

### SS 2023



# **Program Verification**

Part 4 – Checking Well-Definedness of Functional Programs

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## Overview

- recall: a functional program is well-defined if
  - it is pattern disjoint,
  - it is pattern complete, and
  - $\hookrightarrow$  is terminating
- well-definedness is prerequisite for standard model, for derived theorems, ...
- task: given a functional program as input, ensure well-definedness
  - known: type-checking algorithm
  - known: algorithm for checking pattern disjointness
  - missing: algorithm for type-inference
  - missing: algorithm for deciding pattern completeness
  - missing: methods to ensure termination
- all of these missing parts will be covered in this chapter

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## **Type-Inference**

structure of functional programs

- data-type definitions
- function definitions: type of new function + defining equations
- not mentioned: type of variables
- in proseminar: work-around via fixed scheme which does not scale
  - singleton characters get type Nat
  - words ending in "s" get type List
- aim: infer suitable type of variables automatically
- example: given FP

append : List  $\times$  List  $\rightarrow$  List append(Cons(x, y), z) = Cons(x, append(y, z))append(Nil, x) = x

we should be able to infer that x : Nat, y : List and z : List in the first equation,whereas x : List in the second equation

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Type-Checking with Implicit Variables

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**Type-Checking with Implicit Variables** 

<pre>Type-Checking algorithm (variable case omitted) typeCheck :: Sig -&gt; Vars -&gt; Term -&gt; Maybe Type typeCheck sigma vars (Var x) = vars x typeCheck sigma vars (Fun f ts) = do   (tysIn,tyOut) &lt;- sigma f   tysTs &lt;- mapM (typeCheck sigma vars) ts    if tysTs == tysIn then return tyOut else Nothing Maybe-type is only one possibility to represent computational results with failure let us abstract from concrete Maybe-type:    introduce new type Check to represent a result or failure    type Check a = Maybe a    function return :: a -&gt; Check a to produce successful results    function to raise a failure    failure :: Check a    failure = Nothing    convenience function: asserting a property    assert :: Bool -&gt; Check ()    assert p = if p them return () else failure    Part 4-Checking Well-Definedness of functional Programs </pre>		<pre>Making Type-Checking More Abstract • original type-checking algorithm    typeCheck :: Sig -&gt; Vars -&gt; Term -&gt; Maybe Type    typeCheck sigma vars (Var x) = vars x    typeCheck sigma vars (Fun f ts) = do       (tysIn,tyOut) &lt;- sigma f       tysTs &lt;- mapM (typeCheck sigma vars) ts       if tysTs == tysIn then return tyOut else Nothing • with new abstract types and functions    typeCheck :: Sig -&gt; Vars -&gt; Term -&gt; Check Type    typeCheck sigma vars (Fun f ts) = do       (tysIn,tyOut) &lt;- sigma f       tysTs &lt;- mapM (typeCheck sigma vars) ts    assert (tysTs == tysIn    return tyOut • advantage: readability, change Check-type easily</pre>	Type-Checking with Implicit Variables
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<ul> <li>type Sig = FSym -&gt;</li> <li>type Vars = Var -&gt;</li> <li>typeCheck takes Σ and</li> <li>we want a function that for and delivers a suitable type inferType :: Sig -&gt;</li> <li>then type-checking an equipyeCheckEqn :: Sig typeCheckEqn sigma (</li> </ul>	<pre>gorithm Vars -&gt; Term -&gt; Check Type &gt; Check ([Type], Type) - Σ &gt; Check Type - V and V and delivers type of term works in the other direction: it gets an intended type as pe for the variables Type -&gt; Term -&gt; Check [(Var,Type)] uation without explicit Vars is possible -&gt; (Term, Term) -&gt; Check ()</pre>		<pre>Type-Inference Algorithm • note: upcoming algorithm only infers types of v    (in polymorphic setting often also type of funct inferType :: Sig -&gt; Type -&gt; Term -&gt; Check inferType sigma ty (Var x) = return [(x,t inferType sigma ty (Fun f ts) = do    (tysIn,tyOut) &lt;- sigma f    assert (length tysIn == length ts)    assert (tyOut == ty)    varsL &lt;- mapM (\ (ty, t) -&gt; inferType s    let vars = nub (concat varsL) nub ret    assert (distinct (map fst vars))    return vars</pre>	<pre>cion symbols is inferred)  [(Var,Type)] y)] igma ty t) (zip tysIn ts)</pre>
<pre>vars &lt;- inferType sigma ty l tyR &lt;- typeCheck sigma (\ x -&gt; lookup x vars) r assert (ty == tyR)</pre>		<pre>distinct :: Eq a =&gt; [a] -&gt; Bool distinct xs = length (nub xs) == length x</pre>	5	
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### Soundness of Type-Inference Algorithm

### properties

- if  $t \in \mathcal{T}(\Sigma, \mathcal{V})_{\tau}$  then infer\_type  $\Sigma \tau t = return (\mathcal{V} \cap \mathcal{V}ars(t))$
- if  $infer_type \Sigma \tau t = return \mathcal{V}$  then
  - $\mathcal{V}$  is well-defined (no conflicting variable assignments) and
  - $t \in \mathcal{T}(\Sigma, \mathcal{V})_{\tau}$
- properties can be shown in similar way to type-checking algorithm, cf. slides 2/35-42
- note that 'if  $t \in \mathcal{T}(\Sigma, \mathcal{V})_{\tau}$  then  $infer\_type \Sigma \ \tau \ t \neq failure$ ' is a property which is not strong enough when performing induction

## Changing the Error Monad

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Changing the Error Monad

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Changing Implementation of Interface

- current interface for error type
  - type Check a = Maybe a
  - function return :: a -> Check a
  - function assert :: Bool -> Check ()
  - function failure :: Check a
  - do-blocks, monadic-functions such as mapM, etc.
- it is actually easy to change to Either-type for errors
  - type Check a = Either String a
  - return, do-blocks and mapM are unchanged, since these are part of generic monad interface
  - functions assert and failure need to be changed, since they now require error messages
    - failure :: String -> Check a
      failure = Left
    - assert :: Bool -> String -> Check ()
    - assert p err = if p then return () else failure err

Weakness of Maybe-Type for Errors

- situation: several functions for checking properties of terms, equations, which can be assembled to check functional programs w.r.t. slides 3/4 (data-type definitions), 3/15 (function definitions) and partly 3/45 (well-definedness)
  - inferType :: Sig -> Type -> Term -> Check [(Var,Type)]
  - typeCheck :: Sig -> Vars -> Term -> Check Type
  - typeCheckEqn :: Sig -> (Term, Term) -> Check ()
- problem: if checks are not successful, we just get result Nothing
- desired: informative error message why a functional program is refused
- possible solution: use more verbose error type than Maybe
   type Check a = Either String a

Changing the Error Monad

Changing Algorithms for Checking Properties

Changing the Error Monad

Changing Algorithms for Checking Properties, Continued • adapting algorithms often only requires additional error messages • example requiring more changes; with type Check a = Maybe a before change (type Check a = Maybe a) typeCheckEqn sigma (Var x, r) = failure typeCheck :: Sig -> Vars -> Term -> Check Type typeCheckEqn sigma (l @ (Fun f \_), r) = do typeCheck sigma vars (Var x) = vars x (\_,ty) <- sigma f typeCheck sigma vars (Fun f ts) = do vars <- inferType sigma ty 1</pre> (tysIn,tyOut) <- sigma f</pre> tyR <- typeCheck sigma (\ x -> lookup x vars) r tysTs <- mapM (typeCheck sigma vars) ts</pre> assert (ty == tyR)assert (tysTs == tysIn) • new version with type Check a = Either String a return tyOut typeCheckEqn sigma (Var x, r) = failure "var as lhs" • after change (type Check a = Either String a) typeCheckEqn sigma (l @ (Fun f ), r) = do typeCheck :: Sig -> Vars -> Term -> Check Type typeCheck sigma vars (Var x) = ... tyR <- typeCheck sigma (\ x -> lookup x vars) r typeCheck sigma vars t@(Fun f ts) = do assert (ty == tyR) "types of lhs and rhs don't match" • problem: lookup produces Maybe, not Either String assert (tysTs == tysIn) (show t ++ " ill-typed") • solution: use maybeToEither :: e -> Maybe a -> Either e a . . . RT (DCS @ UIBK) RT (DCS @ UIBK) Part 4 - Checking Well-Definedness of Functional Programs 13/101 Part 4 - Checking Well-Definedness of Functional Programs

Changing the Error Monad

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import Data.Either.Utils -- for maybeToEither -- import requires MissingH lib; if not installed, define it yourself: -- maybeToEither e Nothing = Left e -- maybeToEither \_ (Just x) = return x typeCheckEqn sigma (Var x, r) = failure "var as lhs" typeCheckEqn sigma (1 @ (Fun f \_), r) = do (\_,ty) <- sigma f vars <- inferType sigma ty 1</pre> tyR <- typeCheck sigma (\ x -> maybeToEither (x ++ " is unknown variable") (lookup x vars)) r assert (ty == tyR) "types of lhs and rhs don't match" RT (DCS @ UIBK) Part 4 - Checking Well-Definedness of Functional Programs

Fixed Type-Checking Algorithm with Error Messages

### **Processing Functional Programs**

#### Processing Functional Programs

		r rocessing i unceronal i rograms			rocessing runctional rograms
			Recall: Data Type D	Definitions	
			• given: set of types	$\mathcal{T}_{\mathcal{Y}}$ , signature $\Sigma = \mathcal{C} \uplus \mathcal{D}$	
Processing Function	onal Programs			nition has the following form	
• aim: write progra	m which takes		• •	$\tau = c_1 : \tau_{1,1} \times \ldots \times \tau_{1,m_1} \to \tau$	
	pgram as input (data type definitions + function definitions)				where
<ul> <li>checks the syr</li> </ul>	ntactic requirements			$  \dots \\   c_n : \tau_{n,1} \times \dots \times \tau_{n,m_n} \to \tau$	
	evant information in some internal representation		• $\tau \notin \mathcal{T}y$		fresh type name
	ecks well-definedness		• $c_1, \ldots, c_n \notin \Sigma$	and $c_i  eq c_j$ for $i \neq j$ freeh as	nd distinct constructor names
	s essential part of a compiler		• each $\tau_{i,j} \in \{\tau\}$		only known types
<ul> <li>program should b</li> </ul>	e easy to verify			at $ au_{i,j} \in \mathcal{T}_{\mathcal{Y}}$ for all $j$	non-recursive constructor
			<ul> <li>effect: add new typ</li> </ul>	e and new constructors	
			• $\mathcal{T}y := \mathcal{T}y \cup \{\tau\}$		
			• $\mathcal{C} := \mathcal{C} \cup \{c_1 : \tau\}$	$\tau_{1,1} \times \ldots \times \tau_{1,m_1} \to \tau, \ldots, c_n : \tau_{n,1} \times \ldots \times \tau_{n,m_1}$	$a_n \to \tau$
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		Processing Functional Programs	<b>Encoding Functional</b>	Programs in Haskell	Processing Functional Programs
			input: unchecked	data-type definitions and function	n definitions
Existing Encoding	of Part 2: Signatures and Terms			= Data Type [(FSym, FSymInfo)]	
type Check <mark>a</mark> =	Maybe a or Either String a		data FunctionDefini		
• •	· · ·		type FunctionalProg		
<b>type</b> Type = String	7 5		([DataDefinition]	, [FunctionDefinition])	
<b>type</b> Var = String			internal represent	ntation	
type FSym = String				ym, FSymInfo)] signatures as lis	
type Vars = Var ->				list of defined	
type FSymInfo = ([				list of construc	
type Sig = FSym ->	> Gneck FSyminio			[erm, Term)] all function eq	
data Term = Var Va	ar   Fun FSym [Term]			Haskell-type; it also stores known	types
uava icim - Val Va	TI TUT TOAT LIETT		<b>data</b> Proginio = Prog	gInfo [Type] Cons Defs Equations	
			checking single (	lata type definition	
			processDataDefiniti	• =	
			ProgInfo -> DataDo	efinition -> Check ProgInfo	
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#### Processing Functional Programs

