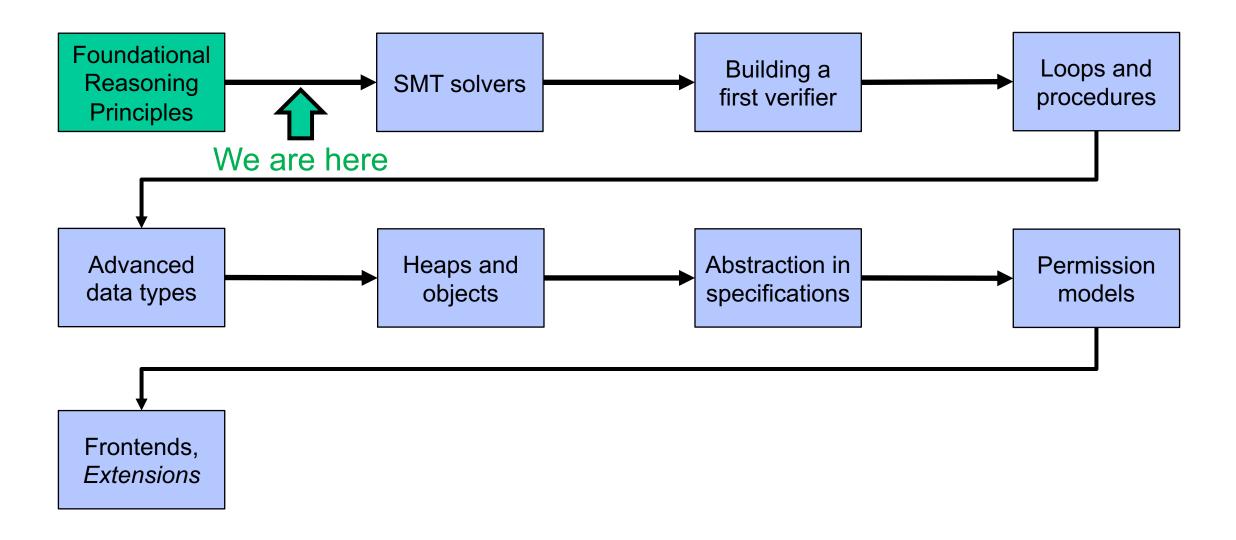
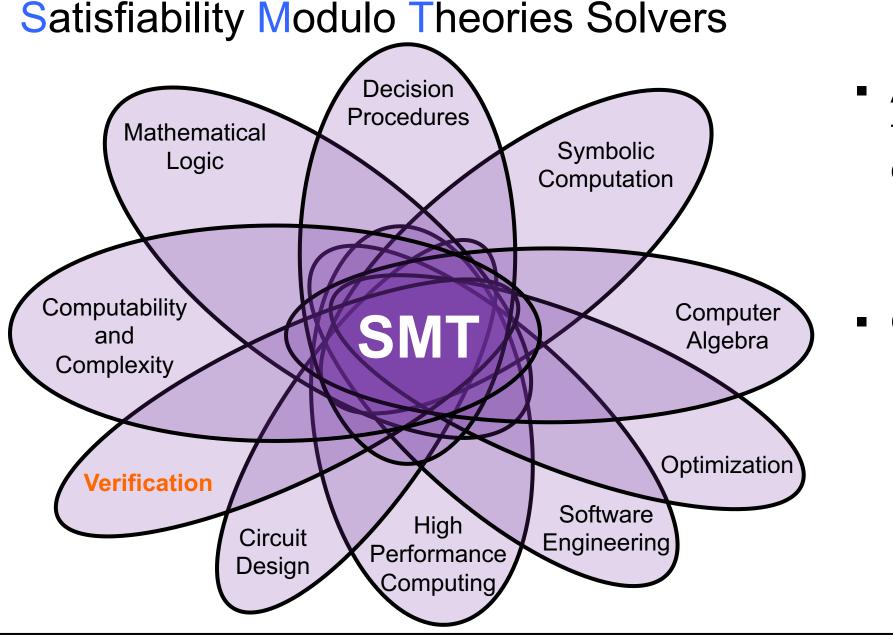
02245 – Lecture 2 FOUNDATIONS & SMT SOLVERS

Tentative course outline



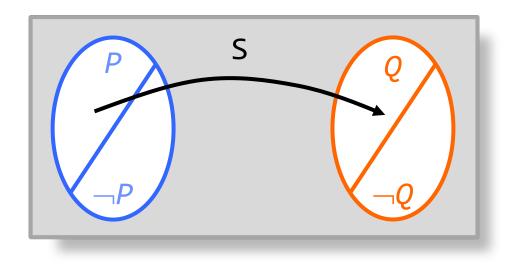


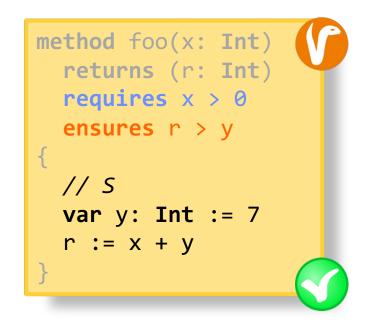
A foundational topic in theoretical and applied computer science

