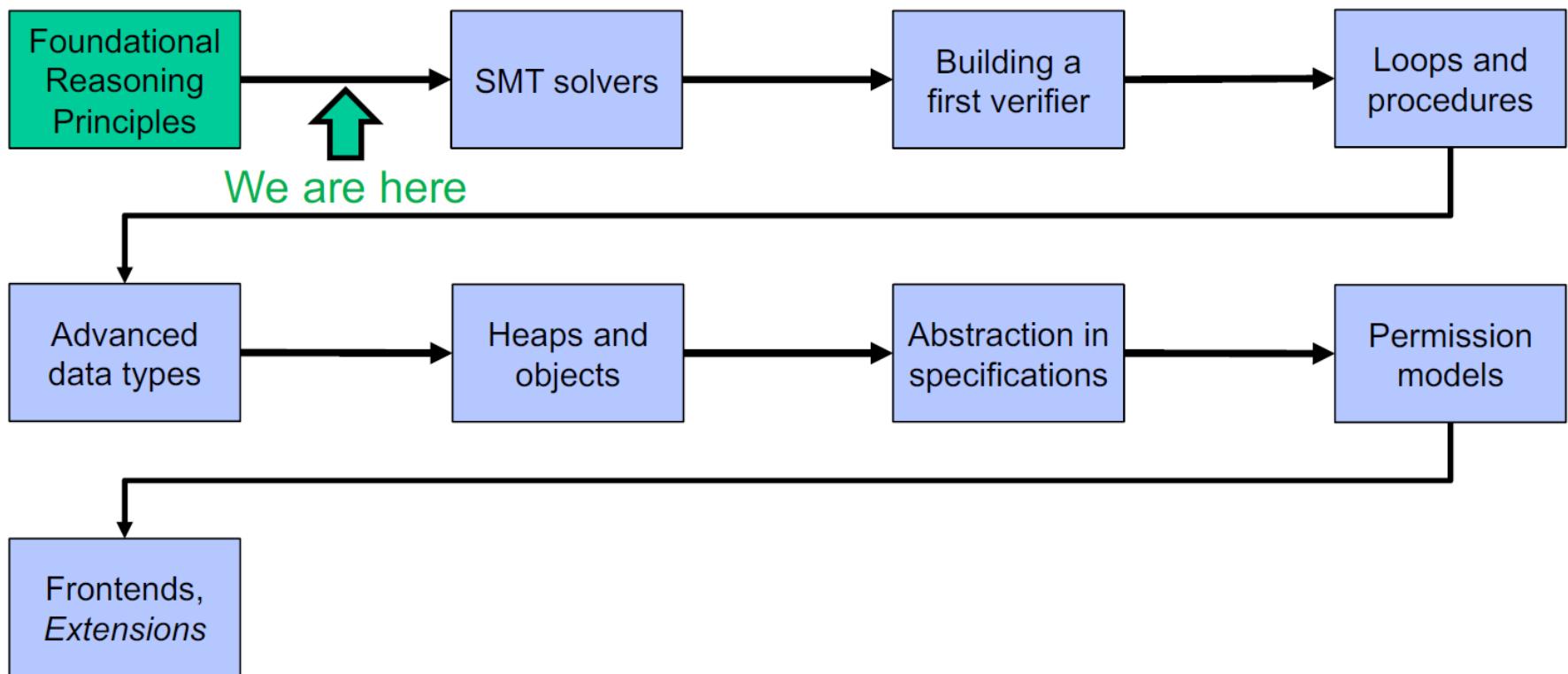


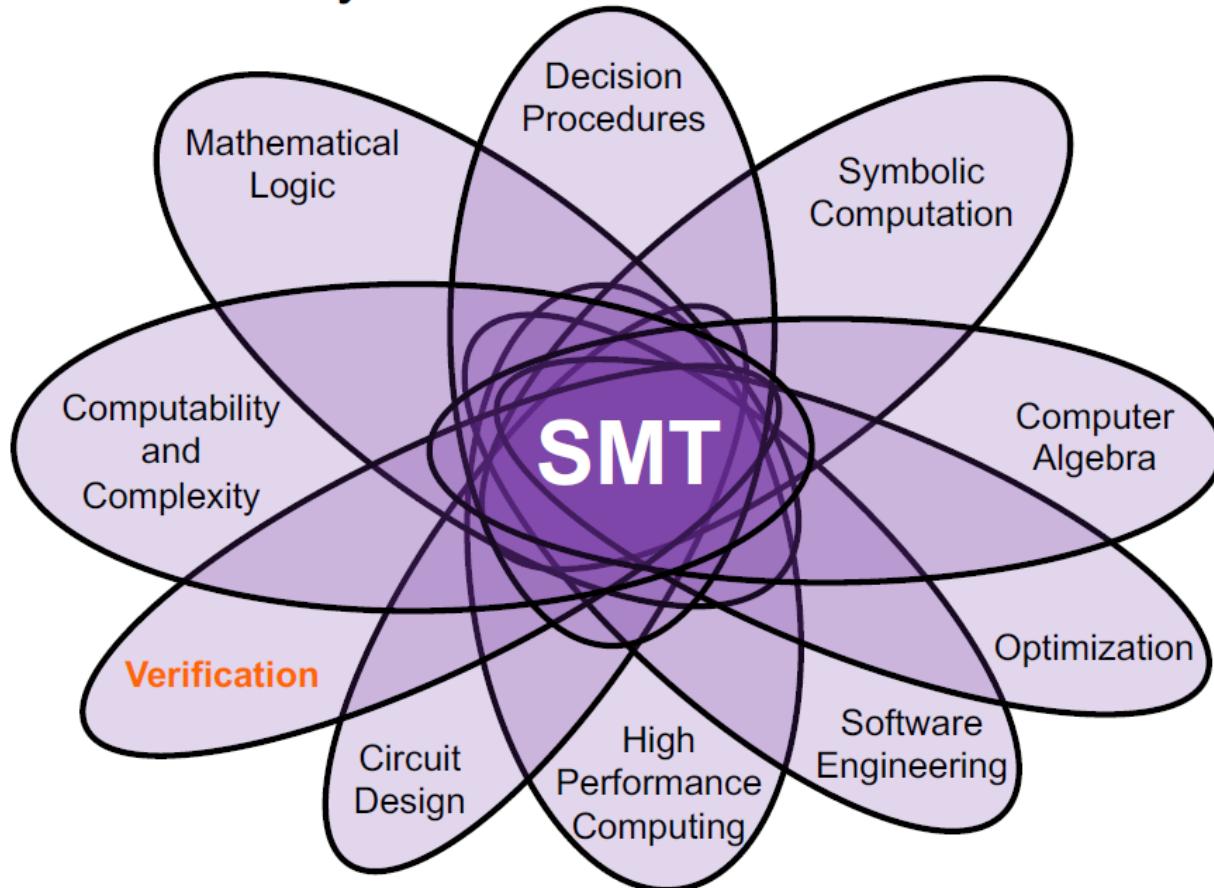
Program Analysis for Software Security

Lecture 3

Tentative course outline



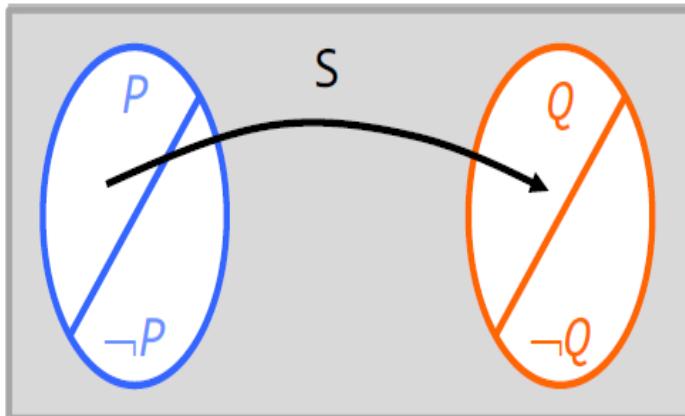
Satisfiability Modulo Theories Solvers



- A foundational topic in theoretical and applied computer science
- Our focus:
effectively applying SMT technology to program verification