### Checking a Single Data Definitions

```
processDataDefinition
 (ProgInfo tys cons defs eqs)
 (Data ty newCs)
= do
 assert (not (elem ty tys))
 let newTys = ty : tys
 assert (distinct (map fst newCs))
 assert (all (\ (c,_) -> all (/= c) (map fst (cons ++ defs))) newCs)
 assert (all (\ (_,(tysIn,tyOut)) ->
 tyOut == ty &&
 all (\ ty -> elem ty newTys) tysIn) newCs)
 assert (any
 (\ (_,(tysIn,_)) -> all (/= ty) tysIn) newCs)
 return (ProgInfo newTys (newCs ++ cons) defs eqs)
```

Part 4 - Checking Well-Definedness of Functional Programs

**Checking Several Data Definitions** 

 processing many data definitions can be easily done by using foldM: predefined monadic version of foldl

foldM :: Monad m => (b -> a -> m b) -> b -> [a] -> m b
foldM f e [] = return e
foldM f e (x : xs) = do
 d <- f e x
 foldM f d xs</pre>

processDataDefinition :: ProgInfo -> DataDefinition -> Check ProgInfo processDataDefinition = ... -- previous slide

processDataDefinitions :: ProgInfo -> [DataDefinition] -> Check ProgInfo processDataDefinitions = foldM processDataDefinition

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Processing Functional Programs
                                                                                                                                                          Processing Functional Programs
Checking Function Definitions w.r.t. Slide 3/15
data FunctionDefinition = Function
                  -- name of function
  FSvm
                                                                                        Checking Functional Programs
  FSymInfo
                -- type of function
                                                                                        initialProgInfo = ProgInfo [] [] []
  [(Term,Term)] -- equations
                                                                                        processProgram :: FunctionalProg -> Check ProgInfo
processFunctionDefinition
                                                                                        processProgram (dataDefs, funDefs) = do
  :: ProgInfo -> FunctionDefinition -> Check ProgInfo
                                                                                          pi <- processDataDefinitions initialProgInfo dataDefs</pre>
processFunctionDefinition = ... -- exercise
                                                                                          processFunctionDefinitions pi funDefs
processFunctionDefinitions ::
  ProgInfo -> [FunctionDefinition] -> Check ProgInfo
processFunctionDefinitions =
  foldM processFunctionDefinition
```

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### **Current State**

- processProgram :: FunctionalProg -> Check ProgInfo is Haskell program to check user provided functional programs, whether they adhere to the specification of functional programs w.r.t. slides 3/4 and 3/15
- its functional style using error monads permits to easily verify its correctness
  - no induction required
  - based on assumption that builtin functions behave correctly, e.g., all, any, nub, ...
- missing: checks for well-defined functional programs w.r.t. slide 3/45

## **Checking Pattern Disjointness**

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Checking Pattern Disjointness

### **Deciding Pattern Disjointness**

- program is pattern disjoint if for all  $f: \tau_1 \times \cdots \times \tau_n \to \tau \in \mathcal{D}$  and all  $t_1 \in \mathcal{T}(\mathcal{C})_{\tau_1}, \ldots, t_n \in \mathcal{T}(\mathcal{C})_{\tau_n}$  there is at most one equation  $\ell = r$  in the program, such that  $\ell$  matches  $f(t_1, \ldots, t_n)$
- in proseminar it was proven that pattern disjointness is equivalent to the following condition: for each pair of distinct equations  $\ell_1 = r_1$  and  $\ell_2 = r_2$ ,  $\ell_1$  and a variable renamed variant of  $\ell_2$  do not unify
- key missing part for checking pattern disjointness is an algorithm for unification:

given two terms s and t, decide 
$$\exists \sigma. s\sigma = t\sigma$$

Checking Pattern Disjointness

Unification Algorithm of Martelli and Montanari

- input: unification problem  $U = \{s_1 \stackrel{?}{=} t_1, \dots, s_n \stackrel{?}{=} t_n\}$
- question: is U solvable, i.e., does there exist a solution  $\sigma$ , a substitution satisfying  $\forall i \in \{1, ..., n\}$ .  $s_i \sigma = t_i \sigma$
- two different kinds of output:
  - unification problem in solved form:

$$\{x_1 \stackrel{?}{=} v_1, \dots, x_m \stackrel{?}{=} v_m\}$$
 with distinct  $x_j$ 's

solved forms can be interpreted as substitution

$$\sigma(x) = \begin{cases} v_i, & \text{if } x = x_i \\ x, & \text{otherwise} \end{cases}$$

and this  $\sigma$  will be solution of U

- $\perp$ , indicating that U is not solvable
- $\bullet\,$  algorithm itself is build via one-step relation  $\rightsquigarrow$  which is applied as long as possible

Unification Algorithm of Martelli and Montanari, continued

Checking Pattern Disjointness

- input: unification problem  $U = \{s_1 \stackrel{?}{=} t_1, \dots, s_n \stackrel{?}{=} t_n\}$
- output: solution of U via solved form or  $\perp$ , indicating unsolvability
- algorithm applies  $\rightsquigarrow$  as long as possible;  $\rightsquigarrow$  is defined as

$$U \cup \{t \stackrel{?}{=} t\} \rightsquigarrow U \tag{delete}$$

$$U \cup \{f(u_1, \dots, u_k) \stackrel{?}{=} f(v_1, \dots, v_k)\} \rightsquigarrow U \cup \{u_1 \stackrel{?}{=} v_1, \dots, v_k \stackrel{?}{=} v_k\} \quad (\text{decompose})$$

$$U \cup \{f(u_1, \dots, u_k) \stackrel{?}{=} g(v_1, \dots, v_\ell)\} \rightsquigarrow \bot, \text{ if } f \neq g \lor k \neq \ell$$
 (clash)

$$U \cup \{f(\dots) \stackrel{?}{=} x\} \rightsquigarrow U \cup \{x \stackrel{?}{=} f(\dots)\}$$
(swap)

$$U \cup \{x \stackrel{?}{=} f(...)\} \rightsquigarrow \bot, \text{ if } x \in \mathcal{V}ars(f(...))$$
 (occurs check)

$$U \cup \{x \stackrel{?}{=} t\} \rightsquigarrow U\{x/t\} \cup \{x \stackrel{?}{=} t\},$$
  
if  $x \notin Vars(t)$  and x occurs in U (eliminate)

notation  $U\{x/t\}$ : apply substitution  $\{x/t\}$  on all terms in U (lhs + rhs)

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Correctness of Unification Algorithm
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- we only state properties (proofs: see term rewriting lecture)
  - $\rightarrow$  terminates
  - normal form of  $\rightsquigarrow$  is  $\perp$  or a solved form
  - whenever  $U \rightsquigarrow V$ , then U and V have same solutions
  - in total: to solve unification problem U
    - determine some normal form V of U
    - if  $V = \bot$  then U is unsolvable
    - otherwise, V represents a substitution that is a solution to U
- note that  $\rightsquigarrow$  is not confluent

• 
$$\{x \stackrel{?}{=} y, y \stackrel{?}{=} x\} \stackrel{x,y}{\rightsquigarrow} \{x \stackrel{?}{=} y, y \stackrel{?}{=} y\} \rightsquigarrow \{x \stackrel{?}{=} y\}$$
  
•  $\{x \stackrel{?}{=} y, y \stackrel{?}{=} x\} \stackrel{y/x}{\rightsquigarrow} \{x \stackrel{?}{=} x, y \stackrel{?}{=} x\} \rightsquigarrow \{y \stackrel{?}{=} x\}$ 

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Checking Pattern Disjointness

Correctness of	an Implementation of a (Unification	a) Algorithm	A Concrete Impleme	enting of the Unification	Algorithm	Checking Pattern [
<ul> <li>any concrete</li> <li>preference</li> </ul>	implementation will make choices e of rules		subst :: Var -> Term -> subst <mark>x t</mark> = applySubst (	Term -> Term (\ y -> if y == x then t els	e Var <mark>y</mark> )	
<ul> <li>represent</li> </ul>	of pairs from $U$ ation of sets $U$ lection in quicksort)		unify :: [(Term, Term)] unify <mark>u</mark> = unifyMain <mark>u</mark> []			
<ul> <li>(order of</li> </ul>	edges in graph-/tree-traversals)		unifyMain :: [(Term, Ter	m)] -> [(Var,Term)] -> Chec	k [(Var, Term)]	
•			unifyMain [] v = return	v	return solved form	
	ensure that implementation is sound		unifyMain ((Fun f ts, Fu assert (f == g && leng		clash	
solution: refi	nement proof		unifyMain (zip <mark>ts ss</mark> +	-+ u) v	decompose	
• aim: reu	se correctness of abstract algorithm ( $\rightsquigarrow$ )		unifyMain ((Fun f ts, x)			
	lation between representations in concrete and t before and done informally)	abstract algorithm (this was called	unifyMain ((x, Fun f t unifyMain ((Var x, t) :		swap	
	t concrete algorithm has less behaviour, i.e., even can be related to some result of (non-determined to some result of the source of the sourc	· · · · · · · · · · · · · · · · · · ·	if Var x == t then uni else do		delete	
0	clear separation between		assert (not (x`elem	varsTerm t))	occurs check	
<ul> <li>soundness of abstract algorithm</li> <li>soundness of implementation</li> </ul>	(solves unification problems) (implements abstract algorithm)	· ·	(subst x t l, subst x t r)) (y, s) -> (y, subst x t s))			
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Checking Pattern Disjointness

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#### Checking Pattern Disjointness

### Notes on Implementation

- it is non-trivial to prove soundness of implementation, since there are several differences w.r.t.  $\rightsquigarrow$ 
  - $\bullet \ unify\_main$  takes two parameters u and v
    - these represent one unification problem  $u \cup v$
  - rule-application is not tried on v, only on u
    - we need to know that v is in normal form w.r.t.  $\leadsto$
  - in (occurs check)-rule, the algorithm has no test that rhs is function application
    we need to show that this will follow from other conditions
  - in (elimination)-rule, the algorithm substitutes only in rhss of  $\boldsymbol{v}$
  - $\hfill \bullet$  we need to know that substituting in lhss of v has no effect
  - in (elimination)-rule, the algorithm does not check that x occurs in remaining problem
    - we need to check that consequences don't harm

- relation  $\sim$  formally aligns parameters of concrete algorithm (u and v) with parameters of abstract algorithm (U);  $\sim$  also includes invariants of implementation
  - set converts list to set, we identify  $s \stackrel{?}{=} t$  with (s, t)
  - $(u,v) \sim U$  iff
    - $U = set \ u \cup set \ v$ ,
    - set v is in normal form w.r.t.  $\rightsquigarrow$  (notation: set  $v \in NF(\rightsquigarrow)$ ), and
    - for all  $(x,t) \in set \ v$ : x does not occur in u
- since alignment between concrete and abstract parameters is specified formally, alignment properties of auxiliary functions can also be made formal