Our focus: effectively applying SMT technology to program verification

But first: Recap

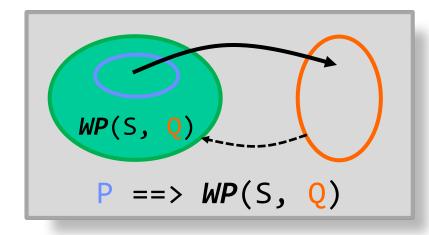
The Floyd-Hoare triple { P } S { Q } is valid if and only if every execution of S that starts in a state satisfying precondition P terminates without an error in a state satisfying postcondition Q.

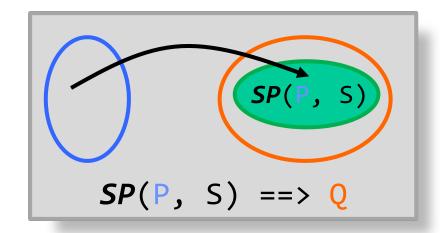






Recap: Weakest Pre & Strongest Post





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

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 $P \implies WP(S, Q)$ 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
```



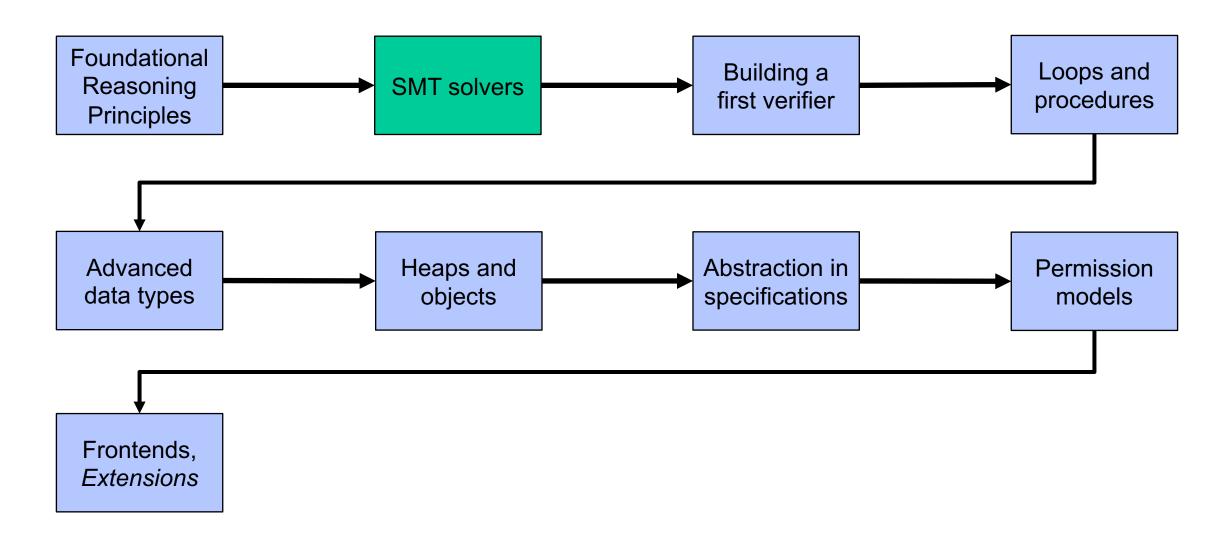
< Homework W1 >

Solutions will be published on course page

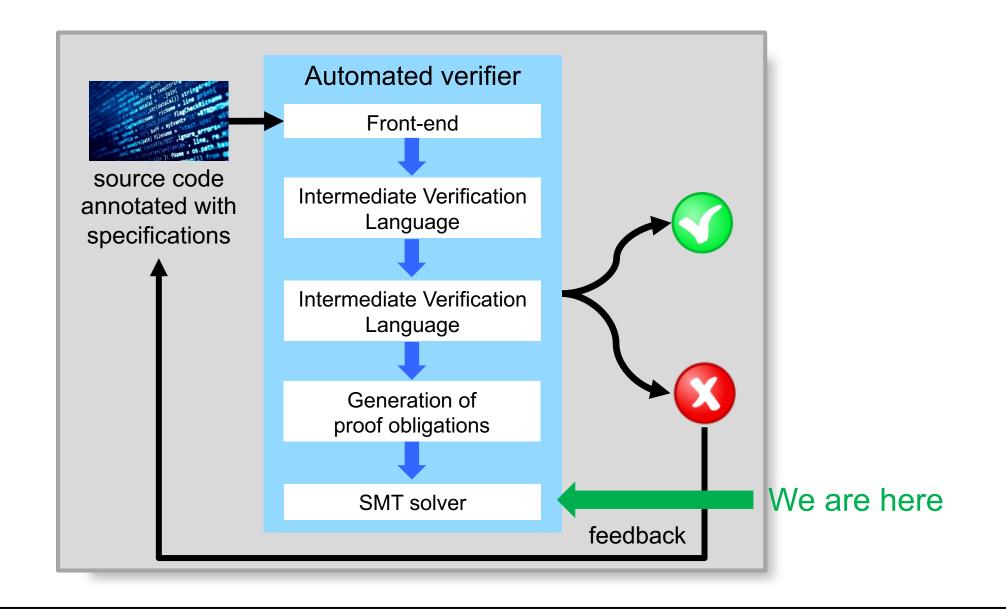


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Tentative course outline



Roadmap

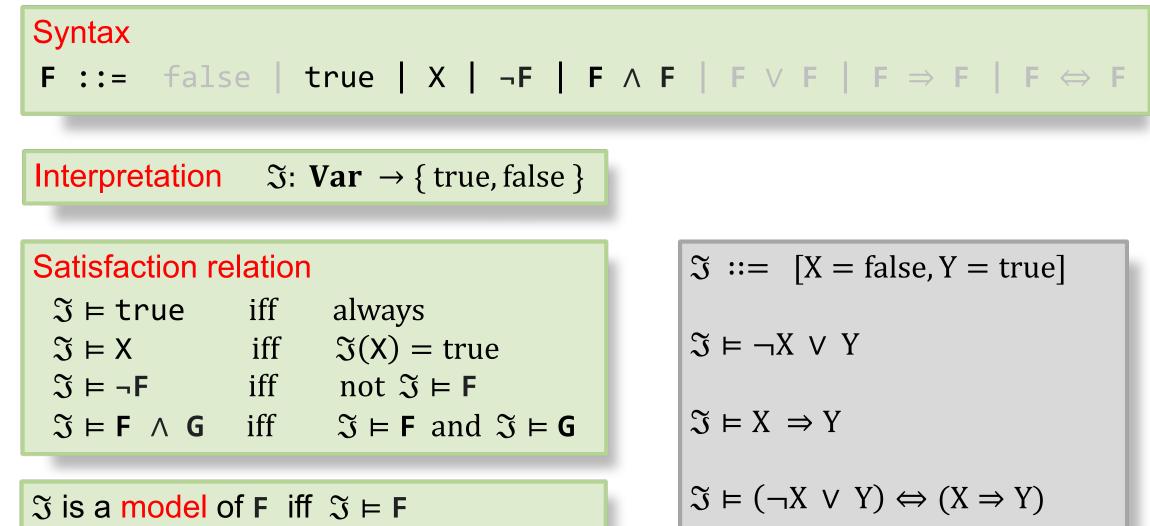


Overview

- 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



Satisfiability & Validity

• F is satisfiable iff F has some model

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

Models: [X = true, Y = true], [X = false, Y = true], [X = true, Y = false]

• F is unsatisfiable iff F has no model

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

 F is valid iff every interpretation is a model of F (¬F is unsatisfiable)

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

• F is not valid iff some interpretation is not a model of F $X \land (X \Rightarrow Y) \Leftrightarrow Y$ (¬F is satisfiable)