But first: Recap

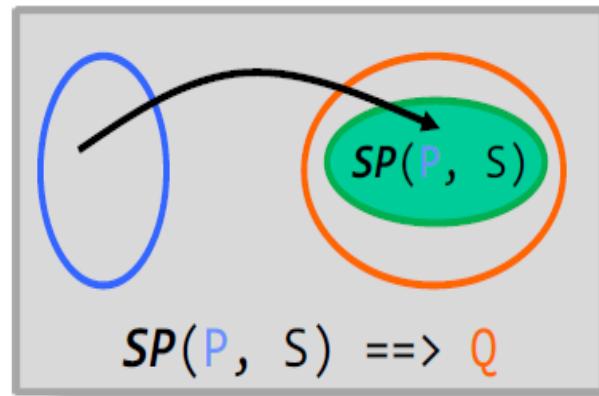
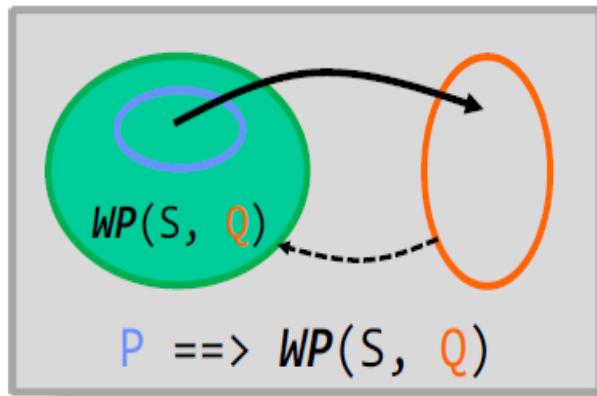
The **Floyd-Hoare triple** $\{ P \} S \{ Q \}$ is **valid** if and only if every execution of S that starts in a state satisfying **precondition** P terminates without an error in a state satisfying **postcondition** Q .



```
method foo(x: Int)
  returns (r: Int)
  requires x > 0
  ensures r > y
{
  // S
  var y: Int := 7
  r := x + y
}
```



Recap: Weakest Pre & Strongest Post



S	$WP(S, Q)$ (total correctness)	$SP(P, S)$ (partial correctness: accepts errors/divergence)
<code>var x</code>	<code>forall x :: Q</code>	<code>exists x :: Q</code>
<code>x := a</code>	<code>Q[x / a]</code>	<code>exists x0 :: P[x / x0] && x == a[x / x0]</code>
<code>assert R</code>	<code>R && Q</code>	<code>P && R</code>
<code>assume R</code>	<code>R ==> Q</code>	<code>P && R</code>
<code>S1; S2</code>	<code>WP(S1, WP(S2, Q))</code>	<code>SP(SP(P, S1), S2)</code>
<code>S1 [] S2</code>	<code>WP(S1, Q) && WP(S2, Q)</code>	<code>SP(P, S1) SP(P, S2)</code>

Automating Program Verification

Mains steps of a tool for checking that $\{ P \} S \{ Q \}$ is valid:

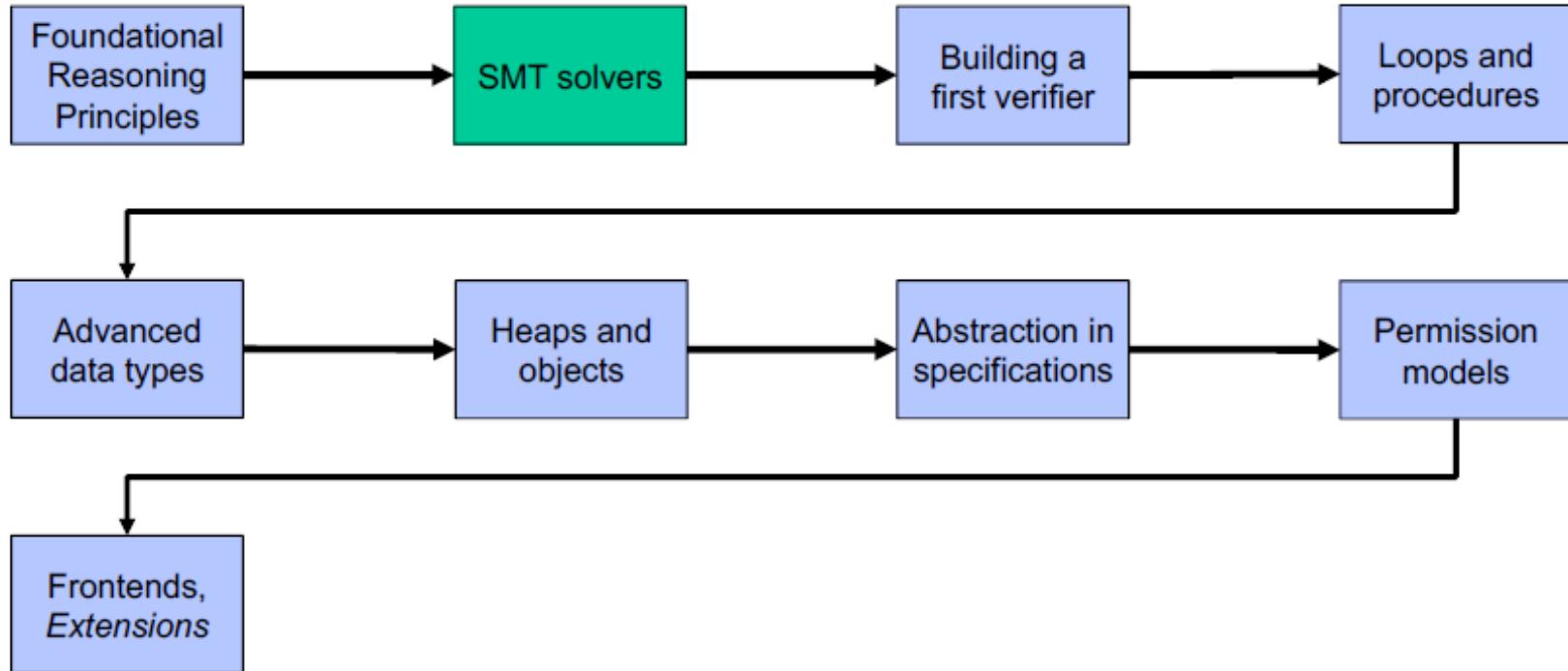
1. Compute $WP(S, Q)$ → last lecture
2. Check whether $P ==> WP(S, Q)$ is valid → delegate to SMT solver

Alternative approach

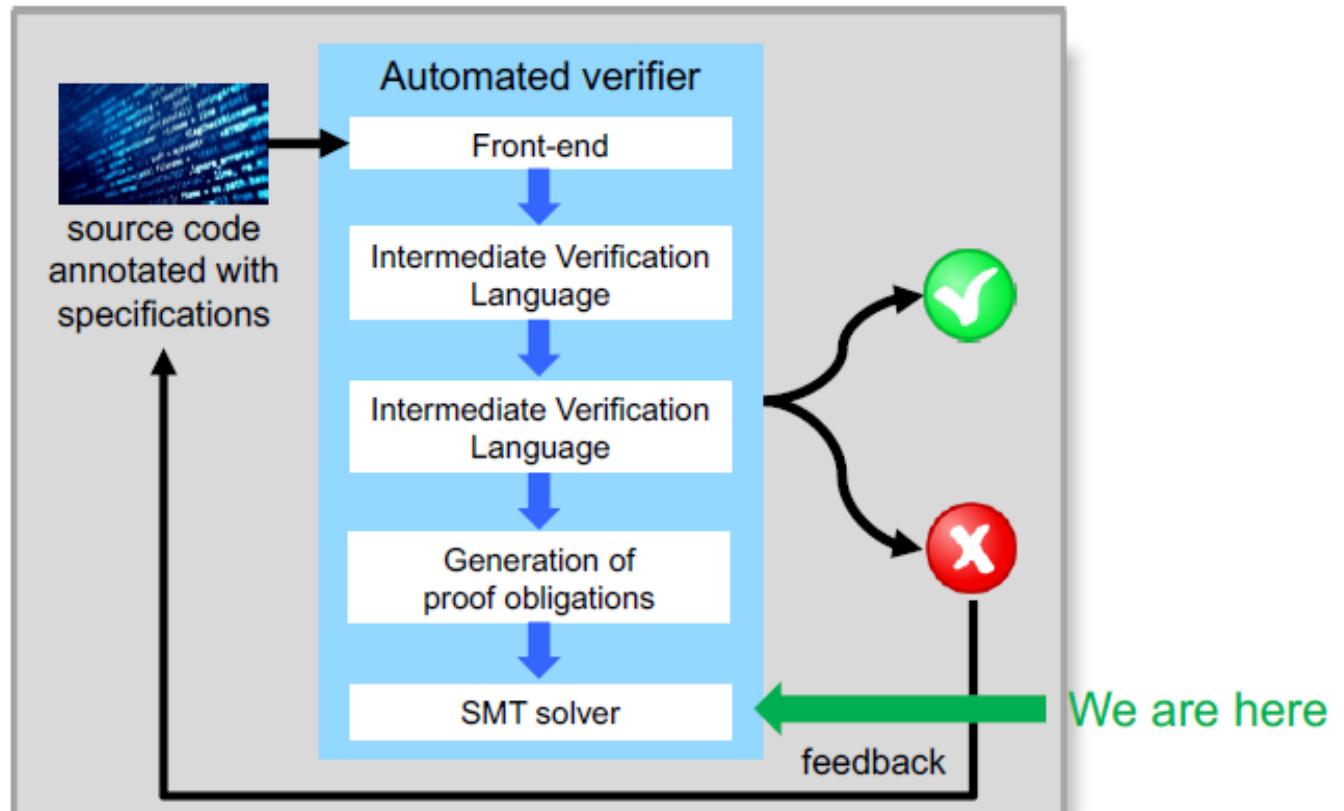
Mains steps of a tool for checking that $\{ P \} S \{ Q \}$ is valid:

1. Compute $SP(P, S)$ and $SAFE(P, S)$
2. Check whether $SP(P, S) ==> Q$ is valid
and $SAFE(P, S)$ is valid

Tentative course outline



Roadmap



Overview

1. Propositional logic and SAT solvers
2. Using Z3 as a SAT solver
3. First-order logic and SMT solvers
4. Using Z3 as an SMT solver

Propositional Logic

X: Boolean variable in **Var**

Syntax

$F ::= \text{false} \mid \text{true} \mid x \mid \neg F \mid F \wedge F \mid F \vee F \mid F \Rightarrow F \mid F \Leftrightarrow F$

Interpretation $\mathfrak{I}: \text{Var} \rightarrow \{\text{true}, \text{false}\}$

Satisfaction relation

$\mathfrak{I} \models \text{true}$ iff always
 $\mathfrak{I} \models X$ iff $\mathfrak{I}(X) = \text{true}$
 $\mathfrak{I} \models \neg F$ iff not $\mathfrak{I} \models F$
 $\mathfrak{I} \models F \wedge G$ iff $\mathfrak{I} \models F$ and $\mathfrak{I} \models G$

\mathfrak{I} is a **model** of F iff $\mathfrak{I} \models F$

$\mathfrak{I} ::= [X = \text{false}, Y = \text{true}]$

$\mathfrak{I} \models \neg X \vee Y$

$\mathfrak{I} \models X \Rightarrow Y$

$\mathfrak{I} \models (\neg X \vee Y) \Leftrightarrow (X \Rightarrow Y)$

Satisfiability & Validity

- \mathbf{F} is **satisfiable** iff \mathbf{F} has **some model**

$$(X \Rightarrow Y) \Rightarrow Y$$

Models: $[X = \text{true}, Y = \text{true}]$, $[X = \text{false}, Y = \text{true}]$, $[X = \text{true}, Y = \text{false}]$

- \mathbf{F} is **unsatisfiable** iff \mathbf{F} has **no model**

$$X \wedge \neg Y \wedge (X \Rightarrow Y)$$

- \mathbf{F} is **valid** iff **every interpretation** is a model of \mathbf{F}
 $(\neg \mathbf{F}$ is unsatisfiable)

$$X \wedge (X \Rightarrow Y) \Rightarrow Y$$

- \mathbf{F} is **not valid** iff **some interpretation is not a model** of \mathbf{F}
 $(\neg \mathbf{F}$ is satisfiable)

$$X \wedge (X \Rightarrow Y) \Leftrightarrow Y$$

Model of $\neg \mathbf{F}$: $[X = \text{false}, Y = \text{true}]$

The Satisfiability Problem

- A formula is **satisfiable** if it has a model

Satisfiability Problem (SAT):

Given a propositional logic formula,
decide whether it is satisfiable.

- If yes, provide a model as a **witness**

$$\begin{aligned} & (X \vee Y \vee \neg Z) \\ & \wedge (U \vee \neg Y) \\ & \wedge (\neg X \vee \neg Z \vee U \vee V) \end{aligned}$$

```
J ::= [  
  U = false  
  V = false  
  X = true  
  Y = false  
  Z = false  
 ]
```

Complexity of SAT

- For formulas in conjunctive normal form (CNF), SAT is the classical NP-complete problem

- Many difficult problems can be efficiently encoded
 - Every known algorithm is exponential in the formula's size

$$\bigwedge_i \bigvee_j C_{i,j} \text{ where } C_{i,j} \in \{x_{i,j}, \neg x_{i,j}\}$$

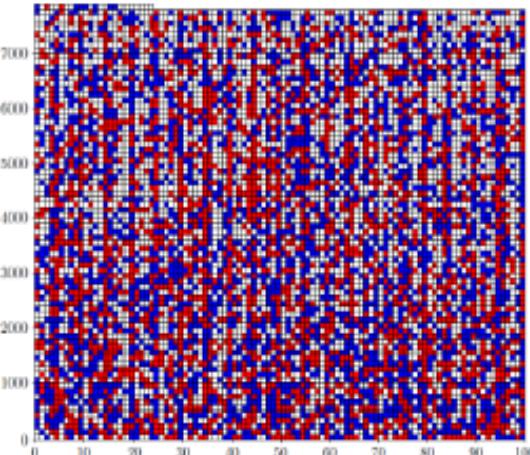
- Modern SAT solvers are extremely efficient in practice

- Scale to formulas with millions of variables
 - May still perform poorly on certain formulas

Example: Boolean Pythagorean Triples

- **BPT:** a triple of natural numbers $1 \leq a \leq b \leq c$ with $a^2 + b^2 = c^2$
- Question: Can we color **all natural numbers** with just two colors such that no BPT is monochromatic?
- Answer: No! The set $\{1, \dots, 7825\}$ always contains a monochromatic BPT

- This was **first proven using a SAT solver**
 - number of combinations: 2^{7825}
 - “the largest math proof ever” (ca. 200 TB)
- **Modern SAT solvers are efficient in practice**



credits: Marijn J.H. Heule, "Everything's Bigger in Texas - The Largest Math Proof Ever", GCAI 2017

Overview

1. Propositional logic and SAT solvers
2. Using Z3 as a SAT solver
3. First-order logic and SMT solvers
4. Using Z3 as an SMT solver

The Z3 Satisfiability Modulo Theories solver

- Developed by Microsoft (under MIT license)
- Building block of many verification tools including Viper
- Various input formats and APIs
 - Z3, [SMTLIB-2](#), C, C++, [Python](#), Java, Rust, OCaml, ...
- For now: Use Z3 as a [SAT solver](#)



A first example (SMTLIB-2)

```
; declare variables
(declare-const X Bool)
(declare-const Y Bool)
(declare-const Z Bool)

; define formula  $(X \Rightarrow Y \Rightarrow Z) \wedge X$ 
(assert ( $\Rightarrow$  X Y Z))
(assert X)

(check-sat)

(get-model) ; fails if unsat
```

```
$ z3 01-example.smt2

sat
(model
  (define-fun Z () Bool
    false)
  (define-fun X () Bool
    true)
  (define-fun Y () Bool
    false)
)
```

A first example (Z3Py)

```
from z3 import *

# declare variables
X = Bool('X')
Y = Bool('Y')
Z = Bool('Z')

# define formula F
F = And( Implies(X, Implies(Y, Z)), X)

solve(F) # find a model for F

# find a counterexample for F
solve(Not(F))
```

F is satisfiable, this is a model

```
$ python .\02-example.py
```

```
[Z = False, X = True, Y = False]
```

```
[Z = False, X = False, Y = True]
```

$\neg F$ is satisfiable, this is a model

Proofs with Z3

```
(declare-const x Bool)
(declare-const y Bool)

(echo "De Morgan's law: !(x && y) == (!x || !y)")
(assert
  (= 
    (not (and x y) )
    (or (not x) (not y)))
  )
(check-sat) ; result: sat
```

What does this tell us about De Morgan's law?