```
• set (x : xs) = \{x\} \cup set xs

• set (xs ++ ys) = set xs \cup set ys

• set (zip \ [x_1, \dots, x_n] \ [y_1, \dots, y_n]) = \{(x_1, y_1), \dots, (x_n, y_n)\}

• set (map \ f \ [x_1, \dots, x_n]) = \{f \ x_1, \dots, f \ x_n\}

• subst x \ t \ s = s\{x/t\}

• \dots
```

these properties can be proven formally and also be applied formally (although we don't do it in the upcoming proof)

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Checking Pattern Disjointness

Soundness via Refinement: Main Statement

- define set\_maybe Nothing =  $\bot$ , set\_maybe (Just w) = set w
- property: whenever  $(u, v) \sim U$  and  $unify\_main \ u \ v = res$  then  $U \rightsquigarrow^! set\_maybe \ res$
- once property is established, we can prove that implementation solves unification problems
  - assume input u, i.e., invocation of unify u which yields result res
  - hence,  $unify_main \ u \ [] = res$
  - moreover,  $(u, []) \sim set \; u$  by definition of  $\sim$
  - via property conclude set  $u \rightsquigarrow$ ! set\_maybe res
  - at this point apply correctness of →: set\_maybe res is the correct answer to the unification problem set u

**Proving the Refinement Property** 

- property P(u, v, U):  $(u, v) \sim U \land unify\_main \ u \ v = res \longrightarrow U \rightsquigarrow^! set\_maybe \ res$
- $\bullet \ (u,v) \sim U \longleftrightarrow U = set \ u \cup set \ v \wedge set \ v \in NF(\leadsto) \wedge \forall (x,t) \in set \ v. \ x \notin \mathcal{V}ars(u)$
- we prove the property P(u, v, U) by induction on u and v w.r.t. the algorithm for arbitrary U, i.e., we consider all left-hand sides and can assume that the property holds for all recursive calls;

induction w.r.t. algorithm gives partial correctness result (assumes termination)

- in the lecture, we will cover a simple, a medium, and the hardest case
- case 1 (arguments [] and v):
  - we have to prove P([], v, U), so assume
    - (\*)  $([],v) \sim U$  and
    - (\*\*)  $unify_main [] v = res$
  - from (\*) conclude U = set v and  $set v \in NF(\rightsquigarrow)$
  - from (\*\*) conclude res = Just v and hence,  $set\_maybe res = set v$
  - we have to show  $U \rightsquigarrow^! set_maybe res$ , i.e., set  $v \rightsquigarrow^! set v$  which is satisfied since set  $v \in NF(\rightsquigarrow)$

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Checking Pattern Disjointness

- P(u, v, U):  $(u, v) \sim U \wedge unify\_main \ u \ v = res \longrightarrow U \rightsquigarrow^! set\_maybe \ res$
- $(u,v) \sim U \longleftrightarrow U = set \ u \cup set \ v \land set \ v \in NF(\rightsquigarrow) \land \forall (x,t) \in set \ v. \ x \notin Vars(u)$

case 2 (arguments (f(ts), g(ss)) : u and v)

- we have to prove P((f(ts), g(ss)) : u, v, U), so assume
  - $(*) \ \left( (f(ts),g(ss)):u,v \right) \sim U$  and
- (\*\*) unify\_main ((f(ts), g(ss)): u) v = res
- consider sub-cases
  - $\neg(f = g \land length \ ts = length \ ss)$ :
    - from (\*\*) conclude  $set\_maybe \ res = \bot$
    - from (\*) conclude  $f(ts) \stackrel{?}{=} g(ss) \in U$  and hence  $U \rightsquigarrow \bot$  by (clash)
    - consequently,  $U \rightsquigarrow^! set\_maybe res$
  - $f = g \land length \ ts = length \ ss$ :
    - from (\*\*) conclude  $res = unify\_main ((f(ts), g(ss)) : u) v = unify\_main (zip ts ss ++ u) v$
    - from (\*) and alignment for zip and ++ conclude U = {f(ts) ? g(ss)} ∪ set u ∪ set v and hence U → set (zip ts ss ++ u) ∪ set v =: V by (decompose)
    - we get  $P(zip \ ts \ ss \ ++ \ u, v, V)$  as IH;  $(zip \ ts \ ss \ ++ \ u, v) \sim V$  follows from (\*), so  $U \rightsquigarrow V \rightsquigarrow^! set\_maybe \ res$

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Part 4 – Checking Well-Definedness of Functional Programs

- P(u, v, U):  $(u, v) \sim U \wedge unify\_main \ u \ v = res \longrightarrow U \rightsquigarrow^! set\_maybe \ res$
- $\bullet \ (u,v) \sim U \longleftrightarrow U = set \ u \cup set \ v \wedge set \ v \in \mathit{NF}(\rightsquigarrow) \land \forall (x,t) \in set \ v. \ x \notin \mathit{Vars}(u)$

case 4 (arguments (x, t) : u and v)

- we have to prove P((x,t):u,v,U), so assume
  - (\*)  $((x,t):u,v) \sim U$  and
- (\*\*) unify\_main ((x,t):u) v = res
- consider sub-cases (where the red part is not triggered by structure of algorithm)
  - $x \neq t \land x \notin \mathcal{V}ars(t) \land x$  occurs in set  $u \cup set v$ :
    - define  $u' = map \ (\lambda(l, r). \ (subst \ x \ t \ l, subst \ x \ t \ r)) \ u$
    - define  $v' = map \ (\lambda(y, s). \ (y, subst \ x \ t \ s)) \ v$
    - define  $V = (set \ u \cup set \ v)\{x/t\} \cup \{x \stackrel{?}{=} t\}$
    - from (\*\*) conclude  $res = unify\_main ((x,t):u) v = unify\_main u' ((x,t):v')$
    - from IH conclude P(u', (x,t): v', V) and hence,  $(u', (x,t): v') \sim V \longrightarrow V \rightsquigarrow^! set\_maybe restriction for the two sets the set of the two sets the$
    - for proving  $U \rightsquigarrow^! set\_maybe\ res$  it hence suffices to show  $(u', (x, t) : v') \sim V$  and  $U \rightsquigarrow V$
    - $U \stackrel{(*)}{=} \{x \stackrel{?}{=} t\} \cup set \ u \cup set \ v \rightsquigarrow (set \ u \cup set \ v)\{x/t\} \cup \{x/t\} = V$  by (eliminate) because of preconditions
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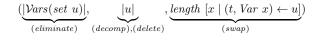
Checking Pattern Disjointness

- $(u,v) \sim U \longleftrightarrow U = set \ u \cup set \ v \land set \ v \in NF(\rightsquigarrow) \land \forall (x,t) \in set \ v. \ x \notin Vars(u)$
- case 4 (arguments (x, t) : u and v)
  - we have to prove P((x,t):u,v,U), so assume (\*)  $((x,t):u,v) \sim U$  and ... and consider sub-case  $x \neq t \land x \notin Vars(t) \land x$  occurs in set  $u \cup set v$ :
    - define  $u' = map (\lambda(l, r). (subst x t l, subst x t r)) u$
    - define  $v' = map(\lambda(y, s). (y, subst x t s)) v$
    - define  $V = (set \ u \cup set \ v) \{x/t\} \cup \{x \stackrel{?}{=} t\}$
    - we still need to show  $(u', (x, t) : v') \sim V$
    - since (\*) holds, we know  $\forall (y, s) \in set \ v. \ x \neq y$
    - hence,  $v' = map (\lambda(y, s). (subst x t y, subst x t s)) v$
    - so,  $V = (set \ u)\{x/t\} \cup \{x \stackrel{?}{=} t\} \cup (set \ v)\{x/t\} = set \ u' \cup set \ ((x,t):v')$
    - we show  $\forall (y,s) \in set \ ((x,t):v'). \ y \notin Vars(u')$  as follows:  $x \notin Vars(u')$  since  $x \notin Vars(t)$ ; and if  $(y,s) \in set \ v'$ , then  $(y,s') \in set \ v$  for some s' and by (\*) we conclude  $y \notin Vars((x,t):u)$ ; thus,  $y \notin Vars((set \ u)\{x/t\}) = Vars(u')$
    - we finally show set ((x,t): v') ∈ NF(~): so, assume to the contrary that some step is applicable; by the shape of set ((x,t): v') we know that the step can only be (eliminate), (delete) or (occurs check); all of these cases result in a contradiction by using the available facts

Checking Pattern Disjointness

### **Proving the Refinement Property**

- remaining cases: similar, cf. exercises
- summary
  - non-trivial implementation of abstract unification algorithm  $\rightsquigarrow$
  - optimizations required additional invariants, encoded in refinement relation
  - proof of correctness can be done formally
    - induction + case analysis proof uses mostly the structure of the Haskell code; exception: case analysis on "x occurs in set u ∪ set v"
    - most cases can easily be solved, after having identified suitable invariants
       fully reuse correctness of →
  - we only proved partial correctness
  - termination of implementation: consider lexicographic measure



#### **Checking Pattern Completeness**

- reminder: program is pattern complete, if for all  $f : \tau_1 \times \ldots \times \tau_n \to \tau \in \mathcal{D}$  and all  $t_i \in \mathcal{T}(\mathcal{C})_{\tau_i}$  there is some lhs that matches  $f(t_1, \ldots, t_n)$
- idea of abstract algorithm
  - a pattern problem is a set P of pairs (t, L) where
    - t is a term, representing the set of all its constructor ground instances
    - L is a set of left-hand sides that potentially match instances of t
  - initially,  $P = \{(f(x_1, \dots, x_n), \text{set of all lhss of } f \text{-equations}) \mid f \in \mathcal{D}\}$
  - whenever some left-hand side  $\ell \in L$  cannot match any instance of t anymore, it can be removed
  - whenever L becomes empty, then no instance of t can be matched
  - whenever all constructor ground instances of t are matched by L, then (t,L) can be removed from  ${\cal P}$
  - $\bullet\,$  when P becomes empty, pattern completeness should be guaranteed
  - $\bullet\,$  if none of the above is applicable, we instantiate t
- initial task: think about exact statement, what kind of property of pattern problem we are investigating (similar to definition of solution of unification problem)

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Part 4 – Checking Well-Definedness of Functional Programs

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Checking Pattern Completeness

#### Checking Pattern Completeness

**Semantics of Pattern Problems** 

- in the following algorithm and proofs, we always consider type-correct terms and substitutions w.r.t.  $\Sigma = C \cup D$ , but do not mention this explicitly
- a pattern problem is a set P of pairs (t, L) consisting of a term t and a set of terms L

**Checking Pattern Completeness** 

- P is complete if for all  $(t, L) \in P$  and all constructor ground substitutions  $\sigma$  there is some  $\ell \in L$  that matches  $t\sigma$