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

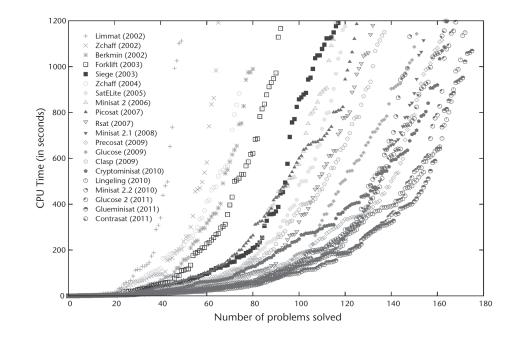
 $(X \lor Y \lor \neg Z)$ $\land (U \lor \neg Y)$ $\land (\neg X \lor \neg Z \lor U \lor V)$

$$\Im$$
 ::= [
U = false
V = false
X = true
Y = false
Z = false
]

Complexity of SAT

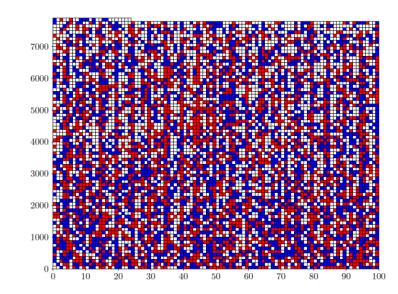
- For formulas in conjunctive normal form (CNF), SAT is the classical NP-complete problem
- Many difficult problems can be efficiently encoded
- Every known algorithm is exponential in the formula's size

$$\bigwedge_{i} \bigvee_{j} C_{i,j} \quad \text{where } C_{i,j} \in \{X_{i,j}, \neg X_{i,j}\}$$



Example: Boolean Pythagorean Triples

- BPT: a triple of natural numbers $1 \le a \le b \le c$ 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, ..., 7825} always contains a monochromatic BPT
- This was first proven using a SAT solver
 - number of combinations: 27825
 - "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

Exercise: Seating of Wedding Guests





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Exercise

- Model the following problem as an instance of the SAT problem.
- There are three chairs in a row: left, middle, right.
- Can we assign chairs to Alice, Bob, and Charlie such that:
 - Alice does not sit next to Charlie,
 - Alice does not sit on the leftmost chair, and
 - Bob does not sit to the right of Charlie?

Overview