Proofs with Z3

```
(declare-const x Bool)
(declare-const y Bool)

(echo "De Morgan's law: !(x && y) == (!x || !y)")
(assert
  (= 
    (not (and x y) )
    (or (not x) (not y)))
  )
)
(check-sat) ; result: sat
```

there is an example for
which the law is true

What does this tell us about De Morgan's law?

Proofs with Z3

```
(declare-const x Bool)
(declare-const y Bool)

(echo "De Morgan's law: !(x && y) == (!x || !y)")
(assert
  (not
    (= 
      (not (and x y) )
      (or (not x) (not y)))
    )
  )
)
(check-sat) ; result: unsat
```

What does this tell us about De Morgan's law?

Proofs with Z3

```
(declare-const x Bool)
(declare-const y Bool)

(echo "De Morgan's law: !(x && y) == (!x || !y)")
(assert
  (not
    (= 
      (not (and x y) )
      (or (not x) (not y))
    )
  )
)
(check-sat) ; result: unsat
```

There is no counterexample

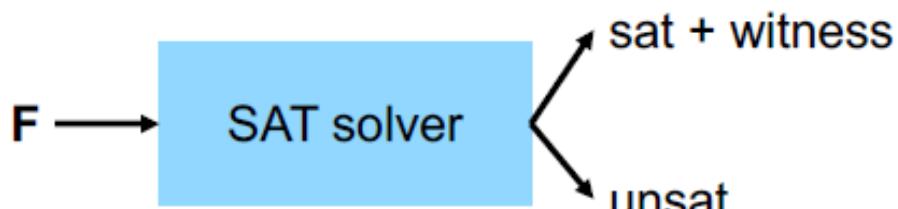
→ the formula is valid

→ De Morgan's law holds

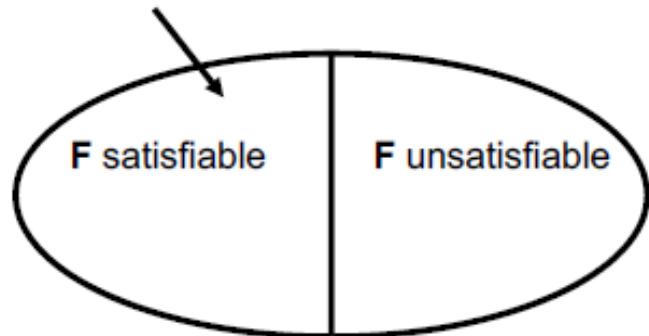
What does this tell us about De Morgan's law?

Using a SAT solver

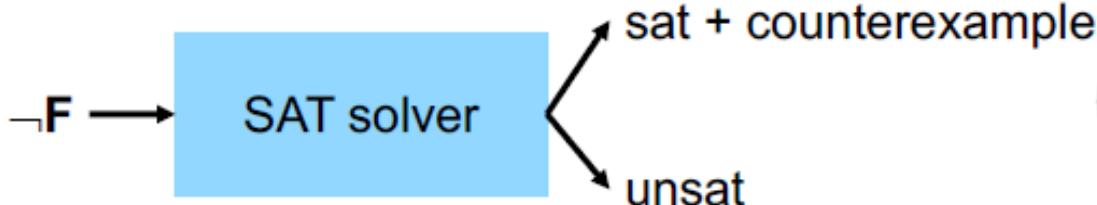
- Is F satisfiable?



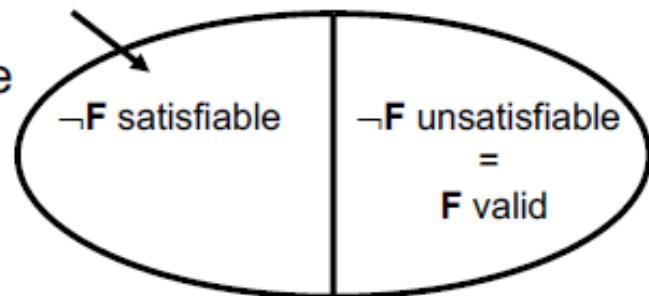
witness: model of F



- Is F valid?



counterexample: model of $\neg F$



Using a SAT Solver for Program Verification

Main steps of a tool for checking that $\{ P \} S \{ Q \}$ is valid:

1. Compute $WP(S, Q)$ → last lecture
2. Check whether entailment $P ==> WP(S, Q)$ is valid → ask SAT solver
 - Check satisfiability of negation: $P \ \&\& \ !WP(S, Q)$
 - unsat → 
 - sat →  model explains why $\{ P \} S \{ Q \}$ is not valid

Using a SAT Solver for Program Verification

```
{ true }
// check that validity of true  $\Rightarrow a \wedge \dots$ 
{ a  $\wedge$  (b  $\wedge$  (true  $\Leftrightarrow$  (a  $\Rightarrow$  b))  $\vee$   $\neg$ b  $\wedge$  (false  $\Leftrightarrow$  (a  $\Rightarrow$  b)))  $\vee$   $\neg$ a  $\wedge$  (true  $\Leftrightarrow$  (a  $\Rightarrow$  b) ) }
if (a) {
{ b  $\wedge$  (true  $\Leftrightarrow$  (a  $\Rightarrow$  b))  $\vee$   $\neg$ b  $\wedge$  (false  $\Leftrightarrow$  (a  $\Rightarrow$  b)) }
  if (b) {
{ true  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
    res := true
{ res  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
  } else {
{ false  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
    res := false
{ res  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
  }
{ res  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
} else {
{ true  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
  res := true
{ res  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
}
{ res  $\Leftrightarrow$  (a  $\Rightarrow$  b) }
```



Propositional logic is not enough

```
{ x == X && y == Y }  
{ y == Y && y - Y == 0 }  
// ... swap X and Y  
{ x == Y && y == X }
```

Entailment to check:

$$(x == X \&\& y == Y) \Rightarrow y == Y \&\& y - Y == 0$$

- Entailment is not in propositional logic
 - Integer-valued variables (x, X, y, Y) and numeric constants (0)
 - Arithmetic operations (-) and comparisons ($==$)
- Logic must support at least the **expressions** appearing in programs
 - It is also useful to support quantifiers (e.g., for array algorithms)
- General framework: **first-order predicate logic (FOL)**

Overview

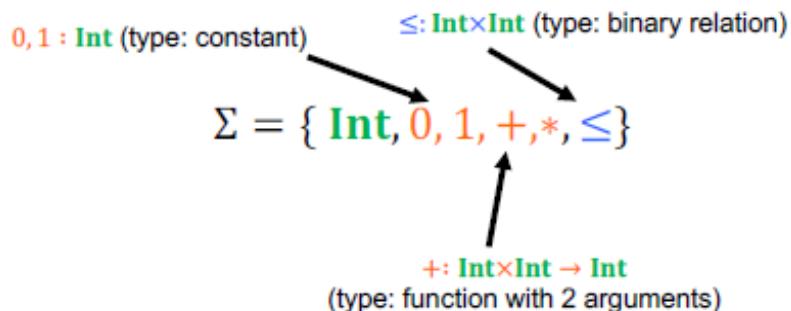
1. Propositional logic and SAT solvers
2. Using Z3 as a SAT solver
3. First-order logic and SMT solvers
4. Using Z3 as an SMT solver

Ingredients of Many-sorted First-order logic (FOL)

1. Sorts	Bool, Int, Real, T
▪ specifies possible types	
2. Typed Variables	x, y, z, ...
3. Typed Function symbols	0, 1.5 +, *, _?_:_
▪ building blocks of terms	0 x ? y - 17 : z*z + 1
4. Typed Relational symbols	< prime R
▪ turn terms into logical propositions	x < 0 prime(y+4) R(x,y,z)
5. Logical symbols	$\wedge \vee \neg \Rightarrow \Leftrightarrow \exists \forall \dots$
6. An equality symbol	=

FOL Formulas

- A **signature Σ** is a set of
 - at least one sort
 - function symbols
 - relational symbols (= does not count)
- A **Σ -formula** is a logical formula over propositions built from **symbols** in Σ



$$\forall x: \text{Int} \exists y: \text{Int} (y = x + 1 \wedge y * y \leq x * x + (1 + 1) * x + 1)$$

Is this Σ -formula satisfiable?

FOL Formulas

Is this Σ -formula satisfiable?