- obviously,  $P=\varnothing$  is complete
- ${\ }{\ }$  we define  ${\ }{\ }{\ }$  as additional pattern problem, which is not complete
- define  $L_{init,f}$  as the set of all lhss of f-equations of the program
- define  $P_{init} = \{(f(x_1, \ldots, x_n), L_{init,f}) \mid f \in \mathcal{D}\}$
- $\bullet\,$  consequence: program is pattern complete iff  $P_{init}$  is complete

**Deciding Completeness of Pattern Problems** 

- we develop abstract algorithm that is similar to abstract unification algorithm, it is defined via a one step relation  $\rightarrow$  that transforms pattern problems into equivalent simpler problems
- it uses the matching algorithm of slides  $3/23\mathchar`-29$  (with detailed error results) as auxiliary algorithm
- $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P$ , if  $\ell$  matches t (match)
- $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P \cup \{(t, L)\}, \text{ if } match \ \ell \ t \ \text{clashes}$  (clash)
- $P \cup \{(t, \emptyset)\} \rightarrow \bot$  (fail)
- $P \cup \{(t,L)\} \rightarrow P \cup \{(t\sigma_1,L),\ldots,(t\sigma_n,L)\}$ , if (split)
  - $\ell \in L$  and  $match \ \ell \ t$  results in fun-var-conflict with variable x
  - the type of x is  $\tau$
  - $\tau$  has n constructors  $c_1, \ldots, c_n$
  - $\sigma_i = \{x/c_i(x_1, \dots, x_k)\}$  where k is the arity of  $c_i$  and the  $x_i$ 's are distinct fresh variables

Decidir

Example

Example

consider

data Bool = True : Bool | False : Bool  

$$\ell_1 := \operatorname{conj}(\operatorname{True}, \operatorname{True}) = \dots$$
  
 $\ell_2 := \operatorname{conj}(\operatorname{False}, y) = \dots$   
 $\ell_3 := \operatorname{conj}(x, \operatorname{False}) = \dots$ 

then we have

$$\begin{split} P_{init} &= \{(\operatorname{conj}(x_1, x_2), \{\ell_1, \ell_2, \ell_3\})\} \\ &\stackrel{(s)}{=} \{(\operatorname{conj}(\operatorname{True}, x_2), \{\ell_1, \ell_2, \ell_3\}), (\operatorname{conj}(\operatorname{False}, x_2), \{\ell_1, \ell_2, \ell_3\})\} \\ &\stackrel{(c)}{=} \{(\operatorname{conj}(\operatorname{True}, x_2), \{\ell_1, \ell_3\}), (\operatorname{conj}(\operatorname{False}, x_2), \{\ell_1, \ell_2, \ell_3\})\} \\ &\stackrel{(c)}{=} \{(\operatorname{conj}(\operatorname{True}, x_2), \{\ell_1, \ell_3\}), (\operatorname{conj}(\operatorname{False}, x_2), \{\ell_2, \ell_3\})\} \\ &\stackrel{(m)}{=} \{(\operatorname{conj}(\operatorname{True}, \operatorname{True}), \{\ell_1, \ell_3\})\} \\ &\stackrel{(s)}{=} \{(\operatorname{conj}(\operatorname{True}, \operatorname{True}), \{\ell_1, \ell_3\}), (\operatorname{conj}(\operatorname{True}, \operatorname{False}), \{\ell_1, \ell_3\})\} \\ &\stackrel{(m)}{=} \{(\operatorname{conj}(\operatorname{True}, \operatorname{False}), \{\ell_1, \ell_3\})\} \\ &\stackrel{(m)}{=} \{(\operatorname{conj}(\operatorname{True}, \operatorname{False}), \{\ell_1, \ell_3\})\} \\ &\stackrel{(m)}{=} \emptyset \\ \end{split}$$
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consider

data Bool = True : Bool | False : Bool

$$l := \mathsf{conj}(\mathsf{True}, \mathsf{True}) = \dots$$
  
 $\ell_2 := \mathsf{conj}(\mathsf{False}, y) = \dots$ 

then we have

$$\begin{split} P_{init} &= \{(\text{conj}(x_1, x_2), \{\ell_1, \ell_2\})\} \\ &\stackrel{(s)}{\leftarrow} \{(\text{conj}(\text{True}, x_2), \{\ell_1, \ell_2\}), (\text{conj}(\text{False}, x_2), \{\ell_1, \ell_2\})\} \\ &\stackrel{(c)}{\leftarrow} \{(\text{conj}(\text{True}, x_2), \{\ell_1\}), (\text{conj}(\text{False}, x_2), \{\ell_1, \ell_2\})\} \\ &\stackrel{(m)}{\leftarrow} \{(\text{conj}(\text{True}, x_2), \{\ell_1\})\} \\ &\stackrel{(s)}{\leftarrow} \{(\text{conj}(\text{True}, \text{True}), \{\ell_1\}), (\text{conj}(\text{True}, \text{False}), \{\ell_1\})\} \\ &\stackrel{(m)}{\leftarrow} \{(\text{conj}(\text{True}, \text{False}), \{\ell_1\})\} \\ &\stackrel{(m)}{\leftarrow} \{(\text{conj}(\text{True}, \text{False}), \{\ell_1\})\} \\ &\stackrel{(c)}{\leftarrow} \{(\text{conj}(\text{True}, \text{False}), \emptyset)\} \\ &\stackrel{(f)}{\leftarrow} \perp \end{split} \end{split}$$
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 $\ell_1$ 

Partial Correctness of  $\rightarrow$ 

- definition: P is complete if for all  $(t, L) \in P$  and all constructor ground substitutions  $\sigma$  there is some  $\ell \in L$  that matches  $t\sigma$
- theorem: whenever  $P \rightarrow Q$ , then P is complete iff Q is complete
- corollary: if P →\* Ø then P is complete, and if P →\* ⊥ then P is not complete
- proof of theorem
  - (match):  $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P$ , if  $\ell$  matches t
    - we only have to show that  $\{(t,\{\ell\}\cup L)\}$  is complete, i.e., for all constructor ground substitutions  $\sigma$  there must be some  $\ell'\in\{\ell\}\cup L$  that matches  $t\sigma$
    - since  $\ell$  matches t, we know that  $t=\ell\gamma$  for some substitution  $\gamma$
    - consequently  $t\sigma = (\ell\gamma)\sigma = \ell(\gamma\sigma)$ , i.e.,  $\ell$  matches  $t\sigma$  and obviously  $\ell \in \{\ell\} \cup L$
  - (fail):  $P \cup \{(t, \emptyset)\} \rightharpoonup \bot$ 
    - both matching problems are not complete:  $\perp$  by definition, and for  $(t, \varnothing)$  there obviously isn't any  $\ell \in \varnothing$  which matches  $t\sigma$

Partial Correctness of  $\rightharpoonup$ , continued

- definition: P is complete if for all  $(t, L) \in P$  and all constructor ground substitutions  $\sigma$  there is some  $\ell \in L$  that matches  $t\sigma$
- proof continued
  - (clash):  $P \cup \{(t, \{\ell\} \cup L)\} \rightharpoonup P \cup \{(t, L)\}$ , if match  $\ell$  t clashes
    - if suffices to show that  $\ell$  cannot match any instance of t, i.e.,  $match \ \ell \ (t\sigma)$  will always fail
    - to this end we require an auxiliary property of the matching algorithm
    - for a matching problem M, define  $M\sigma=\{(\ell,r\sigma)\mid (\ell,r)\in M\},$  i.e., where  $\sigma$  is applied on rhss, and  $\bot\sigma=\bot$
    - lemma: whenever M is transformed into M' by rule (decompose) or (clash), then  $M\sigma$  is transformed into  $M'\sigma$  by the same rule
    - hence, since  $match~\ell~t$  clashes, we conclude that  $match~\ell~(t\sigma)$  clashes

### Partial Correctness of $\rightarrow$ , final part

Checking Pattern Completeness

(fail)

- definition: P is complete if for all  $(t, L) \in P$  and all constructor ground substitutions  $\sigma$  there is some  $\ell \in L$  that matches  $t\sigma$
- proof continued Correctness of  $\rightarrow$ , Missing Parts • (split):  $P \cup \{(t, L)\} \rightarrow P \cup \{(t\sigma_1, L), \dots, (t\sigma_n, L)\}$ , where  $x : \tau$ ,  $\tau$  has constructors  $c_1, \ldots, c_n$  and  $\sigma_i = \{x/c_i(x_1, \ldots, x_k)\}$  for fresh  $x_i$  already proven • we only consider one direction of the proof: we assume that the rhs of  $\rightarrow$  is complete and • if  $P \rightarrow^* \emptyset$  then P is complete prove that the lhs is complete • if  $P \rightharpoonup^* \bot$  then P is not complete • to this end, consider an arbitrary constructor ground substitution  $\sigma$  and we have to show that • open: termination of  $\rightarrow$  $t\sigma$  is matched by some element of L • since  $\sigma$  is constructor ground, we know  $\sigma(x) = c_i(t_1, \ldots, t_k)$  for some constructor  $c_i$  and open: can → get stuck? constructor ground terms  $t_1, \ldots, t_k$ • define  $\gamma(y) = \begin{cases} t_j, \\ \ddots \end{cases}$ if  $y = x_i$  $\sigma(y)$ , otherwise •  $\gamma$  is well-defined since the  $x_i$ 's are distinct •  $\gamma$  is a constructor ground substitution •  $t\sigma = t\sigma_i\gamma$  since the  $x_i$ 's are fresh • since  $(t\sigma_i, L)$  is an element of the rhs of  $\rightarrow$  and the assumed completeness, we conclude that there is some element of L that matches  $(t\sigma_i)\gamma$  and consequently, also  $t\sigma$ RT (DCS @ UIBK) Part 4 - Checking Well-Definedness of Functional Programs RT (DCS @ UIBK) 49/101 Part 4 - Checking Well-Definedness of Functional Programs

→ Cannot Get Stuck

- $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P$ , if  $\ell$  matches t (match)
- $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P \cup \{(t, L)\}, \text{ if } match \ \ell \ t \text{ results in clash}$  (clash)
- $P \cup \{(t, \emptyset)\} \rightarrow \bot$
- P ∪ {(t,L)} → P ∪ {(tσ<sub>1</sub>,L),...,(tσ<sub>n</sub>,L)}, if (split)
   ℓ ∈ L and match ℓ t results in fun-var-conflict with variable x and ...
- lemma: whenever P is in normal form w.r.t.  $\rightarrow$  and for all  $(t, L) \in P$  and all  $\ell \in L$ , the
- In  $\ell$  is linear, then  $P \in \{\emptyset, \bot\}$
- proof by contradiction
  - assume P is such a normal form,  $P \notin \{ \varnothing, \bot \}$
  - hence,  $(t,L)\in P$  for some t and L
  - since (fail) is not applicable,  $L \neq \varnothing$ , i.e.,  $\ell \in L$  for some  $\ell$
  - as (match) is not applicable,  $match \ \ell \ t$  must fail
  - ${\ }^{\bullet}$  as (clash) and (split) are not applicable the failure can only be a var-clash
  - however, a var-clash cannot occur since  $\ell$  is linear

Termination of  $\rightharpoonup$ 

- $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P$ , if  $\ell$  matches t (match) •  $P \cup \{(t, \{\ell\} \cup L)\} \rightarrow P \cup \{(t, L)\}$ , if match  $\ell$  t clashes (clash)
