

COMP4161: Advanced Topics in Software Verification



based on slides by J. Blanchette, L. Bulwahn and T.

Nipkow June Andronick, Christine Rizkallah, Miki Tanaka, Johannes Åman Pohjola T3/2019

**CSIRO** 

### Content



→ Intro & motivation, getting started

→ Foundations & Principles

 Lambda Calculus, natural deduction [1,2]Higher Order Logic, Isar (part 1)

[4]

Term rewriting

→ Proof & Specification Techniques

 Inductively defined sets, rule induction [5]  $[6, 7^b]$ 

Datatypes, recursion, induction, Isar (part 2)

 Hoare logic, proofs about programs, invariants [8]

C verification

[9]

Practice, questions, exam prep

 $[10^{c}]$ 

<sup>a</sup>a1 due; <sup>b</sup>a2 due; <sup>c</sup>a3 due

### **Overview**



#### Automatic Proof and Disproof

→ Sledgehammer: automatic proofs

→ Quickcheck: counter example by testing

→ Nipick: counter example by SAT

Based on slides by Jasmin Blanchette, Lukas Bulwahn, and Tobias Nipkow (TUM).

### **Automation**



Dramatic improvements in fully automated proofs in the last 2 decades.

- → First-order logic (ATP): Otter, Vampire, E, SPASS
- → Propositional logic (SAT): MiniSAT, Chaff, RSat
- → SAT modulo theory (SMT): CVC3, Yices, Z3

### The key:

Efficient reasoning engines, and restricted logics.

### **Automation in Isabelle**



- 1980s rule applications, write ML code
- 1990s simplifier, automatic provers (blast, auto), arithmetic
- 2000s embrace external tools, but don't trust them (ATP/SMT/SAT)

## Sledgehammer



#### Sledgehammer:

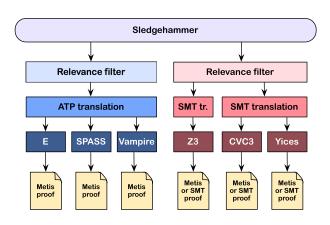
- → Connects Isabelle with ATPs and SMT solvers: E, SPASS, Vampire, CVC3, Yices, Z3
- → Simple invocation:
  - → Users don't need to select or know facts
  - → or ensure the problem is first-order
  - → or know anything about the automated prover
- → Exploits local parallelism and remote servers



**Demo: Sledgehammer** 

# **Sledgehammer Architecture**



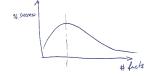


### **Fact Selection**



#### Provers perform poorly if given 1000s of facts.

- → Best number of facts depends on the prover
- → Need to take care which facts we give them
- → Idea: order facts by relevance, give top n to prover (n = 250, 1000, ...)
- → Meng & Paulson method: lightweight, symbol-based filter
- → Machine learning method: look at previous proofs to get a probability of relevance



### From HOL to FOL



**Source:** higher-order, polymorphism, type classes

Target: first-order, untyped or simply-typed

#### → First-order:

- $\rightarrow$  SK combinators,  $\lambda$ -lifting
- → Explicit function application operator

### → Encode types:

- → Monomorphise (generate multiple instances), or
- → Encode polymorphism on term level

### Reconstruction



#### We don't want to trust the external provers.

Need to check/reconstruct proof.

- → Re-find using Metis
  Usually fast and reliable (sometimes too slow)
- → Rerun external prover for trusted replay Used for SMT. Re-runs prover each time!
- → Recheck stored explicit external representation of proof Used for SMT, no need to re-run. Fragile.
- → Recast into structured Isar proof Fast, not always readable.

# Judgement Day (up to 2013)

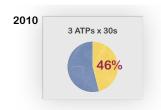


### **Evaluating Sledgehammer:**

- → 1240 goals out of 7 existing theories.
- → How many can sledgehammer solve?
- → 2010: E, SPASS, Vampire (for 5-120s). 46%  $ESV \times 5s \approx V \times 120s$
- → **2011**: Add E-SInE, CVC2, Yices, Z3 (30s). Z3 > V
- → 2012: Better integration with SPASS. 64% SPASS best (small margin)
- → 2013: Machine learning for fact selection. 69% Improves a few percent across provers.