$\Sigma = \{ \text{Int}, 0, 1, +, *, \leq \}$

$$\forall x: \text{Int} \exists y: \text{Int} (y = x + 1 \wedge y * y \leq x * x + (1 + 1) * x + 1)$$

Yes, if the symbols $+$, $*$, $=$ have the canonical meaning

No, if

- 1 actually means 2, or
- $+$ actually means maximum

Satisfiability of Σ -formulas depends on the admissible interpretations of symbols in Σ

determined by Σ -theories

FOL Σ -Interpretations

- A **Σ -structure** \mathfrak{A} assigns
 - a **non-empty domain** (set) $\mathbf{U}^{\mathfrak{A}}$ to each sort \mathbf{U} in Σ
 - a **function** $f^{\mathfrak{A}}$ over domains (respecting types) to each function symbol f in Σ
 - a **relation** $R^{\mathfrak{A}}$ over domains (respecting types) to each relational symbol R in Σ
- A **Σ -assignment** β maps variables x of sort \mathbf{U} to domain elements in $\mathbf{U}^{\mathfrak{A}}$
- A **Σ -interpretation** is a pair $\mathfrak{I} = (\mathfrak{A}, \beta)$
- $\mathfrak{I}(t)$ denotes the domain element obtained by evaluating term t in \mathfrak{I}

$$\Sigma = \{ \text{Int}, \text{one}, \text{plus}, \text{leq} \}$$

$$\mathfrak{A} ::= (\text{Int}^{\mathfrak{A}}, \text{one}^{\mathfrak{A}}, \text{plus}^{\mathfrak{A}}, \text{leq}^{\mathfrak{A}})$$

$$\text{Int}^{\mathfrak{A}} ::= \mathbb{Z}$$

$$\text{one}^{\mathfrak{A}} ::= 1$$

$$\text{plus}^{\mathfrak{A}}: \text{Int}^{\mathfrak{A}} \times \text{Int}^{\mathfrak{A}} \rightarrow \text{Int}^{\mathfrak{A}}, (a, b) \mapsto a + b$$

$$\text{leq}^{\mathfrak{A}} ::= \{ (a, b) \in \text{Int}^{\mathfrak{A}} \times \text{Int}^{\mathfrak{A}} \mid a \leq b \}$$

$$\beta: \text{Var} \rightarrow \text{Int}^{\mathfrak{A}}$$

$$\begin{aligned}\mathfrak{I}(\text{plus}(\text{plus}(\text{one}, \text{one}), x)) \\ &= \text{plus}^{\mathfrak{A}}(\text{plus}^{\mathfrak{A}}(\text{one}^{\mathfrak{A}}, \text{one}^{\mathfrak{A}}), \beta(x)) \\ &= (1+1) + \beta(x)\end{aligned}$$

FOL Semantics

\mathfrak{I} is a **model** of F iff $\mathfrak{I} \models F$

we can always
express equality
between terms

FOL formula F (excerpt)	$\mathfrak{I} = (\mathfrak{A}, \beta) \models F$ if and only if
$t_1 = t_2$	$\mathfrak{I}(t_1) = \mathfrak{I}(t_2)$
$R(t_1, \dots, t_n)$	$(\mathfrak{I}(t_1), \dots, \mathfrak{I}(t_n)) \in R^{\mathfrak{A}}$
$G \wedge H$	$\mathfrak{I} \models G$ and $\mathfrak{I} \models H$
$G \Rightarrow H$	If $\mathfrak{I} \models G$, then $\mathfrak{I} \models H$
$\exists x: T(G)$	For some $v \in T^{\mathfrak{A}}$, $\mathfrak{I}[x := v] \models G$
$\forall x: T(G)$	For all $v \in T^{\mathfrak{A}}$, $\mathfrak{I}[x := v] \models G$

F is **satisfiable** iff F has some model

Issues with FOL Satisfiability

- All symbols are **uninterpreted**
- The meaning of functions and relations is determined in the chosen model
- Many formulas are satisfiable if we can choose Σ -structures that defy the intended meaning of functions and relations

→ Filter out unwanted Σ -structures

$$\Sigma = \{ \text{Nat}, \text{zero}, \text{one}, \text{plus}, \text{leq} \}$$

$$F ::= \exists x: \text{Nat} (x \text{ plus one leq zero})$$

infix notation for $\text{leq}(\text{plus}(x, \text{one}), \text{zero})$

sat: $\text{Nat} = \mathbb{N}, \text{one}^{\mathfrak{U}} ::= 0, \text{leq} ::= \leq$
 $\text{zero}^{\mathfrak{U}} ::= 1, \text{plus}^{\mathfrak{U}} ::= +$

sat: $\text{Nat} = \mathbb{N}, \text{one}^{\mathfrak{U}} ::= 1, \text{leq} ::= \leq$
 $\text{zero}^{\mathfrak{U}} ::= 0, \text{plus}^{\mathfrak{U}} ::= -$

Satisfiability Modulo Theories

- A Σ -sentence is a formula without free variables
- An axiomatic system \mathbf{AX} is a set of Σ -sentences
- The Σ -theory \mathbf{Th} given by \mathbf{AX} is the set of all Σ -sentences implied by \mathbf{AX}

A Σ -formula F is **satisfiable modulo \mathbf{Th}** iff there exists a Σ -interpretation \mathfrak{I} such that

- $\mathfrak{I} \models F$, and
- $\mathfrak{I} \models G$ for every sentence G in \mathbf{Th} .

A Σ -formula F is **valid modulo \mathbf{Th}** iff $\neg F$ is *not* satisfiable modulo \mathbf{Th} .

Some important theories

- Arithmetic (with canonical axioms)
 - Presburger arithmetic: $\Sigma = \{ \text{Int}, <, 0, 1, + \}$ **decidable**
 - Peano arithmetic: $\Sigma = \{ \text{Int}, <, 0, 1, +, * \}$ **undecidable**
 - Real arithmetic: $\Sigma = \{ \text{Real}, <, 0, 1, +, * \}$ **decidable**
- EUF: Equality logic with Uninterpreted Functions **decidable**
 - $\Sigma = \{ \text{U}, =, f, g, h, \dots \}$
 - arbitrary non-empty domain **U**
 - axioms ensure that $=$ is an equivalence relation
 - arbitrary number of **uninterpreted function symbols** of any arity
 - axioms do *not* constrain function symbols
- We typically need a combination of multiple theories
 - Program verification: theories for modeling different data types

Overview

1. Propositional logic and SAT solvers
2. Using Z3 as a SAT solver
3. First-order logic and SMT solvers
4. Using Z3 as an SMT solver

Using Theories (SMTLIB-2)

- Sorts
 - Bool, Int, Real, BitVec(precision)
 - DeclareSort(name)
(uninterpreted)
- Uninterpreted functions are declared with parameter and return types
- Variables are uninterpreted functions of arity 0
 - Const(name, sort)

```
(declare-sort Pair)

(declare-fun cons (Int Int) Pair)
(declare-fun first (Pair) Int)

(declare-const null Pair)

; first axiom
(assert (= null (cons 0 0)))
; second axiom
(assert (forall ((x Int) (y Int))
                (= x (first (cons x y)))))
))

; formula (negated for validity check)
(assert (not (= (first null) 0)))

(check-sat)
```

Using Theories (Z3Py)

- Sorts
 - Bool, Int, Real,
BitVec(precision)
 - DeclareSort(name)
(uninterpreted)
- Uninterpreted functions are declared with parameter and return types
- Variables are uninterpreted functions of arity 0
 - Const(name, sort)

```
from z3 import *
Pair = DeclareSort('Pair')
null = Const('null', Pair)
cons = Function('cons', IntSort(), IntSort(), Pair)
first = Function('first', Pair, IntSort())
ax1 = (null == cons(0, 0))
x, y = Ints('x y')
ax2 = ForAll([x, y], first(cons(x, y)) == x)
s = Solver()
s.add(ax1)
s.add(ax2)
F = first(null) == 0
# check validity
s.add(Not(F))
print( s.check() )
```

Custom theories for user-defined data types and operations

- Encoding via
 - uninterpreted sorts
 - constants
 - uninterpreted functions
 - axioms enforcing the data type's properties
- We call such an encoding an **axiomatization**
- Week 5: advanced data types
 - sets, sequences, trees
 - accessors, mutators
 - recursive functions

```
(declare-sort Pair)

(declare-fun cons (Int Int) Pair)
(declare-fun first (Pair) Int)

(declare-const null Pair)

; first axiom
(assert (= null (cons 0 0)))
; second axiom
(assert (forall ((x Int) (y Int))
                (= x (first (cons x y)))))
))

; formula (negated for validity check)
(assert (not (= (first null) 0)))

(check-sat)
```

Incorporating custom theories

Original verification condition: $P \implies WP(S, Q)$ valid

A Σ -formula F is **valid modulo Th** iff $\neg F$ is *not* satisfiable modulo **Th**.

