. Check if it satisfies the specification for *all* inputs (using an SMT solver).
3. If a counterexample input is found, add it to the test suite and re-synthesize.

Common pattern: both CEGAR and CEGIS use SMT solvers in a *verify-refine loop*. The solver alternates between finding counterexamples and checking candidates.

Syntax-Guided Synthesis (SyGuS)

Definition 41 (SyGuS Problem): Given a *specification* $\varphi(x, f(x))$ and a *grammar* G defining the space of candidate expressions, find a function f expressible in G such that $\forall x. \varphi(x, f(x))$.

Example: **Synthesize** $\max(x, y)$:

- Spec: $f(x, y) \geq x \wedge f(x, y) \geq y \wedge (f(x, y) = x \vee f(x, y) = y)$
- Grammar: $f ::= x \mid y \mid \text{ite}(f \geq f, f, f)$
- Solution: $\text{ite}(x \geq y, x, y)$

SyGuS competitions (since 2014) drive solver development. Key solvers: **CVC5** (built-in SyGuS), **EUSolver**, **DryadSynth**.

The big picture: SMT solvers are not just decision procedures — they are the *engine* behind program analysis, synthesis, and verification. Everything from Dafny to KLEE to CVC5's synthesizer relies on efficient SMT solving.

Exercises

Exercise: Theory Satisfiability

Determine whether the following sets of literals are \mathcal{T} -satisfiable or \mathcal{T} -unsatisfiable. If satisfiable, provide a \mathcal{T} -interpretation. If unsatisfiable, show why.

1. In \mathcal{T}_{EUF} : $\{a \doteq b, b \doteq c, f(a) \not\doteq f(c)\}$
2. In \mathcal{T}_{RA} : $\{x + y \leq 3, 2x - y \geq 5, x \leq 1\}$
3. In DL: $\{x - y \leq 2, y - z \leq 3, z - x \leq -6\}$

Exercise: Nelson-Oppen and SMT-LIB

1. Purify the following formula into \mathcal{T}_{EUF} and \mathcal{T}_{RA} components:

$$f(x + 1) \doteq f(y) \wedge x - y \geq 0 \wedge x \leq y$$

Run the Nelson-Oppen equality propagation. Is the formula satisfiable?

2. Encode the following problem in SMT-LIB (QF_LIA): “Find integers x, y, z such that $x + y + z = 15$, $x \geq 1$, $y \geq 1$, $z \geq 1$, and $x \leq y \leq z$.”
3. Write a z3py script to verify that for all integers x : if $x > 0$, then $x + x > x$. (*Hint: show the negation is UNSAT.*)