- 1. Propositional logic and SAT solvers
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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) \land X
(assert (=> 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')	
<pre># define formula F F = And(Implies(X, Implies(Y, Z)), X)</pre>	
solve(F) # find a model for F	
<pre># find a counterexample for F solve(Not(F))</pre>	

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

Example: Course Selection

- You have to take CS Modeling, Physics, or Chemistry
- For CS Modeling, you also need Discrete Math
- For Verification, you need CS Modeling
- For Physics and Chemistry, you need Statistics
- Statistics and Discrete Math are at the same time
- CS Modeling and Physics are at the same time
- Verification and Chemistry are at the same time

Is it possible to take Verification and all preliminaries?

Is it possible to take Physics and Discrete Math?

Exercise: Seating of Wedding Guests

- Use Z3 to check whether we can assign suitable seats to all wedding guests
- There are three chairs in a row: left, middle, right.
- We want to assign chairs to Alice, Bob, and Charlie such that:
 - Alice does not sit next to Charlie,
 - Alice does not sit on the leftmost chair, and
 - Bob does not sit to the right of Charlie.



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?

Using Z3 for Homework

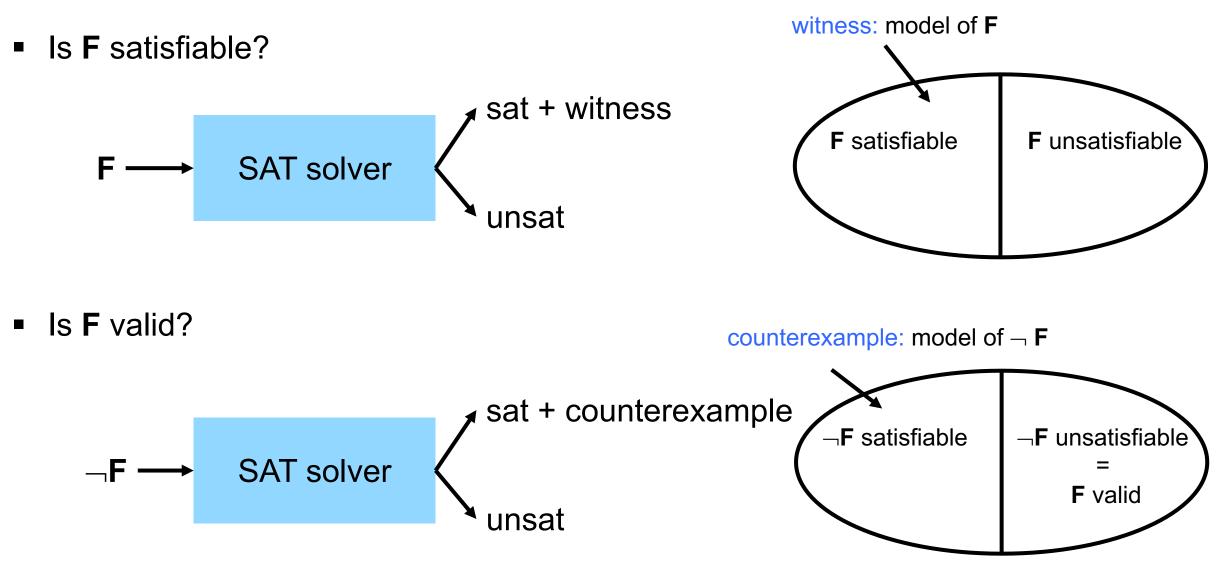
Here is an excerpt from a proof in the first homework assignment:

```
valid: P ==> WP(assert R, Q)
iff
valid: P ==> (R && Q)
iff
valid: P ==> R
and valid: (P && R) ==> Q
iff
valid: SAFE(p, assert R)
and valid: SP(P, assert R) ==> Q
```

Use Z3 to prove the blue equivalence.



Using a SAT solver



Using a SAT Solver for Program Verification

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

1. Compute wp(S, Q)

- 2. Check whether entailment P => WP(S, Q) is valid
 - Check satisfiability of negation: P && !WP(S, Q)

 - sat → model explains why { P } S { Q } is not valid

→ last lecture

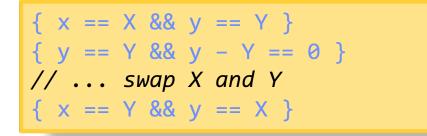
ask SAT solver

Using a SAT Solver for Program Verification

```
{ true }
// check that validity of true \Rightarrow a \land \ldots
\{a \land (b \land (true \Leftrightarrow (a \Rightarrow b)) \lor \neg b \land (false \Leftrightarrow (a \Rightarrow b))) \lor \neg a \land (true \Leftrightarrow (a \Rightarrow b))\}
if (a) {
\{ b \land (true \Leftrightarrow (a \Rightarrow b)) \lor \neg b \land (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



```
Entailment to check:
(x == X && y == Y) ==> 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

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

- 1. Sorts
 - specifies possible types
 - we assume a corresponding set of values
- 2. Typed Variables
- 3. Typed Function symbols
 - building blocks of terms
- 4. Typed Relational symbols
 - turn terms into logical propositions
- 5. Logical symbols

Bool, Int, Real, T

Bool = {true, false} X, Y, Z, ... 0, 1.5 +, *, _?_:_ x ? y - 17 : z*z + 1 0 prime R < prime(y+4) R(x,y,z) $\mathbf{X} = \mathbf{0}$ $\land \lor \lor \neg \Rightarrow \Leftrightarrow \exists \lor \sqcup \sqcup$

FOL Formulas

- A signature Σ is a set of
 - at least one sort
 - function symbols
 - relational symbols (including =)
- A Σ-formula is a logical formula over propositions built from symbols in Σ

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

 $\forall x: Int \exists y: Int (y = x + 1 \land y * y = x * x + (1 + 1) * x + 1)$

Is this Σ -formula satisfiable?

FOL Σ-Interpretations

- A Σ-structure 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
 U to domain elements in U^A
- A **\Sigma-interpretation** is a pair $\mathfrak{T} = (\mathfrak{A}, \beta)$
- ℑ(t) denotes the domain element obtained by evaluating term t in ℑ

 $\Sigma = \{ Int, one, plus, eq \}$

 $\mathfrak{A} :::= (\mathbf{Int}^{\mathfrak{A}}, one^{\mathfrak{A}}, plus^{\mathfrak{A}}, eq^{\mathfrak{A}})$ $\mathbf{Int}^{\mathfrak{A}} :::= \mathbb{Z}$ $one^{\mathfrak{A}} :::= 1$ $plus^{\mathfrak{A}} ::\mathbf{Int}^{\mathfrak{A}} \times \mathbf{Int}^{\mathfrak{A}} \to \mathbf{Int}^{\mathfrak{A}}, (a, b) \mapsto a + b$ $eq^{\mathfrak{A}} ::= \{ (a, b) \in \mathbf{Int}^{\mathfrak{A}} \times \mathbf{Int}^{\mathfrak{A}} | a = b \}$

 β : Var \rightarrow Int^{\mathfrak{A}}

 $\Im(plus(plus(one, one), x))$ = plus^{\mathfrak{A}}(plus^{\mathfrak{A}}(one^{\mathfrak{A}}, one^{\mathfrak{A}}), $\beta(x)$) = (1+1) + $\beta(x)$

FOL Semantics

\Im is a model of F iff $\Im \models F$