A Σ -formula F is **satisfiable modulo Th** iff there exists a Σ -interpretation \mathfrak{I} such that

- $\mathfrak{I} \models F$, and
- $\mathfrak{I} \models G$ for every sentence G in **Th**.

Enriched verification condition:

$P \implies WP(S, Q)$ valid modulo custom theory
iff $BP \And P \And \neg WP(S, Q)$ unsat
iff $BP \implies P \implies WP(S, Q)$ valid

Background Predicate:
conjunction of all our axioms
defining our theory

Automating Program Verification

Mains steps of a tool for checking that $\{ P \} S \{ Q \}$ is valid:

0. Determine axioms of underlying theory → background predicate BP
 - *Reusable*: once for every type or function → Week 5
 - Z3 has built-in theories for common theories (e.g. arithmetic) → Today
1. Compute $WP(S, Q)$ → Week 1 & 2
2. Check whether $BP ==> P ==> WP(S, Q)$ is valid → SMT solver
 - Check satisfiability of negation: $BP \And P \And !WP(S, Q)$
 - unsat → 
 - sat →  model explains why $\{ P \} S \{ Q \}$ is not valid

Axiomatize with Care

- Axiomatizations are part of the trusted codebase
- Inconsistent axioms invalidate verification results

$\text{false} \Rightarrow P \Rightarrow \text{WP}(S, Q)$ is always valid

- Axioms do not show up in verification problems

$\{ x == \text{null} \} S \{ y > \text{null} \}$

- Axiomatizations require separate validation
 - proofs, testing, profiling, ...

```
(declare-const null Int)

; inconsistent axioms
(assert (= null 0))
(assert (= null 17))

(assert (not
        (> 42 23) ; wrong statement
))

(check-sat) ; unsat
```



Theory applications

Linear integer/real arithmetic

$$19 * x + 2 * y = 42$$

- (Unbounded) arithmetic is often used to approximate int and float
- Multiplication by constants is supported

Non-linear integer/real arithmetic

$$x * y + 2 * x * y + 1 = (x + y) * (x + y)$$

- Useful for programs that perform multiplication and division, e.g., crypto libraries

Equality logic with uninterpreted functions

$$(x = y \wedge u = v) \Rightarrow f(x, u) = f(y, v)$$

- Universal mechanism to encode operations not natively supported by a theory

Fixed-size bitvector arithmetic

$$x \& y \leq x \mid y$$

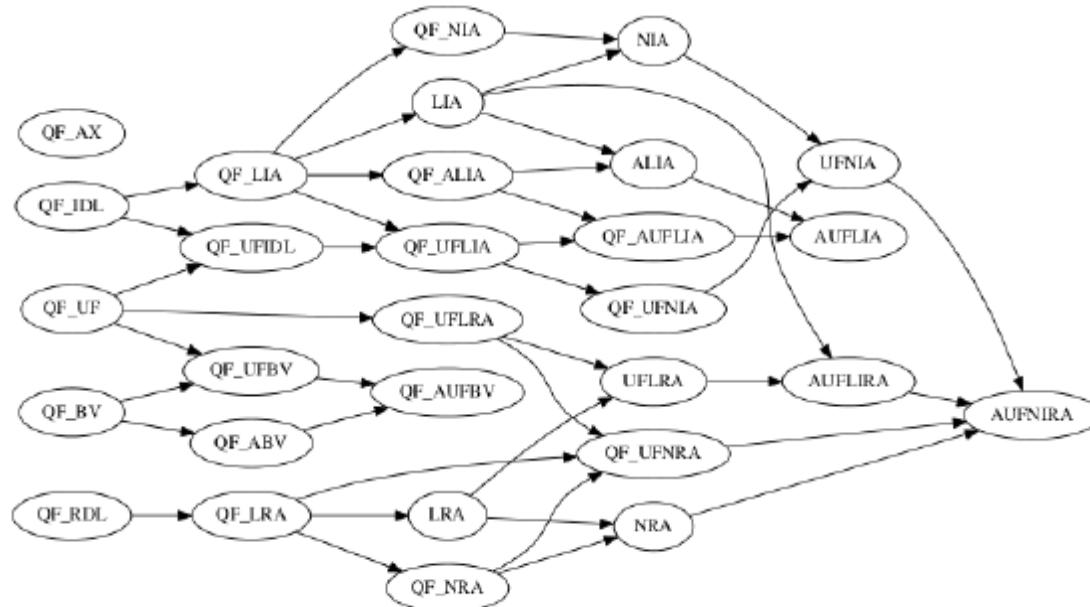
- To encode bit-level operations
- To perform bit-precise reasoning, e.g., floats