- $P \cup \{(t, \emptyset)\} \rightarrow \bot$  (fail)
- $P \cup \{(t,L)\} \rightharpoonup P \cup \{(t\sigma_1,L),\ldots,(t\sigma_n,L)\}$ , if (split)
  - $\ell \in L$  and  $match \ \ell \ t$  results in fun-var-conflict with variable x and  $\dots$
- define  $|\ell-t|$  as a measure of difference of  $\ell$  and t
  - $\bullet \ |\ell-x| = {\rm number \ of \ function \ symbols \ in \ } \ell$

• 
$$|f(\ell_1, ..., \ell_n) - f(t_1, ..., t_n)| = \sum_i |\ell_i - t_i|$$

- $|\ell t| = 0$ , in all other cases
- map each pattern problem P to multiset  $\left\{\sum_{\ell \in L} |\ell t| \mid (t, L) \in P\right\}$
- this multiset decreases in (match) and (split) and is not increased in (clash) (multiset decrease:  $M \cup N >^{mul} M \cup N'$  if  $N \neq \emptyset$  and  $\forall y \in N'$ .  $\exists x \in N. x > y$ )
- since (clash) on its own also terminates,  $\rightharpoonup$  must terminate

Checking Pattern Completeness

Implementing  $\rightharpoonup$ 

- implementing  $\rightharpoonup$  naively has the disadvantage that the matching algorithm is executed from scratch every time
- an improved algorithm might therefore interleave both algorithms
- a pair (t, {l<sub>1</sub>,..., l<sub>n</sub>}) in the abstract algorithm corresponds to an entry {{(t, l<sub>1</sub>)},..., {(t, l<sub>n</sub>)}} in the improved algorithm, where each {(t, l<sub>i</sub>)} corresponds to an initial matching problem: does l<sub>i</sub> match t?
- the improved algorithm is described by the following inference rules
  - $P \cup \{\emptyset\} \rightarrow' \bot$  (fail)
  - $P \cup \{\{\emptyset\} \cup p\} \rightarrow' P$  (match-empty)
  - $P \cup \{\{\{(t,x)\} \cup mp\} \cup p\} \rightharpoonup' P \cup \{\{mp\} \cup p\}$  (match-var)

**Termination – Dependency Pairs** 

- $P \cup \{\{\{(f(\ldots), g(\ldots))\} \cup mp\} \cup p\} \rightarrow P \cup \{p\}, \text{ if } f \neq g$  (clash)
- $P \cup \{\{\{(f(t_1, \ldots), f(\ell_1, \ldots))\} \cup mp\} \cup p\} \rightharpoonup P \cup \{\{\{(t_1, \ell_1), \ldots\} \cup mp\} \cup p\} \text{ (decompose)}\}$
- $P \cup \{\{\{(x,\ell)\} \cup mp\} \cup p\} \rightharpoonup' P \cup \{(\{\{\{(x,\ell)\} \cup mp\} \cup p\}) \sigma_i \mid \sigma_i = \{x/c_i(x_1, \dots, x_{n_i})\}$  (split)

where the substitutions are only applied on the left components of pairs of terms and  $\ell \notin \mathcal{V}$ 

• theorem:  $\rightharpoonup'$  is an implementation of  $\rightharpoonup$ , and  $\rightharpoonup'$  is terminating

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Termination - Dependency Pairs

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#### Termination of Programs

Summary on Pattern Completeness

• pattern completeness of functional programs is decidable:

• proof required additional properties of matching algorithm

• termination of  $\rightarrow$  was shown via multisets and a dedicated measure

• termination proof was tricky, definitely required human interaction

• in contrast: upcoming part is on automated termination proving

• partial correctness was proven via invariant of  $\rightarrow$ 

program is pattern complete iff  $P_{init} \rightharpoonup^! \varnothing$ 

- the question of termination is a famous problem
  - Turing showed that "halting problem" is undecidable
  - halting problem
    - question: does program (Turing machine) terminate on given input
    - problem is semi-decidable: positive instances can always be identified
    - algorithm: just simulate the program and then say "yes, terminates"
- we here consider universal termination, i.e., termination on all inputs
- universal termination is not even semi-decidable
- despite theoretical limits: often termination can be proven automatically

**Termination of Functional Programs** 

#### Termination - Dependency Pairs

- for termination, we mainly consider functional programs which are pattern-disjoint; hence,  $\hookrightarrow$  is confluent
- consequence: it suffices to prove innermost termination, i.e., the restriction of  $\hookrightarrow$  such that arguments  $t_i$  will be fully evaluated before evaluating a function invocation  $f(t_1,\ldots,t_n)$
- example without confluence

f(True, False, x) = f(x, x, x) $\mathsf{f}(\ldots,\ldots,x) = x$ (all other cases) coin = Truecoin = False

- both f and coin terminate if seen as separate programs
- program is innermost terminating, but not terminating in general

$$f(\mathsf{True},\mathsf{False},\mathsf{coin}) \hookrightarrow f(\mathsf{coin},\mathsf{coin},\mathsf{coin}) \hookrightarrow^2 f(\mathsf{True},\mathsf{False},\mathsf{coin}) \hookrightarrow \dots$$

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Subterm Relation and Innermost Evaluation

• define  $\triangleright$  as the strict subterm relation and  $\triangleright$  as its reflexive closure

$$\frac{t_i \triangleright s}{F(t_1, \dots, t_n) \triangleright t_i} \qquad \qquad \frac{t_i \triangleright s}{F(t_1, \dots, t_n) \triangleright s}$$

• innermost evaluation  $\stackrel{i}{\rightarrow}$  is defined similar to one-step evaluation  $\rightarrow$ 

$$\begin{array}{c} \underbrace{s_i \stackrel{ \ \, \leftarrow \ \, \rightarrow \ \, t_i}{F(s_1,\ldots,s_i,\ldots,s_n) \stackrel{ \ \, \leftarrow \ \, \rightarrow \ \, F(s_1,\ldots,t_i,\ldots,s_n)}} \ \text{rewrite in context} \\ \underbrace{\ell = r \ \text{is equation in program} \quad \forall s \lhd \ell\sigma. \ s \in NF(\hookrightarrow)}_{\ell\sigma \stackrel{ \ \, \leftarrow \ \, \rightarrow \ \, r\sigma}} \ \text{root step} \end{array}$$

example

$$f(True, False, coin) \not\rightarrow f(coin, coin, coin)$$

since coin  $\triangleleft$  f(True, False, coin) and coin  $\notin NF(\hookrightarrow)$ 

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Termination – Dependency Pairs

#### Termination - Dependency Pairs

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Strong Normalization

• relation  $\succ$  is strongly normalizing, written  $SN(\succ)$ , if there is no infinite sequence

$$a_1 \succ a_2 \succ a_3 \succ \dots$$

- strong normalization is other notion for termination
- strong normalization of a relation is equivalent to soundness of induction principle w.r.t. that relation;

the following two conditions are equivalent

- $SN(\succ)$
- $\forall P. \ (\forall x. \ (\forall y. \ x \succ y \longrightarrow P \ y) \longrightarrow P \ x) \longrightarrow (\forall x. \ P \ x)$
- equivalence shows why it is possible to perform induction w.r.t. algorithm for terminating programs

**Termination Analysis with Dependency Pairs** 

- aim: prove  $SN(\stackrel{i}{\hookrightarrow})$
- only reason for potential non-termination: recursive calls
- for each recursive call of equation  $f(t_1, \ldots, t_n) = \ell = r \ge f(s_1, \ldots, s_n)$  build one dependency pair with fresh (constructor) symbol  $f^{\sharp}$ :

$$f^{\sharp}(t_1,\ldots,t_n) \to f^{\sharp}(s_1,\ldots,s_n)$$

define *DP* as the set of all dependency pairs

• example program for Ackermann function has three dependency pairs

$$ack(Zero, y) = Succ(y)$$

$$ack(Succ(x), Zero) = ack(x, Succ(Zero))$$

$$ack(Succ(x), Succ(y)) = ack(x, ack(Succ(x), y))$$

$$ack^{\sharp}(Succ(x), Zero) \rightarrow ack^{\sharp}(x, Succ(Zero))$$

$$ack^{\sharp}(Succ(x), Succ(y)) \rightarrow ack^{\sharp}(x, ack(Succ(x), y))$$

$$ack^{\sharp}(Succ(x), Succ(y)) \rightarrow ack^{\sharp}(Succ(x), y)$$

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#### Termination – Dependency Pairs

Example of Evaluation and Chain minus(x, Zero) = xTermination Analysis with Dependency Pairs, continued minus(Succ(x), Succ(y)) = minus(x, y)• dependency pairs provide characterization of termination  $\operatorname{div}(\operatorname{Zero}, \operatorname{Succ}(y)) = \operatorname{Zero}$ • definition: let  $P \subseteq DP$ : a *P*-chain is a possible infinite sequence  $\mathsf{div}(\mathsf{Succ}(x),\mathsf{Succ}(y)) = \mathsf{Succ}(\mathsf{div}(\mathsf{minus}(x,y),\mathsf{Succ}(y)))$ minus<sup> $\sharp$ </sup>(Succ(x), Succ(y))  $\rightarrow$  minus<sup> $\sharp$ </sup>(x, y)  $s_1\sigma_1 \rightarrow t_1\sigma_1 \stackrel{i}{\leftrightarrow} s_2\sigma_2 \rightarrow t_2\sigma_2 \stackrel{i}{\leftrightarrow} s_3\sigma_3 \rightarrow t_2\sigma_3 \stackrel{i}{\leftrightarrow} \dots$  $\operatorname{div}^{\sharp}(\operatorname{Succ}(x),\operatorname{Succ}(y)) \to \operatorname{div}^{\sharp}(\operatorname{minus}(x,y),\operatorname{Succ}(y))$ such that all  $s_i \to t_i \in P$  and all  $s_i \sigma_i \in NF(\hookrightarrow)$ • example innermost evaluation •  $s_i \sigma_i \rightarrow t_i \sigma_i$  represent the "main" recursive calls that may lead to non-termination div(Succ(Zero), Succ(Zero)) •  $t_i \sigma_i \stackrel{i}{\hookrightarrow} s_{i+1} \sigma_{i+1}$  corresponds to evaluation of arguments of recursive calls  $\stackrel{i}{\hookrightarrow}$  Succ(div(minus(Zero, Zero), Succ(Zero))) • theorem:  $SN(\stackrel{i}{\hookrightarrow})$  iff there is no infinite DP-chain  $\stackrel{i}{\hookrightarrow}$  Succ(div(Zero, Succ(Zero))) advantage of dependency pairs  $\stackrel{i}{\hookrightarrow}$  Succ(Zero) • in infinite chain, non-terminating recursive calls are always applied at the root and its (partial) representation as DP-chain • simplifies termination analysis  $div^{\sharp}(Succ(Zero), Succ(Zero))$  $\rightarrow div^{\sharp}(minus(Zero, Zero), Succ(Zero))$  $\stackrel{i}{\hookrightarrow}^* \operatorname{div}^{\sharp}(\operatorname{Zero}, \operatorname{Succ}(\operatorname{Zero}))$ RT (DCS @ UIBK) Part 4 - Checking Well-Definedness of Functional Programs RT (DCS @ UIBK) Part 4 - Checking Well-Definedness of Functional Programs 61/101