FOL formula F (excerpt)	$\mathfrak{I} = (\mathfrak{A}, \beta) \vDash F$ if and only if
$t_{1} = t_{2}$	$\Im(t_1) = \Im(t_2)$
$R(t_1, \dots, t_n)$	$(\Im(t_1), \dots, \Im(t_n)) \in R^{\mathfrak{A}}$
$\mathbf{G} \wedge \mathbf{H}$	$\mathfrak{I} \vDash \mathbf{G}$ and $\mathfrak{I} \vDash \mathbf{H}$
$G \Rightarrow H$	If $\mathfrak{I} \models \mathbf{G}$, then $\mathfrak{I} \models \mathbf{H}$
$\exists x: \mathbf{T} (\mathbf{G})$	For some $v \in T^{\mathfrak{A}}$, $\mathfrak{I}[x := v] \models G$
$\forall x: \mathbf{T} (\mathbf{G})$	For all $v \in \mathbf{T}^{\mathfrak{A}}$, $\mathfrak{I}[x \coloneqq v] \vDash \mathbf{G}$

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
- \rightarrow Filter out unwanted Σ-structures

 $\Sigma = \{ Nat, zero, one, plus, eq \}$

$$\mathbf{F} ::= \exists x: \mathbf{Nat}(x \ plus \ one \ eq \ zero)$$

sat:
$$Nat = \mathbb{N}, one^{\mathfrak{A}} ::= 0, eq ::= =$$

 $zero^{\mathfrak{A}} ::= 1, plus^{\mathfrak{A}} ::= +$

sat:
$$Nat = \mathbb{N}, one^{\mathfrak{A}} ::= 1, eq ::= =$$

 $zero^{\mathfrak{A}} ::= 0, plus^{\mathfrak{A}} ::= -$

Satisfiability Modulo Theories

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

A Σ -formula F is satisfiable modulo Th iff there exists a Σ -interpretation \Im such that

- $\Im \models F$, and
- $\Im \models G$ for every sentence G in Th.

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

Exercise

- Consider the signature Σ = { Nat, zero, one, plus, eq },
- the theory **Th** given by the axioms

 $\forall x: \mathbf{Nat} (x \ eq \ x) \qquad \forall x: \mathbf{Nat} \ \forall x: \mathbf{Nat} \ (x \ plus \ y \ eq \ y \ plus \ x)$

- and the formula $F ::= \exists x: Nat(x plus one eq zero)$.
- a) Give a model witnessing that **F** is satisfiable modulo **Th**.
- b) Propose an axiom such that **F** is also valid modulo **Th**.



Some important theories

- Arithmetic (with canonical axioms)
 - Presburger arithmetic: $\Sigma = \{ Int, =, <, 0, 1, + \}$
 - Peano arithmetic:
 - Real arithmetic:
- $\Sigma = \{ Int, =, <, 0, 1, +, * \}$ $\Sigma = \{ Real, =, <, 0, 1, +, * \}$
- EUF: Equality logic with Uninterpreted Functions

decidable

decidable

decidable

undecidable

- $\Sigma = \{ \mathbf{U}, =, f, g, h, ... \}$
- 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

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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)

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

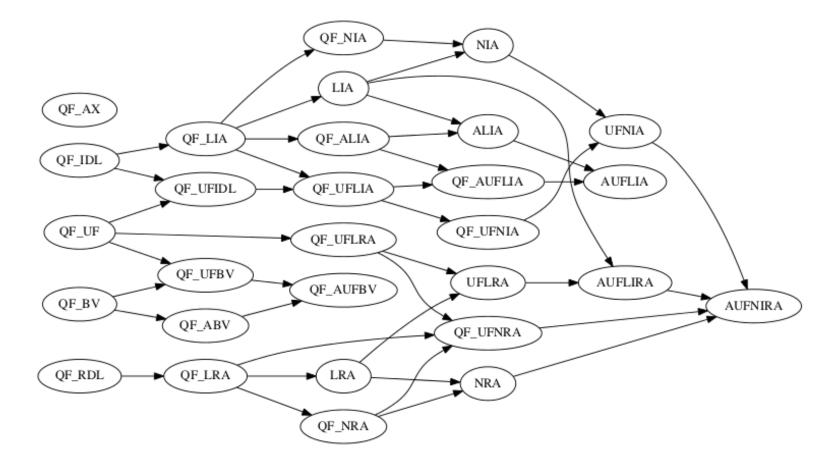
(check-sat)

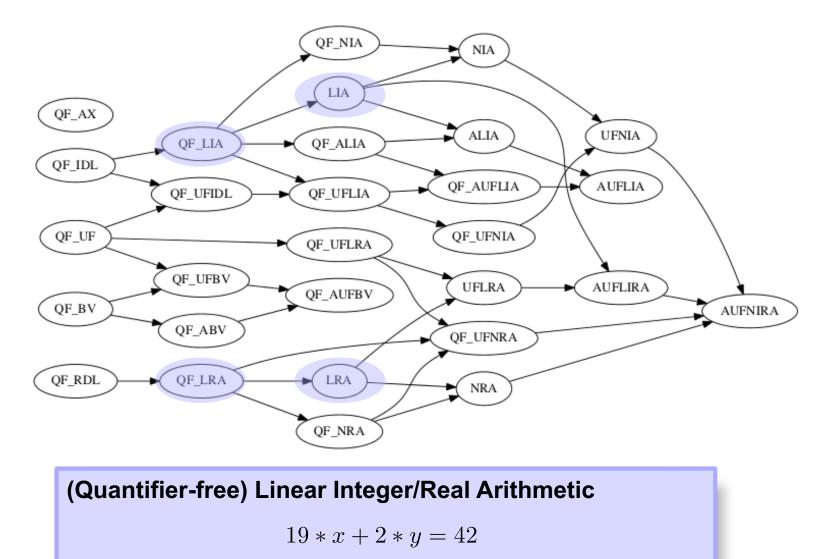


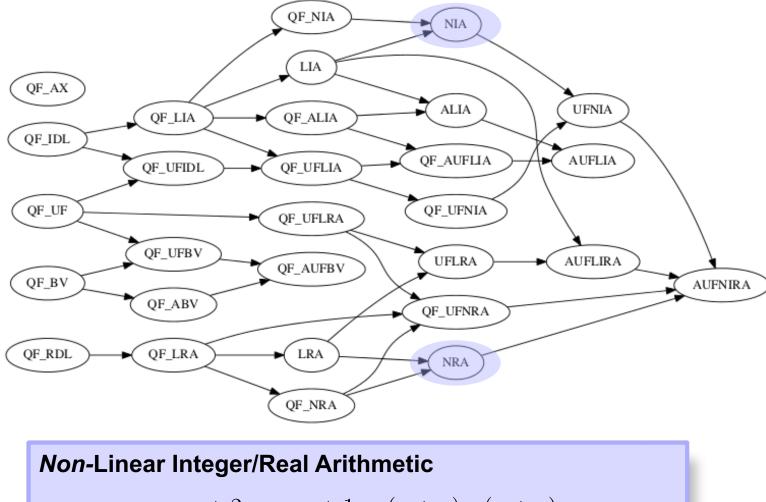
Using Theories (Z3Py)

- 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)

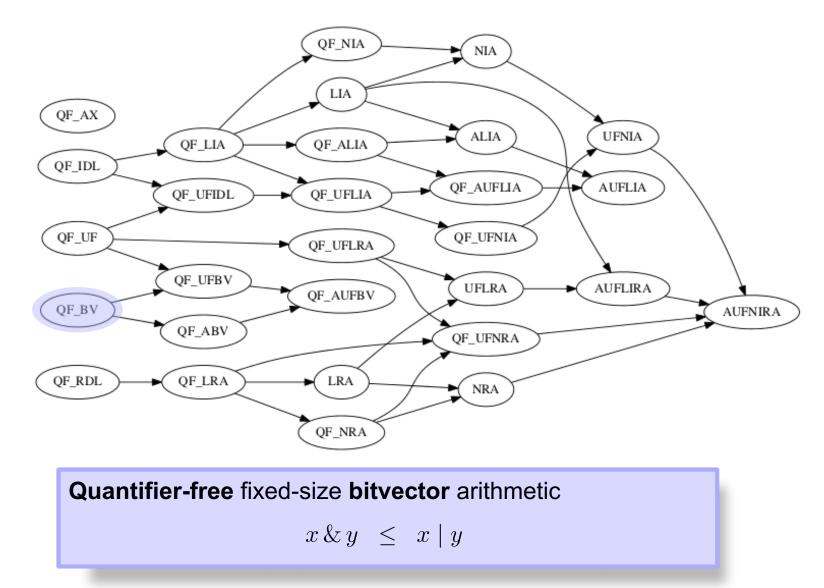
```
from z3 import *
Pair = DeclareSort('Pair')
null = Const('null', Pair)
cons = Function('cons', IntSort(), IntSort(), Pair)
first = Function('first', Pair, IntSort())
ax1 = (null == cons(0, 0))
x, y = Ints('x y')
ax2 = ForAll([x, y], first(cons(x, y)) == x)
s = Solver()
s.add(ax1)
s.add(ax2)
F = first(null)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