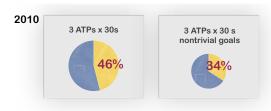
## **Evaluation**





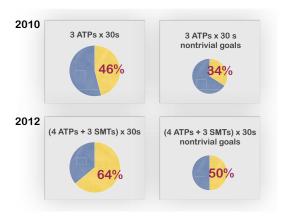
### **Evaluation**





### **Evaluation**





# Judgement Day (2016)



Prover	MePo	MaSh	MeSh	Any selector
CVC4 1.5pre	679	749	783	830
E 1.8	622	601	665	726
SPASS 3.8ds	678	684	739	789
Vampire 3.0	703	698	740	789
veriT 2014post	543	556	590	655
Z3 4.3.2pre	638	668	703	788
Any prover	801	885	919	943

Fig. 15 Number of successful Sledgehammer invocations per prover on 1230 Judgment Day goals

$$919/1230 = 74\%$$

## Sledgehammer rules!



#### **Example application:**

- → Large Isabelle/HOL repository of algebras for modelling imperative programs (Kleene Algebra, Hoare logic, ..., ≈ 1000 lemmas)
- → Intricate refinement and termination theorems
- → Sledgehammer and Z3 automate algebraic proofs at textbook level.

"The integration of ATP, SMT, and Nitpick is for our purposes very very helpful." – G. Struth



# Theorem proving and testing



Testing can show only the presence of errors, but not their absence. (Dijkstra)

Testing cannot prove theorems, but it can refute conjectures!

#### Sad facts of life:

- → Most lemma statements are wrong the first time.
- → Theorem proving is expensive as a debugging technique.

Find counter examples automatically!

## Quickcheck



#### Lightweight validation by testing.

- → Motivated by Haskell's QuickCheck
- → Uses Isabelle's code generator
- → Fast
- → Runs in background, proves you wrong as you type.

## Quickcheck



### Covers a number of testing approaches:

- → Random and exhausting testing.
- → Smart test data generators.
- → Narrowing-based (symbolic) testing.

Creates test data generators automatically.



# Test generators for datatypes



### Fast iteration in continuation-passing-style

**datatype** 
$$\alpha$$
 list = Nil | Cons  $\alpha$  ( $\alpha$  list)

#### **Test function:**

$$\mathsf{test}_{\alpha\ \mathit{list}}\ \mathsf{P}\ =\ \mathsf{P}\ \mathsf{Nil}\ \mathit{andalso}\ \mathsf{test}_{\alpha}\ (\lambda \mathsf{x}.\ \mathsf{test}_{\alpha\ \mathit{list}}\ (\lambda \mathsf{xs.}\ \mathsf{P}\ (\mathsf{Cons}\ \mathsf{x}\ \mathsf{xs})))$$

## Test generators for predicates



distinct  $xs \implies distinct (remove1 \times xs)$ 

#### **Problem:**

Exhaustive testing creates many useless test cases.

#### Solution:

Use definitions in precondition for smarter generator. Only generate cases where distinct xs is true.

test- $distinct_{\alpha}$  list P = P Nil and also  $test_{\alpha}$   $(\lambda x. test$ - $distinct_{\alpha}$  list  $(if x \notin xs then (\lambda xs. P (Cons x xs))$  else True))

Use data flow analysis to figure out which variables must be computed and which generated.

# **Narrowing**



#### Symbolic execution with demand-driven refinement

- → Test cases can contain variables
- → If execution cannot proceed: instantiate with further symbolic terms

#### Pays off if large search spaces can be discarded:

distinct (Cons 1 (Cons 1 x))

False for any x, no further instantiations for x necessary.

### Implementation:

Lazy execution with outer refinement loop. Many re-computations, but fast.

## **Quickcheck Limitations**



#### Only executable specifications!

- → No equality on functions with infinite domain
- → No axiomatic specifications



# **Nitpick**



#### Finite model finder

- → Based on SAT via Kodkod (backend of Alloy prover)
- → Soundly approximates infinite types

## **Nitpick Successes**



- → Algebraic methods
- → C++ memory model
- → Found soundness bugs in TPS and LEO-II

#### Fan mail:

"Last night I got stuck on a goal I was sure was a theorem. After 5–10 minutes I gave Nitpick a try, and within a few secs it had found a splendid counterexample—despite the mess of locales and type classes in the context!"



## We have seen today ...

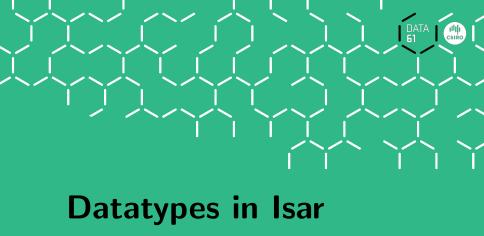


→ Proof: Sledgehammer

→ Counter examples: Quickcheck

→ Counter examples: Nitpick





## Datatype case distinction



```
proof (cases term)
   case Constructor<sub>1</sub>
next
next
   case (Constructor<sub>k</sub> \vec{x})
   \vec{x} ...
ged
       case (Constructor, \vec{x}) \equiv
       fix \vec{x} assume Constructor<sub>i</sub>: "term = Constructor<sub>i</sub> \vec{x}"
```