Termination - Dependency Pairs

### **Proving Termination**

- global approaches
  - try to find one termination argument that no infinite chain exists
- iterative approaches
  - identify dependency pairs that are harmless, i.e., cannot be used infinitely often in a chain
  - remove harmless dependency pairs from set of dependency pairs
  - until no dependency pairs are left
- we focus on iterative approaches, in particular those that are incremental
  - incremental: a termination proof of some function stays valid
  - if later on other functions are added to the program
  - incremental termination proving is not possible in general case (for non-confluent programs), consider coin-example on slide 57

**Termination – Subterm Criterion** 

#### Termination – Subterm Criterion

A First Termination Technique – The Subterm Criterion

- the subterm criterion works as follows
  - let  $P \subseteq DP$
  - choose  $f^{\sharp}$ , a symbol of arity n
  - choose some argument position  $i \in \{1, \ldots, n\}$
  - demand  $s_i \succeq t_i$  for all  $f^{\sharp}(s_1, \ldots, s_n) \to f^{\sharp}(t_1, \ldots, t_n) \in P$
  - define  $P_{\triangleright} = \{ f^{\sharp}(s_1, \dots, s_n) \to f^{\sharp}(t_1, \dots, t_n) \in P \mid s_i \triangleright t_i \}$
  - then for proving absence of infinite *P*-chains it suffices to prove absence of infinite  $P \setminus P_{\triangleright}$ -chains, i.e., one can remove all pairs in  $P_{\triangleright}$
- observations

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- easy to test: just find argument position i such that each i-th argument of all  $f^{\sharp}$ -dependency pairs decreases w.r.t.  $\supseteq$  and then remove all strictly decreasing pairs
- incremental method: adding other dependency pairs for  $g^{\sharp}$  later on will have no impact

Part 4 - Checking Well-Definedness of Functional Programs

- can be applied iteratively
- fast, but limited power

- Subterm Criterion Example
- consider a program with the following set of dependency pairs

$$\operatorname{ack}^{\sharp}(\operatorname{Succ}(x),\operatorname{Zero}) \to \operatorname{ack}^{\sharp}(x,\operatorname{Succ}(\operatorname{Zero}))$$
 (1)

- $\operatorname{ack}^{\sharp}(\operatorname{Succ}(x),\operatorname{Succ}(y)) \to \operatorname{ack}^{\sharp}(x,\operatorname{ack}(\operatorname{Succ}(x),y))$  (2)
- $\operatorname{ack}^{\sharp}(\operatorname{Succ}(x),\operatorname{Succ}(y)) \to \operatorname{ack}^{\sharp}(\operatorname{Succ}(x),y)$  (3)
- $\operatorname{minus}^{\sharp}(\operatorname{Succ}(x), \operatorname{Succ}(y)) \to \operatorname{minus}^{\sharp}(x, y)$ (4)

$$\operatorname{div}^{\sharp}(\operatorname{Succ}(x),\operatorname{Succ}(y)) \to \operatorname{div}^{\sharp}(\operatorname{minus}(x,y),\operatorname{Succ}(y))$$
(5)

$$\mathsf{plus}^{\sharp}(\mathsf{Succ}(x), y) \to \mathsf{plus}^{\sharp}(y, x)$$
 (6)

- it is easy to remove (4) by choosing any argument of minus<sup>#</sup>
- we can remove (1) and (2) by choosing argument 1 of  $ack^{\sharp}$
- afterwards we can remove (3) by choosing argument 2 of  $\mathsf{ack}^\sharp$
- it is not possible to remove any of the remaining dependency pairs (5) and (6) by the subterm criterion
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Subterm Criterion – Soundness Proof

Termination – Subterm Criterion

- assume the chosen parameters in the subterm criterion are  $f^{\sharp}$  and i
- it suffices to prove that there is no infinite chain

 $s_1\sigma_1 \to t_1\sigma_1 \stackrel{i}{\hookrightarrow} s_2\sigma_2 \to t_2\sigma_2 \stackrel{i}{\hookrightarrow} s_3\sigma_3 \to t_3\sigma_3 \stackrel{i}{\hookrightarrow} \dots$ 

- such that all  $s_j \to t_j \in P$ , all  $s_j$  and  $t_j$  have  $f^{\sharp}$  as root and there are infinitely many  $s_j \to t_j \in P_{\triangleright}$ ; perform proof by contradiction
- hence all  $s_j \to t_j$  are of the form  $f^{\sharp}(s_{j,1}, \ldots, s_{j,n}) \to f^{\sharp}(t_{j,1}, \ldots, t_{j,n})$
- from condition  $s_{j,i} \ge t_{j,i}$  of criterion conclude  $s_{j,i}\sigma_j \ge t_{j,i}\sigma_j$ and if  $s_j \to t_j \in P_{\triangleright}$  then  $s_{j,i} \triangleright t_{j,i}$  and thus  $s_{j,i}\sigma_j \triangleright t_{j,i}\sigma_j$
- we further know  $t_{j,i}\sigma_j \stackrel{\cdot}{\hookrightarrow}^* s_{j+1,i}\sigma_{j+1}$  since  $f^{\sharp}$  is a constructor
- this implies  $t_{j,i}\sigma_j = s_{j+1,i}\sigma_{j+1}$  since  $t_{j,i}\sigma_j \in NF(\hookrightarrow)$  as  $t_{j,i}\sigma_j \leq s_{j,i}\sigma_j < f^{\sharp}(s_{j,1}\sigma_j, \ldots, s_{j,n}\sigma_j) = s_j\sigma_j \in NF(\hookrightarrow)$
- obtain an infinite sequence with infinitely many  $\triangleright$ ; this is a contradiction to  $SN(\triangleright)$

$$s_{1,i}\sigma_1 \succeq t_{1,i}\sigma_1 = s_{2,i}\sigma_2 \succeq t_{2,i}\sigma_2 = s_{3,i}\sigma_3 \succeq t_{3,i}\sigma_3 = \dots$$

Part 4 - Checking Well-Definedness of Functional Programs

**Termination – Size-Change Principle** 

### The Size-Change Principle

- the size-change principle abstracts decreases of arguments into size-change graphs
- size-change graph
  - let  $f^{\sharp}$  be a symbol of arity n
  - a size-change graph for  $f^{\sharp}$  is a bipartite graph G = (V, W, E)
  - the nodes are  $V = \{1_{in}, \dots, n_{in}\}$  and  $W = \{1_{out}, \dots, n_{out}\}$
  - E is a set of directed edges between in- and out-nodes labelled with  $\succ$  or  $\succeq$
  - the size-change graph G of a dependency pair  $f^{\sharp}(s_1, \ldots, s_n) \to f^{\sharp}(t_1, \ldots, t_n)$  defines E as follows
    - $i_{in} \stackrel{\succ}{\to} j_{out} \in E$  whenever  $s_i \triangleright t_j$  (strict decrease)
    - $i_{in} \stackrel{\succeq}{\rightarrow} j_{out} \in E$  whenever  $s_i = t_j$  (weak decrease)
- in representation, in-nodes are on the left, out-nodes are on the right, and subscripts are omitted

- Example Size-Change Graphs
- consider the following dependency pairs; they include permutations that cannot be solved by the subterm criterion

$$f^{\sharp}(\mathsf{Succ}(x), y) \to f^{\sharp}(x, \mathsf{Succ}(x))$$
(7)

$$f^{\sharp}(x, \mathsf{Succ}(y)) \to f^{\sharp}(y, x) \tag{8}$$

 $G_{(8)}: \quad 1 \underset{\swarrow}{\succ} 1$ 

• obtain size-change graphs that contain more information than just the size-decrease in one argument, as we had in subterm criterion



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Termination – Size-Change Principle

**Multigraphs and Concatenation** 

- graphs can be glued together, tracing size-changes in chains, i.e., subsequent dependency pairs
- definition: let G be a set of size-change graphs for the same symbol f<sup>#</sup>; then the set of multigraphs for f<sup>#</sup> is defined as follows
  - every  $G \in \mathcal{G}$  is a multigraph
  - whenever there are multigraphs G<sub>1</sub> and G<sub>2</sub> with edges E<sub>1</sub> and E<sub>2</sub> then also the concatenated graph G = G<sub>1</sub> • G<sub>2</sub> is a multigraph; here, the edges of E of G are defined as
    - if  $i \to j \in E_1$  and  $j \to k \in E_2$ , then  $i \to k \in E$
    - if at least one of the edges  $i \to j$  and  $j \to k$  is labeled with  $\succ$  then  $i \to k$  is labeled with  $\succ$ , otherwise with  $\succeq$
    - if the previous rules would produce two edges  $i\xrightarrow{\succ}k$  and  $i\xrightarrow{\succeq}k$ , then only the former is added to E
- a multigraph G is maximal if  $G = G \cdot G$
- since there are only finitely many possible sets of edges, the set of multigraphs is finite and can easily be computed

Example – Multigraphs

• consider size-change graphs

 $\begin{array}{ccc} G_{(7)}: & 1 \xrightarrow{\succ} 1 \\ & & & \\ & & & \\ & & & 2 \end{array}$ 

• this leads to three maximal multigraphs

$$\begin{array}{cccc} G_{(7)} \bullet G_{(8)} : 1 \stackrel{\succ}{\longrightarrow} 1 & G_{(8)} \bullet G_{(7)} : 1 & 1 & G_{(8)} \bullet G_{(8)} : 1 \stackrel{\succ}{\rightarrow} 1 \\ & & & & \\ 2 & 2 & 2 & 2 & 2 \\ \end{array}$$

• and a non-maximal multigraph

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Termination – Size-Change Principle

Size-Change Termination

- instead of multigraphs, one can also glue two graphs  $G_1$  and  $G_2$  by just identifying the out-nodes of  $G_1$  with the in-nodes of  $G_2$ , defined as  $G_1 \circ G_2$ ; in this way it is also possible to consider an infinite sequence of graphs  $G_1 \circ G_2 \circ G_3 \circ \ldots$
- example:

$$\begin{array}{ccc} G_{(7)} \circ G_{(8)} \circ G_{(8)} \circ G_{(7)} : & 1 \stackrel{\succ}{\searrow} 1 & 1 \stackrel{\succ}{\swarrow} 1 \\ & 2 & 2 & 2 & 2 & 2 \\ \end{array}$$

- definition: a set  $\mathcal{G}$  of size-change graph is size-change terminating iff for every infinite concatenation of graphs of  $\mathcal{G}$  there is a path with infinitely many  $\xrightarrow{\succ}$ -edges
- theorem: let P be a set of dependency pairs for symbol  $f^{\sharp}$  and  $\mathcal{G}$  be the corresponding size-change graphs; if  $\mathcal{G}$  is size-change terminating, then there is no infinite P-chain
- the proof is mostly identical to the one of the subterm criterion

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### Proof of Theorem

- the direction that size-change termination implies the property on maximal multigraphs can be done in a straight-forward way