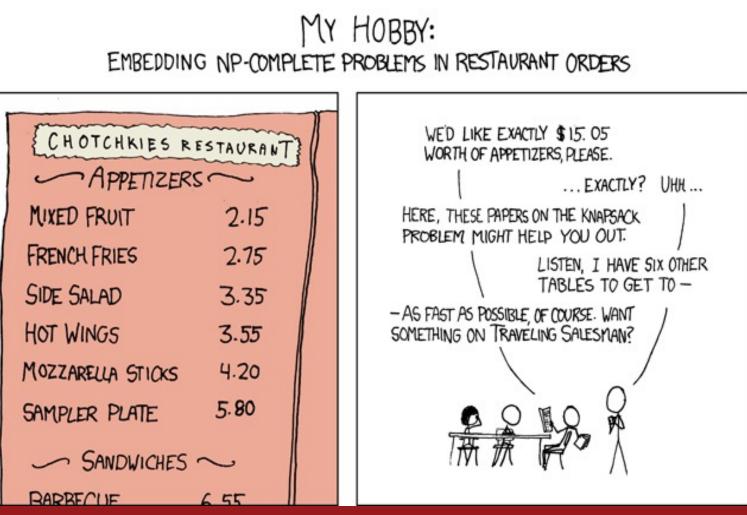




x * y + 2 * x * y + 1 = (x + y) * (x + y)