Array theory

$$\text{read}(\text{write}(a, i, v), i) = v$$

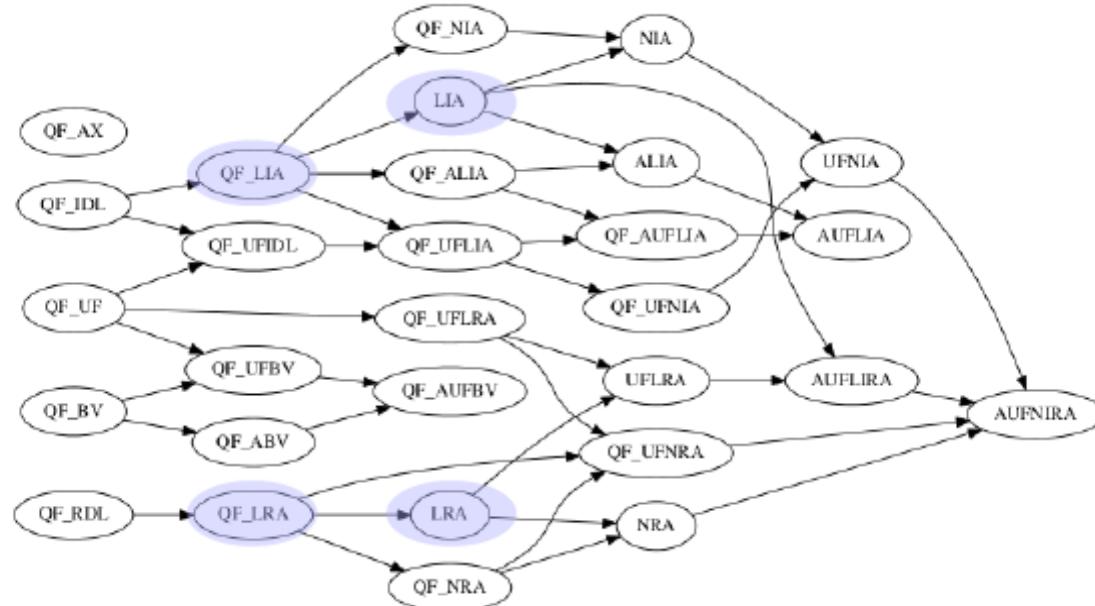
- To encode data types such as arrays

Z3 built-in theories



→ no explicit background predicate needed

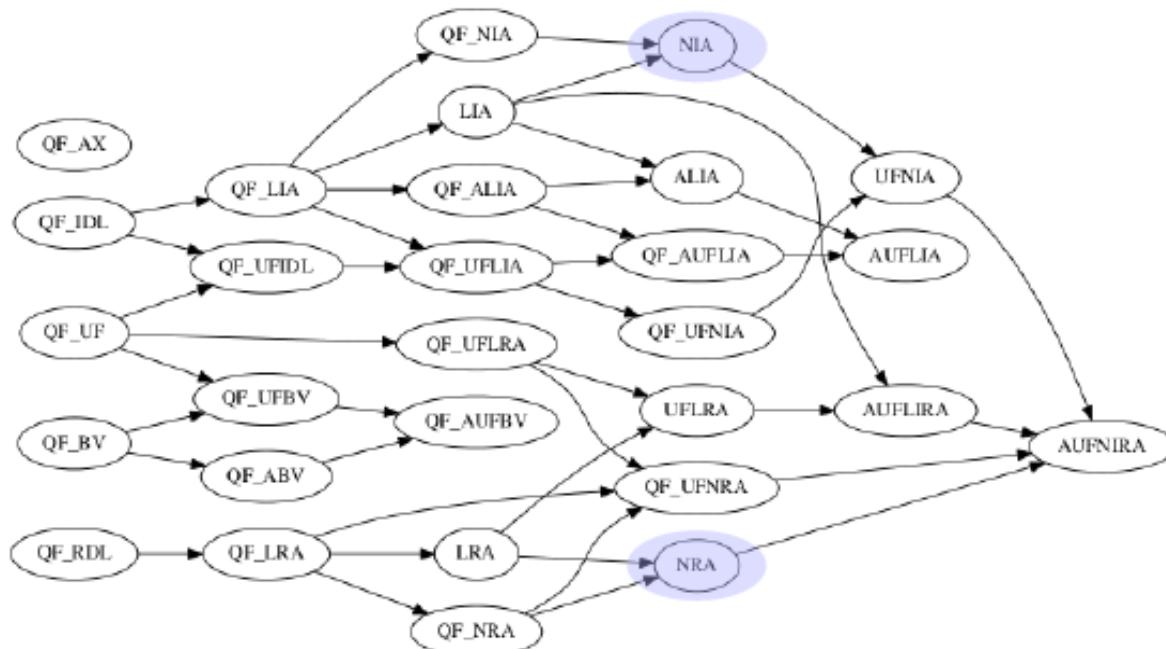
Z3 built-in theories



(Quantifier-free) Linear Integer/Real Arithmetic

$$19 * x + 2 * y = 42$$

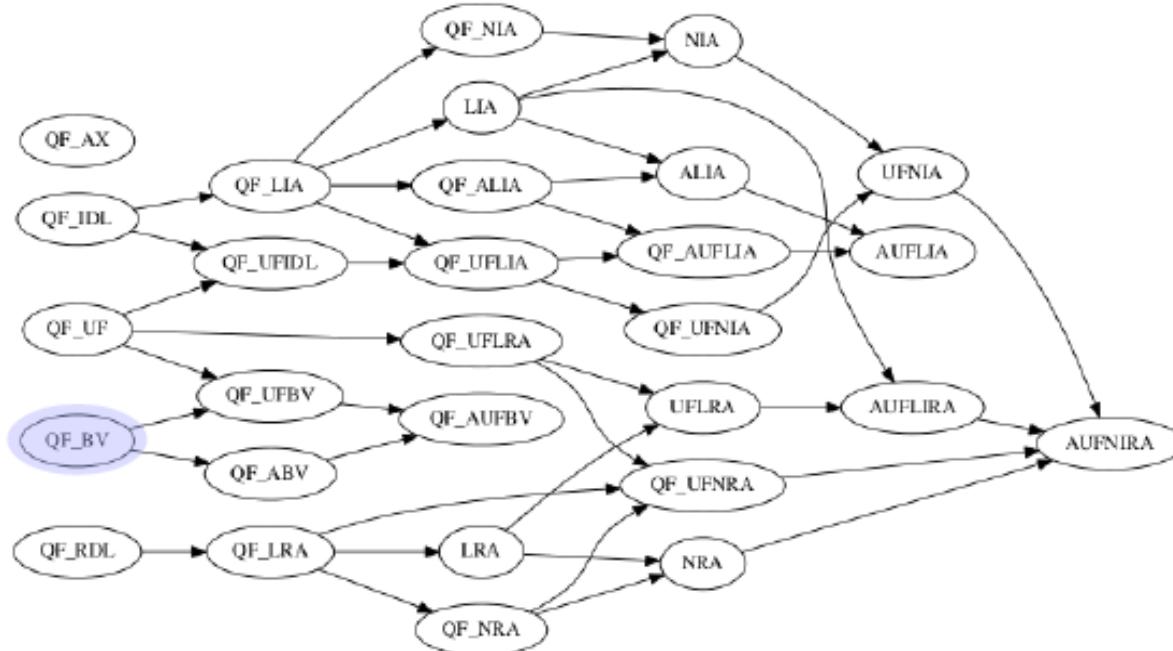
Z3 built-in theories



Non-Linear Integer/Real Arithmetic

$$x * y + 2 * x * y + 1 = (x + y) * (x + y)$$

Z3 built-in theories



Quantifier-free fixed-size bitvector arithmetic

$$x \& y \leq x \mid y$$

Using Z3 to verify a program

```
{ a = 1 ∧ 0 ≤ b*b - 4*c }
discriminant := b*b - 4*a*c;
if (discriminant < 0) {
    assert false
} else {
    x := (b + √discriminant) / 2
}
{ a*x*x + b*x + c = 0 }
```

Step 1: use **WP** to determine the verification condition

Using Z3 to verify a program

```
{ a = 1 ∧ 0 ≤ b*b - 4*c }
// ==>
{ b*b - 4*a*c < 0 ∧ false ∨
  ¬(b*b - 4*a*c < 0) ∧ a*((-b + √b*b - 4*a*c) / 2)^2 + b*((-b + √b*b - 4*a*c) / 2) + c = 0 }
discriminant := b*b - 4*a*c;
{ discriminant < 0 ∧ false ∨
  ¬discriminant < 0 ∧ a*((-b + √discriminant) / 2)^2 + b*((-b + √discriminant) / 2) + c = 0 }
if (discriminant < 0) {
{ false }
  assert false
{ a*x^2 + b*x + c = 0 }
} else {
{ a*((-b + √discriminant) / 2)^2 + b*((-b + √discriminant) / 2) + c = 0 }
  x := (-b + √discriminant) / 2
{ a*x^2 + b*x + c = 0 }
}
{ a*x^2 + b*x + c = 0 }
```

Using Z3 to verify a program

- Step 1: use **WP** to determine the verification condition

```
{ a = 1 ∧ 0 ≤ b*b - 4*c }
// ==>
{ b*b - 4*a*c < 0 ∧ false ∨
  ¬(b*b - 4*a*c < 0) ∧ a*((-b + √b*b - 4*a*c) / 2)^2 + b*((-b + √b*b - 4*a*c) / 2) + c = 0 }
```

- Step 2: check whether the verification condition is valid
 - Check satisfiability of negation: `Pre && !WP(S, Post)`

```
; declarations ... (full example available online)
; precondition
(assert (and (= a 1) (≤ 0 (- (* b b) (* 4 c)))))

; negated weakest precondition
(assert (not <complicated expression here>))

(check-sat) ; want: unsat
```

Z3's Theory Reasoning

- Z3 selects theories based on the features appearing in formulas
 - Most verification problems require a combination of many theories

Quantifier-free linear integer arithmetic with uninterpreted functions

$$17 * x + 23 * f(y) > x + y + 42$$
- Some theories are decidable, e.g., quantifier-free linear arithmetic
 - SMT solver will terminate and report either “sat” or “unsat”
- Some theories are undecidable, e.g., nonlinear integer arithmetic
 - Especially in combination with quantifiers
 - SMT solver uses heuristics and may not terminate or return “unknown”
 - Results can be flaky, e.g., depend on order of declarations or random seeds

SMT solvers

1. Propositional logic and satisfiability solvers
2. Using Z3 as a SAT solver
3. First-order logic and SMT solvers
4. Using Z3 as an SMT solver
5. Quantifiers

Quantifiers

- Program specifications will often require quantifiers to express **complex properties**
 - “Array a is sorted”: $\forall i, j. 0 \leq i < j < |a| \Rightarrow a[i] \leq a[j]$
 - “All entries in a are non-null”: $\forall i. 0 \leq i < |a| \Rightarrow a[i] \neq \text{null}$
- Quantifiers are also useful to model **additional theories and data types**
 - Function definitions: $\forall x. f(x) = e$
 - We will see an example of this soon
- Addition of quantifiers makes many problems undecidable
 - **Quantifier-free** linear arithmetic is decidable
 - Linear arithmetic **with quantifiers** is not

Existential quantifiers

- Existential quantifiers in positive positions can be eliminated via [skolemization](#)
 - Positive position means non-negated.
 - Quantifier in $(\exists x. P) \wedge Q$ is in a positive position, in $(\exists x. P) \Rightarrow Q$ it is not.

$\exists x. P$ is \mathcal{T} -satisfiable iff $P[x/x']$, where x' is fresh, is \mathcal{T} -satisfiable.