- the other direction is much more advanced and relies upon Ramsey's theorem in its infinite version

**Deciding Size-Change Termination** 

- definition: a set G of size-change graph is size-change terminating iff for every infinite concatenation of graphs of  $\mathcal{G}$  there is a path with infinitely many  $\xrightarrow{\succ}$ -edges
- checking size-change termination directly is not possible
- still, size-change termination is decidable
- theorem: let  $\mathcal{G}$  be a set of size-change graphs; the following two properties are equivalent 1.  $\mathcal{G}$  is size-change terminating
  - 2. every maximal multigraph of  $\mathcal{G}$  contains an edge  $i \xrightarrow{\succ} i$
- although the above theorem only gives rise to an EXPSPACE-algorithm, size-change termination is in PSPACE: in fact, size-change termination is PSPACE-complete
- despite the high theoretical complexity class, for sets of size-change graphs arising from usual algorithms, the number of multigraphs is rather low

Proof of Theorem: Easy Direction (1. implies 2.)

- assume that  $\mathcal{G}$  is size-change terminating, and consider any maximal graph G
- since G is a multigraph, it can be written as  $G = G_1 \cdot \ldots \cdot G_n$  with each  $G_i \in \mathcal{G}$
- consider infinite graph  $G_1 \circ \ldots \circ G_n \circ G_1 \circ \ldots \circ G_n \circ \ldots$
- because of size-change termination, this graph contains path with infinitely many  $\rightarrow$ -edges
- hence  $G \circ G \circ \ldots$  also has a path with infinitely many  $\stackrel{\succ}{\rightarrow}$ -edges
- on this path some index *i* must be visited infinitely often
- hence there is a path of length k such that  $G \circ G \circ \ldots \circ G$  (k-times) contains a path from the leftmost argument i to the rightmost argument i with at least one  $\stackrel{\succ}{\rightarrow}$ -edge
- consequently  $G \cdot G \cdot \ldots \cdot G$  (k-times) contains an edge  $i \xrightarrow{\succ} i$
- by maximality,  $G = G \cdot G \cdot \ldots \cdot G$ , and thus G contains an edge  $i \xrightarrow{\succ} i$

#### Termination – Size-Change Principle

### Ramsey's Theorem

• definition: given set X and  $n \in \mathbb{N}$ , we define  $X^{(n)}$  as the set of all subsets of X of size n; formally:

$$X^{(n)} = \{ Z \mid Z \subseteq X \land |Z| = n \}$$

- Ramsey's Theorem Infinite Version
  - let  $n \in \mathbb{N}$

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- let C be a finite set of colors
- let X be an infinite set
- let c be a coloring of the size n sets of X, i.e.,  $c:X^{(n)}\to C$
- theorem: there exists an infinite subset  $Y\subseteq X$  such that all size n sets of Y have the same color

- consider some arbitrary infinite graph  $G_0 \circ G_1 \circ G_2 \circ \ldots$
- for n < m define  $G_{n,m} = G_n \bullet \ldots \bullet G_{m-1}$
- by Ramsey's theorem there is an infinite set  $I \subseteq \mathbb{N}$  such that  $G_{n,m}$  is always the same graph G for all  $n,m \in I$  with n < m
  - $(n = 2, C = multigraphs, X = \mathbb{N}, c(\{n, m\}) = G_{min\{n,m\},max\{n,m\}})$
- G is maximal: for  $n_1 < n_2 < n_3$  with  $\{n_1, n_2, n_3\} \subseteq I$ , we have  $G_{n_1,n_3} = G_{n_1} \cdot \ldots \cdot G_{n_2-1} \cdot G_{n_2} \cdot \ldots \cdot G_{n_3-1} = G_{n_1,n_2} \cdot G_{n_2,n_3}$ , and thus  $G = G \cdot G$
- by assumption, G contains edge  $i \xrightarrow{\succ} i$
- let  $I = \{n_1, n_2, \ldots\}$  with  $n_1 < n_2 < \ldots$  and obtain

$$G_0 \circ G_1 \circ \dots$$
  
=  $G_0 \circ \dots \circ G_{n_1-1} \circ G_{n_1} \circ \dots \circ G_{n_2-1} \circ G_{n_2} \circ \dots \circ G_{n_3-1} \circ \dots$   
~  $G_0 \circ \dots \circ G_{n_1-1} \circ G$   $\circ G$   $\circ \dots$ 

so that edge  $i \xrightarrow{\succ} i$  of G delivers path with infinitely many  $\xrightarrow{\succ}$ -edges

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Termination – Size-Change Principle

## Proof of Ramsey's Theorem

- Ramsey's Theorem Infinite Version
  - let  $n \in \mathbb{N}$
  - let C be a finite set of colors
  - let X be an infinite set
  - let c be a coloring of the size n sets of X, i.e.,  $c: X^{(n)} \to C$
  - theorem: there exists an infinite subset  $Y\subseteq X$  such that all size n sets of Y have the same color
- proof of Ramsey's theorem is interesting
  - it is simple, in that it only uses standard induction on n with arbitrary c and X
  - it is complex, in that it uses a non-trivial construction in the step-case, in particular applying the IH infinitely often
- base case n = 0 is trivial, since there is only one size-0 set: the empty set

**Proof of Ramsey's Theorem – Step Case** n = m + 1• define  $X_0 = X$ 

- pick an arbitrary element  $a_0$  of  $X_0$
- define  $Y_0 = X_0 \setminus \{a_0\}$ ; define coloring  $c': Y_0^{(m)} \to C$  as  $c'(Z) = c(Z \cup \{a_0\})$
- IH yields infinite subset  $X_1 \subseteq Y_0$  such that all size m sets of  $X_1$  have the same color  $c_0$  w.r.t. c'
- hence,  $c(\{a_0\} \cup Z) = c_0$  for all  $Z \in X_1^{(m)}$
- next pick an arbitrary element  $a_1$  of  $X_1$  to obtain infinite set  $X_2 \subseteq X_1 \setminus \{a_1\}$  such that  $c(\{a_1\} \cup Z) = c_1$  for all  $Z \in X_2^{(m)}$
- by iterating this obtain elements  $a_0, a_1, a_2, \ldots$ , colors  $c_0, c_1, c_2 \ldots$  and sets  $X_0, X_1, X_2, \ldots$  satisfying the above properties
- since C is finite there must be some color d in the infinite list  $c_0, c_1, \ldots$  that occurs infinitely often; define  $Y = \{a_i \mid c_i = d\}$
- Y has desired properties since all size n sets of Y have color d: if  $Z \in Y^{(n)}$  then Z can be written as  $\{a_{i_1}, \ldots, a_{i_n}\}$  with  $i_1 < \ldots < i_n$ ; hence,  $Z = \{a_{i_1}\} \cup Z'$  with  $Z' \in X^{(m)}_{i_1+1}$ , i.e.,  $c(Z) = c_{i_1} = d$
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Termination - Size-Change Principle

### Summary of Size-Change Principle

- size-change principle abstracts dependency pairs into set of size-change graphs
- if no critical graph exists (multigraph without edge  $i \xrightarrow{\succ} i$ ), termination is proven
- soundness relies upon Ramsey's theorem
- subsumes subterm criterion
- still no handling of defined symbols in dependency pairs as in

## $\operatorname{div}^{\sharp}(\operatorname{Succ}(x), \operatorname{Succ}(y)) \to \operatorname{div}^{\sharp}(\operatorname{minus}(x, y), \operatorname{Succ}(y))$

## **Termination – Reduction Pairs**

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Termination - Reduction Pairs

### **Reduction Pairs**

• recall definition: *P*-chain is sequence

$$s_1\sigma_1 \to t_1\sigma_1 \stackrel{i}{\hookrightarrow} s_2\sigma_2 \to t_2\sigma_2 \stackrel{i}{\hookrightarrow} s_3\sigma_3 \to t_3\sigma_3 \stackrel{i}{\hookrightarrow} \dots$$

such that all  $s_i \to t_i \in P$  and all  $s_i \sigma_i \in NF(\hookrightarrow)$ 

- previously we used  $\triangleright$  on  $s_i \rightarrow t_i$  to ensure decrease  $s_i \sigma_i \triangleright t_i \sigma_i$
- previously we used  $s_i \sigma \in NF(\hookrightarrow)$  and  $\triangleright$  to turn  $\stackrel{i}{\hookrightarrow}^*$  into =
- now generalize  $\triangleright$  to strongly normalizing relation  $\succ$
- now demand  $\ell \succeq r$  for equations to ensure decrease  $t_i \sigma_i \succeq s_{i+1} \sigma_{i+1}$
- definition: reduction pair  $(\succ, \succeq)$  is pair of relations such that
  - $SN(\succ)$
  - $\succeq$  is transitive
  - $\succ$  and  $\succeq$  are compatible:  $\succ \circ \succeq \subseteq \succ$
  - both  $\succ$  and  $\succeq$  are closed under substitutions:  $s \succeq t \longrightarrow s\sigma \succeq t\sigma$   $\succeq$  is closed under contexts:  $s \succeq t \longrightarrow F(\dots, s, \dots) \succeq F(\dots, t, \dots)$

  - note:  $\succ$  does not have to be closed under contexts

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### **Applying Reduction Pairs**

• recall definition: *P*-chain is sequence

 $s_1\sigma_1 \rightarrow t_1\sigma_1 \stackrel{i}{\leftrightarrow} s_2\sigma_2 \rightarrow t_2\sigma_2 \stackrel{i}{\leftrightarrow} s_3\sigma_3 \rightarrow t_3\sigma_3 \stackrel{i}{\leftrightarrow} \dots$ 

- such that all  $s_i \to t_i \in P$  and all  $s_i \sigma \in NF(\hookrightarrow)$
- demand  $s \succeq t$  for all  $s \to t \in P$  to ensure  $s_i \sigma_i \succeq t_i \sigma_i$
- demand  $\ell \succeq r$  for all equations to ensure  $t_i \sigma_i \succeq s_{i+1} \sigma_{i+1}$
- define  $P_{\succ} = \{s \to t \in P \mid s \succ t\}$
- effect: pairs in  $P_{\succ}$  cannot be applied infinitely often and can therefore be removed
- theorem: if there is an infinite P-chain, then there also is an infinite  $P \setminus P_{\succ}$ -chain

Termination - Reduction Pairs

### Example

### • remaining termination problem

$$\begin{split} \min(x, \mathsf{Zero}) &= x \\ \min(\mathsf{Succ}(x), \mathsf{Succ}(y)) &= \min(x, y) \\ \operatorname{div}(\mathsf{Zero}, \mathsf{Succ}(y)) &= \mathsf{Zero} \\ \operatorname{div}(\mathsf{Succ}(x), \mathsf{Succ}(y)) &= \operatorname{Succ}(\operatorname{div}(\min(x, y), \mathsf{Succ}(y))) \\ \operatorname{div}^{\sharp}(\mathsf{Succ}(x), \mathsf{Succ}(y)) &\to \operatorname{div}^{\sharp}(\min(x, y), \mathsf{Succ}(y)) \end{split}$$

• constraints