Exercise: use Z3 to find all suitable restaurant orders



https://xkcd.com/287/

Using Z3 to verify a program

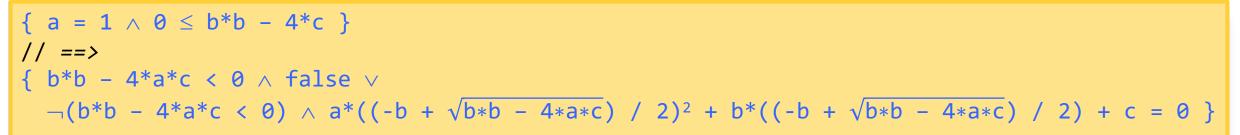
Step 1: use *WP* to determine the verification condition

Using Z3 to verify a program

```
\{ a = 1 \land 0 \le b*b - 4*c \}
// ==>
{ b*b − 4*a*c < 0 ∧ false ∨
  \neg (b*b - 4*a*c < 0) \land a*((-b + \sqrt{b*b} - 4*a*c) / 2)<sup>2</sup> + b*((-b + \sqrt{b*b} - 4*a*c) / 2) + c = 0 }
discriminant := b*b - 4*a*c;
{ discriminant < 0 \land false \lor
  \neg \text{discriminant} < 0 \land a^*((-b + \sqrt{\text{discriminant}}) / 2)^2 + b^*((-b + \sqrt{\text{discriminant}}) / 2) + c = 0 \}
if (discriminant < 0) {</pre>
{ false }
  abort
\{a^*x^2 + b^*x + c = 0\}
} else {
{ a^{*}((-b + \sqrt{discriminant}) / 2)^{2} + b^{*}((-b + \sqrt{discriminant}) / 2) + c = 0 }
  x := (-b + \sqrt{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



- 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
```

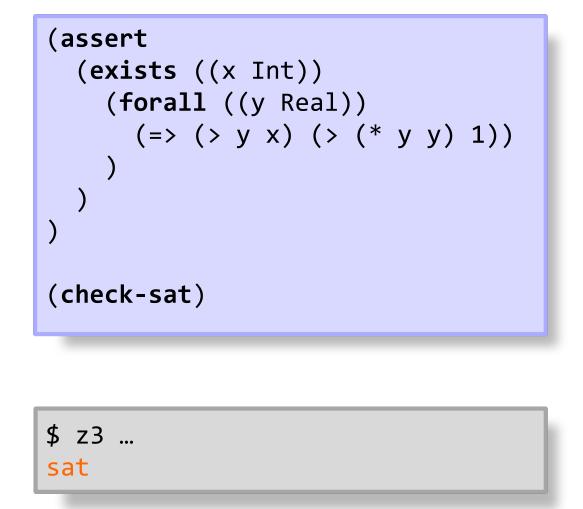
Z3's Theory Reasoning

- 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

Working with quantifiers is non-trivial



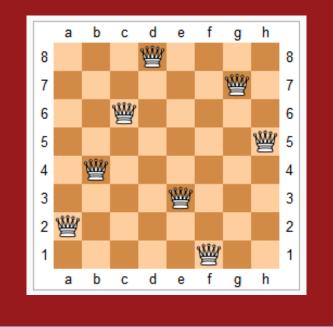
(assert (forall ((x Real)) (exists ((y Real)) (= x (* y y))(check-sat)

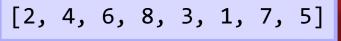




Exercise

- The N-queens problem is to place N-queens on an N x N chess board such that no two queens threaten each other.
- Use Z3 to compute a solution to the N-queens problem for any given N.
- Hints:
 - It is ok to give a solution using just Z3 for a fixed instance but we recommend using Z3Py or another Z3 API such that you can write programs around your Z3 queries.
 - Represent the board using multiple integer variables, e.g. X2 = 5 means the queen is in row 5 in column 2.
 - distinct(1) is a shortcut for stating all elements of list I are pairwise disjoint.
 - You can easily check the diagonals by shifting the queens vertically and then checking the rows.







Wrap-Up

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

1. Compute wp(S, Q)

- 2. Check whether entailment P ==> WP(S, Q) is valid
 - Check satisfiability of negation: P && !WP(S, Q)

 - sat → model explains why { P } S { Q } is not valid

→ ask SMT solver

→ last lecture

What next?