- Quantifier alternation requires introducing fresh functions
 - Example: $\forall x. \exists y. P$ is satisfiable iff $\forall x. P[y/f(x)]$ is satisfiable, where f is fresh
- Quantifiers in negative positions can be rewritten:

$$\neg \exists x. P \Leftrightarrow \forall x. \neg P$$

$$\neg \forall x. P \Leftrightarrow \exists x. \neg P$$

Universal quantifiers

- Universal quantifiers **cannot** be eliminated in general
 - Checking whether some formula is satisfiable for **all values in an infinite sort** is a hard problem
- Recall: Our goal is to prove **validity** of P , i.e., to prove $\neg P$ unsatisfiable
 - Usually, to prove unsatisfiability, only **specific instantiations** are needed

Example: To prove unsatisfiability of

$$\mathbf{F} ::= (f(0) = 1) \wedge (\forall x. f(x) = x)$$

we only need to instantiate x with 0 to get

$$(f(0) = 1) \wedge (f(0) = 0)$$

Unsatisfiable

Universal quantifier instantiation

- Idea: Let users provide hints about relevant quantifier instantiations
- Treat quantified constraints separately from non-quantified constraints
 - Do not attempt to satisfy quantifier in general
 - Quantifier only generates specific instantiations, solver satisfies all generated instantiations
 - Resulting formulas can be solved by decision procedures for quantifier-free problems
- New format of universal quantifiers is $\forall x. \{e(x)\}P$
 - $e(x)$ is the pattern or trigger, must contain all quantified variables
 - Quantifier body P is instantiated for any term with the shape $e(x)$ occurring in the current constraints or model
 - Trigger can consist of multiple terms, quantifier can have multiple triggers
 - Most SMT solvers impose syntactic restrictions on possible trigger terms

Quantifier instantiation using triggers

- A **ground term** is a term in the formula containing no quantified variables

A quantified constraint $\forall x. \{e(x)\} P$ in the current formula leads to an instantiation $P[x/v]$ iff the current formula/model contains a ground term $e(v)$

- Example: $(\forall x. \{f(x)\} f(x) > 0) \wedge f(2) < 2$ instantiates $f(2) > 0$
- Trigger matching is performed **modulo equality**
 - Example: $g(1) = 2 \wedge h(2) = 3 \wedge (\forall x. \{h(g(x))\} h(g(x)) < x)$ instantiates $h(g(1)) < 1$
 - This instantiation technique is called **e-matching** for this reason

Problematic triggers

- Too strict or wrong triggering leads to incompleteness
 - Example: $f(0) = 1 \wedge (\forall x. \{g(f(x))\} f(x) = x)$ Unknown
- Generally, SMT solvers using e-matching cannot prove satisfiability
 - Quantifiers might not be satisfied for values for which they were not triggered
 - Only report unsat or unknown
- Too liberal triggering leads to many or infinite instantiations
 - Cause of most performance issues in automated verifiers
 - Always write triggers yourself! Otherwise, the solver will infer them
- Matching loop: Quantifier instantiation (transitively) triggers same quantifier again, ad infinitum
 - Example: $\forall i. \{a[i]\} 0 \leq i < |a| - 1 \Rightarrow a[i] \leq a[i + 1]$
 - In practice, tools eventually stop after some limit is reached

Mathematical data types

- Our language so far has a very restricted set of types
- Mathematical data types are useful to encode languages that have them
 - Many functional languages offer lists, tuples, and abstract data types (ADTs)
- Mathematical data types will also be important to specify imperative code
 - “*Array sort leaves the *multiset* of elements unchanged*”
 - “*All implementations of Java’s List interface store a *sequence* of elements*”
 - We will discuss this kind of specification in Lecture 7

Types

$T ::= \text{Bool} \mid \text{Int} \mid \text{Rational} \mid \text{Real}$

Encoding of custom data types

- We encode custom data types into SMT using

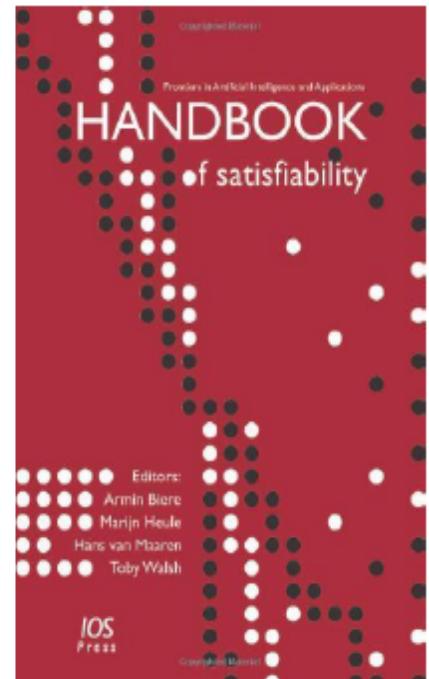
- uninterpreted types
- constants
- uninterpreted functions
- axioms that express properties of the constants and functions

```
from z3 import *
Set = DeclareSort('Set')
empty = Const('empty', Set)
card = Function('card', Set, IntSort())
...
ax1 = (card(empty) == 0)
...
```

- We call such an encoding an **axiomatization**
- Generics can be handled via monomorphization, that is, by generating a separate axiomatization for each instantiation of a generic type

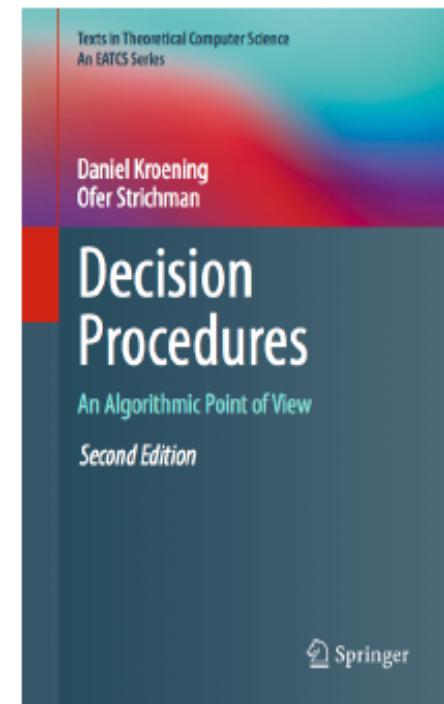
More background on SAT solvers

- DPLL: Davis-Putnam-Logemann-Loveland Algorithm
 - [A machine program for theorem-proving](#). Martin Davis, George Logemann, and Donald Loveland. 1962.
- CDCL: Conflict-Driven Clause Learning Algorithm
 - [GRASP – A New Search Algorithm for Satisfiability](#). João P. Marques Silva and Karem A. Sakallah. 1996.
- Further developments
 - [Chaff: engineering an efficient SAT solver](#). Matthew W. Moskewicz, Conor F. Madigan, Ying Zhao, Lintao Zhang, and Sharad Malik. 2001.
 - [SAT-solving in practice](#). Koen Claessen, Niklas Een, Mary Sheeran, Niklas Sörensson. 2008.
- Annual SAT competition:
 - <http://www.satcompetition.org/>



More background on SMT solvers

- <http://www.decision-procedures.org/> (website of book)
- [Programming Z3](#), Nikolaj Bjørner, Leonardo de Moura, Lev Nachmanson, Christoph M. Wintersteiger, 2018
- [SMT-LIB standard](#)
- Other teaching material
 - SMT solvers: Theory and Implementation. Leonardo de Moura
 - SMT Solvers: Theory and Practice. Clark Barrett
 - Satisfiability Checking, Erika Ábrahám
- Efficient E-Matching for SMT Solvers. Leonardo de Moura, Nikolaj Bjørner, 2007



Wrap-up

Main steps of a tool for checking that $\{ P \} S \{ Q \}$ is valid:

0. Determine axioms of underlying theory

→ background predicate BP

- *Reusable*: once for every type or function
- Z3 has **built-in theories** for common theories (e.g. arithmetic)

→ Week 5

1. Compute $WP(S, Q)$

→ Week 1 & 2

2. Check whether $BP ==> P ==> WP(S, Q)$ is valid

→ SMT solver

- Check satisfiability of negation: $BP \And P \And !WP(S, Q)$
- unsat →  model explains why $\{ P \} S \{ Q \}$ is not valid
- sat →  model explains why $\{ P \} S \{ Q \}$ is not valid
- unknown → decidability issues, strengthen theory, hacks

→ future classes