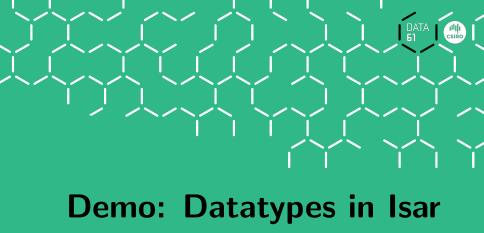
## Structural induction for nat

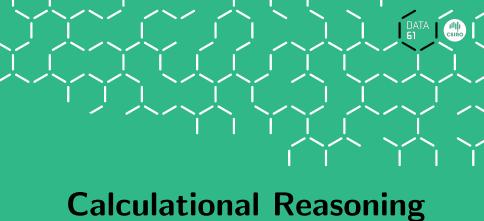


```
show P n
proof (induct n)
                     \equiv let ?case = P 0
  case 0
  show ?case
next
  case (Suc n) \equiv fix n assume Suc: P n
                         let ?case = P (Suc n)
  \cdots n \cdots
  show ?case
qed
```

# Structural induction: $\Longrightarrow$ and $\bigwedge$







### The Goal



Prove:  $x \cdot x^{-1} = 1$ assoc:  $(x \cdot y) \cdot z = x \cdot (y \cdot z)$ using:

left\_inv:  $x^{-1} \cdot x = 1$ left\_one:  $1 \cdot x = x$ 

### The Goal



#### Prove:

$$\begin{array}{lll} x \cdot x^{-1} &= 1 \cdot (x \cdot x^{-1}) & \text{assoc:} & (x \cdot y) \cdot z = x \cdot (y \cdot z) \\ \dots &= 1 \cdot x \cdot x^{-1} & \text{left\_inv:} & x^{-1} \cdot x = 1 \\ \dots &= (x^{-1})^{-1} \cdot x^{-1} \cdot x \cdot x^{-1} & \text{left\_one:} & 1 \cdot x = x \\ \dots &= (x^{-1})^{-1} \cdot (x^{-1} \cdot x) \cdot x^{-1} & \\ \dots &= (x^{-1})^{-1} \cdot 1 \cdot x^{-1} & \\ \dots &= (x^{-1})^{-1} \cdot (1 \cdot x^{-1}) & \\ \dots &= (x^{-1})^{-1} \cdot x^{-1} & \\ \dots &= 1 & \end{array}$$

#### Can we do this in Isabelle?

→ Simplifier: too eager

→ Manual: difficult in apply style

→ Isar: with the methods we know, too verbose

# Chains of equations



#### The Problem

Each step usually nontrivial (requires own subproof) **Solution in Isar:** 

- → Keywords **also** and **finally** to delimit steps
- → ...: predefined schematic term variable, refers to right hand side of last expression
- → Automatic use of transitivity rules to connect steps

## also/finally



```
have "t_0 = t_1" [proof]
also
have "... = t_2" [proof]
also
also
have "\cdots = t_n" [proof]
finally
show P
— 'finally' pipes fact "t_0 = t_n" into the proof
```

```
calculation register "t_0 = t_1" "t_0 = t_2" \vdots "t_0 = t_{n-1}" t_0 = t_n
```

### More about also



- $\rightarrow$  Works for all combinations of =,  $\leq$  and <.
- → Uses all rules declared as [trans].
- → To view all combinations: print\_trans\_rules

# Designing [trans] Rules



have = "
$$I_1 \odot r_1$$
" [proof] also have "...  $\odot r_2$ " [proof] also

### Anatomy of a [trans] rule:

- ightharpoonup Usual form: plain transitivity  $\llbracket l_1 \odot r_1; r_1 \odot r_2 \rrbracket \Longrightarrow l_1 \odot r_2$
- ightharpoonup More general form:  $\llbracket P \ l_1 \ r_1; Q \ r_1 \ r_2; A \rrbracket \Longrightarrow C \ l_1 \ r_2$

### Examples:

- → pure transitivity:  $[a = b; b = c] \implies a = c$
- $\rightarrow$  mixed:  $\llbracket a \leq b; b < c \rrbracket \Longrightarrow a < c$
- $\rightarrow$  substitution:  $\llbracket P \ a; a = b \rrbracket \Longrightarrow P \ b$
- $\rightarrow$  antisymmetry:  $[a < b; b < a] \Longrightarrow False$
- → monotonicity:

$$\llbracket a = f \ b; b < c; \land x \ y. \ x < y \Longrightarrow f \ x < f \ y \rrbracket \Longrightarrow a < f \ c$$