$$\begin{split} \minus(x, \mathsf{Zero}) \succeq x \\ \minus(\mathsf{Succ}(x), \mathsf{Succ}(y)) \succeq \minus(x, y) \\ \operatorname{div}(\mathsf{Zero}, \mathsf{Succ}(y)) \succeq \mathsf{Zero} \\ \operatorname{div}(\mathsf{Succ}(x), \mathsf{Succ}(y)) \succeq \mathsf{Succ}(\operatorname{div}(\minus(x, y), \mathsf{Succ}(y))) \\ \operatorname{div}^{\sharp}(\mathsf{Succ}(x), \mathsf{Succ}(y)) \succ \operatorname{div}^{\sharp}(\minus(x, y), \mathsf{Succ}(y)) \end{split}$$

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Termination – Reduction Pairs

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Applying Reduction Pairs with Usable Equations

• recall definition: P-chain is sequence

$$s_1\sigma_1 \to t_1\sigma_1 \stackrel{i}{\hookrightarrow} s_2\sigma_2 \to t_2\sigma_2 \stackrel{i}{\hookrightarrow} s_3\sigma_3 \to t_3\sigma_3 \stackrel{i}{\hookrightarrow} \dots$$

such that all  $s_i \to t_i \in P$  and all  $s_i \sigma \in NF(\hookrightarrow)$ 

- choose a symbol  $f^{\sharp}$  and define  $P_{f^{\sharp}} = \{s \to t \in P \mid root \ s = f^{\sharp}\}$
- demand  $s \succeq t$  for all  $s \to t \in P_{f^{\sharp}}$
- demand  $\ell \succsim r$  for all  $l=r \in \mathcal{U}$  where  $\mathcal U$  are usable equations w.r.t.  $P_{f^{\sharp}}$
- define  $P_{\succ} = \{s \rightarrow t \in P_{f^{\sharp}} \mid s \succ t\}$
- effect: pairs in  $P_\succ$  cannot be applied infinitely often and can therefore be removed
- theorem: if there is an infinite P-chain, then there also is an infinite  $P \setminus P_{\succ}$ -chain

Termination - Reduction Pairs

Termination - Reduction Pairs

$$\operatorname{div}^{\sharp}(\operatorname{Succ}(x),\operatorname{Succ}(y)) \to \operatorname{div}^{\sharp}(\operatorname{minus}(x,y),\operatorname{Succ}(y))$$

- requiring  $\ell \gtrsim r$  for all program equations  $\ell = r$  is quite demanding • not incremental, i.e., adding other functions later will invalidate proof
  - not necessary, i.e., argument evaluation in example only requires minus
- definition: the usable equations  $\mathcal{U}$  w.r.t. a set P are program equations of those symbols that occur in P or that are invoked by (other) usable equations; formally, let  $\mathcal{E}$  be set of equations of program, let root (f(...)) = f; then  $\mathcal{U}$  is defined as

$$\frac{s \to t \in P \quad t \succeq u \quad \ell = r \in \mathcal{E} \quad root \ u = root \ \ell}{\ell = r \in \mathcal{U}}$$
$$\frac{\ell' = r' \in \mathcal{U} \quad r' \trianglerighteq u \quad \ell = r \in \mathcal{E} \quad root \ u = root \ \ell}{\ell = r \in \mathcal{U}}$$

• observation whenever  $t_i \sigma_i \stackrel{{}_{\leftarrow}}{\rightarrow}^* s_{i+1} \sigma_{i+1}$  in chain, then only usable equations of  $\{s_i \rightarrow t_i\}$ can be used RT (DCS © UIBK) Part 4 - Checking Well-Definedness of Functional Programs 86/101

• remaining termination problem

$$\begin{split} \min(x, \mathsf{Zero}) &= x \\ \min(\mathsf{Succ}(x), \mathsf{Succ}(y)) &= \min(x, y) \\ \operatorname{div}(\mathsf{Zero}, \mathsf{Succ}(y)) &= \mathsf{Zero} \\ \operatorname{div}(\mathsf{Succ}(x), \mathsf{Succ}(y)) &= \mathsf{Succ}(\operatorname{div}(\min(x, y), \mathsf{Succ}(y))) \\ \operatorname{div}^{\sharp}(\mathsf{Succ}(x), \mathsf{Succ}(y)) &\to \operatorname{div}^{\sharp}(\min(x, y), \mathsf{Succ}(y)) \end{split}$$

constraints

$$\begin{split} \min(x, \mathsf{Zero}) &\succsim x \\ \min(\mathsf{Succ}(x), \mathsf{Succ}(y)) &\succsim \min(x, y) \\ \mathsf{div}^{\sharp}(\mathsf{Succ}(x), \mathsf{Succ}(y)) &\succ \mathsf{div}^{\sharp}(\min(x, y), \mathsf{Succ}(y)) \end{split}$$

 because of usable equations, applying reduction pairs becomes incremental: new function definitions won't increase usable equations of DPs of previously defined equations

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## **Remaining Problem**

• given constraints

$$\begin{split} \min_{x} & (x, \text{Zero}) \succeq x \\ \min_{x} & (\text{Succ}(x), \text{Succ}(y)) \succeq \min_{x} (x, y) \\ & \text{div}^{\sharp} (\text{Succ}(x), \text{Succ}(y)) \succ \text{div}^{\sharp} (\min_{x} (x, y), \text{Succ}(y)) \end{split}$$

find a suitable reduction pair such that these constraints are satisfied

- many such reduction pairs are available (cf. term rewriting lecture)
  - Knuth-Bendix order (constraint solving is in P)
  - recursive path order (NP-complete)
  - polynomial interpretations (undecidable)
    - powerful
    - intuitive
    - automatable

• matrix interpretations (undecidable)

weighted path order (undecidable)

Example – Polynomial Interpretation

given constraints

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Polynomial Interpretation
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- interpret each *n*-ary symbol *F* as polynomial  $p_F(x_1, \ldots, x_n)$
- $\bullet\,$  variables in polynomials range over  $\mathbb N$  and polynomials have to be weakly monotone

$$x_i \ge y_i \longrightarrow p_F(x_1, \dots, x_i, \dots, x_n) \ge p_F(x_1, \dots, y_i, \dots, x_n)$$

sufficient criterion: forbid subtraction and negative numbers in  $p_F$ 

• interpretation is lifted to terms by composing polynomials

$$\|x\| = x$$
$$[F(t_1, \dots, t_n)] = p_F([t_1], \dots, [t_n])$$

- $(\succeq)$  is defined as
- $s \underset{(\sim)}{\succ} t \text{ iff } \forall \vec{x} \in \mathbb{N}^*. \llbracket s \rrbracket_{(\geq)} \llbracket t \rrbracket$
- $(\succ, \succeq)$  is a reduction pair, e.g.,

•  $SN(\succ)$  follows from strong-normalization of > on  $\mathbb N$ 

- $\succeq$  is closed under contexts since each  $p_F$  is weakly monotone
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- each polynomial constraint over  $\mathbb N$  can be brought into simple form " $p\geq 0$  " for some polynomial p

**Solving Polynomial Constraints** 

- replace  $p_1 > p_2$  by  $p_1 \ge p_2 + 1$
- replace  $p_1 \ge p_2$  by  $p_1 p_2 \ge 0$
- the question of " $p \ge 0$ " over  $\mathbb{N}$  is undecidable (Hilbert's 10th problem)
- approximation via absolute positiveness: if all coefficients of p are non-negative, then  $p\geq 0$  for all instances over  $\mathbb N$
- division example has trivial constraints

original	simplified
$x \ge x$	$0 \ge 0$
$1+x \ge x$	$1 \ge 0$
4 + x + 3y > 3 + x + 3y	$0 \ge 0$

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 $\begin{aligned} \min_{\substack{\mathsf{v} \in \mathsf{v}(x), \mathsf{Succ}(y) \\ \mathsf{div}^{\sharp}(\mathsf{Succ}(x), \mathsf{Succ}(y)) \\ \mathsf{div}^{\sharp}(\mathsf{minus}(x, y), \mathsf{Succ}(y))} \\ \mathsf{div}^{\sharp}(\mathsf{minus}(x, y), \mathsf{Succ}(y)) \\ \mathsf{div}^{\sharp}(\mathsf{minus}(x, y), \mathsf{Succ}(y)) \\ \mathsf{minus}(x_1, x_2) &= x_1 \\ p_{\mathsf{Tero}} &= 2 \end{aligned}$ 

 $\min(x, Zero) \succeq x$ 

$$p_{Succ}(x_1) = 1 + x_1$$
  
 $p_{div^{\sharp}}(x_1, x_2) = x_1 + 3x_2$ 

we obtain polynomial constraints

$$\llbracket \minus(x, \operatorname{Zero}) \rrbracket = x \ge x = \llbracket x \rrbracket$$
$$\llbracket \minus(\operatorname{Succ}(x), \operatorname{Succ}(y)) \rrbracket = 1 + x \ge x = \llbracket \minus(x, y) \rrbracket$$
$$\llbracket \operatorname{div}^{\sharp}(\operatorname{Succ} \dots) \rrbracket = 4 + x + 3y > 3 + x + 3y = \llbracket \operatorname{div}^{\sharp}(\minus \dots) \rrbracket$$
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**Symbolic Polynomial Interpretations** 

• fix shape of polynomial, e.g., linear

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$$p_F(x_1, \ldots, x_n) = F_0 + F_1 x_1 + \cdots + F_n x_n$$

where the  $F_i$  are symbolic coefficients

$$p_{minus}(x_1, x_2) = x_1$$

$$p_{Zero} = 2$$

$$p_{Succ}(x_1) = 1 + x_1$$

$$p_{div} \#(x_1, x_2) = x_1 + 3x_2$$

concrete interpretation above becomes symbolic

$$\begin{split} p_{\mathsf{minus}}(x_1, x_2) &= \mathsf{m}_0 + \mathsf{m}_1 x_1 + \mathsf{m}_2 x_2 \\ p_{\mathsf{Zero}} &= \mathsf{Z}_0 \\ p_{\mathsf{Succ}}(x_1) &= \mathsf{S}_0 + \mathsf{S}_1 x_1 \\ p_{\mathsf{div}^{\sharp}}(x_1, x_2) &= \mathsf{d}_0 + \mathsf{d}_1 x_1 + \mathsf{d}_2 x_2 \\ \text{Part 4 - Checking Well-Definedness of Functional Programs} \end{split}$$

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Symbolic Polynomial Constraints • given constraints

**Finding Polynomial Interpretations** 

aim: search for suitable interpretationapproach: perform everything symbolically

• in division example, interpretation was given on slides

$$\begin{aligned} \minus(x, \operatorname{Zero}) &\succeq x\\ \minus(\operatorname{Succ}(x), \operatorname{Succ}(y)) &\succeq \minus(x, y)\\ \operatorname{div}^{\sharp}(\operatorname{Succ}(x), \operatorname{Succ}(y)) &\succ \operatorname{div}^{\sharp}(\minus(x, y), \operatorname{Succ}(y)) \end{aligned}$$

Part 4 - Checking Well-Definedness of Functional Programs

• obtain symbolic polynomial constraints

$$\begin{split} \mathsf{m}_0 + \mathsf{m}_1 x + \mathsf{m}_2 \mathsf{Z}_0 &\geq x \\ \mathsf{m}_0 + \mathsf{m}_1 (\mathsf{S}_0 + \mathsf{S}_1 x) + \mathsf{m}_2 (\mathsf{S}_0 + \mathsf{S}_1 y) &\geq \mathsf{m}_0 + \mathsf{m}_1 x + \mathsf{m}_2 y \\ \mathsf{d}_0 + \mathsf{d}_1 (\mathsf{S}_0 + \mathsf{S}_1 x) + \mathsf{d}_2 (\mathsf{S}_0 + \mathsf{S}_1 y) &\geq \mathsf{d}_0 + \mathsf{d}_1 (\mathsf{m}_0 + \mathsf{m}_1 x + \mathsf{m}_2 y) \\ &+ \mathsf{d}_2 (\mathsf{S}_0 + \mathsf{S}_1 y) \end{split}$$

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• and simplify to

$$\begin{aligned} (\mathsf{m}_0 + \mathsf{m}_2 \mathsf{Z}_0) + (\mathsf{m}_1 - 1)x &\geq 0\\ (\mathsf{m}_1\mathsf{S}_0 + \mathsf{m}_2\mathsf{S}_0) + (\mathsf{m}_1\mathsf{S}_1 - \mathsf{m}_1)x + (\mathsf{m}_2\mathsf{S}_1 - \mathsf{m}_2)y &\geq 0\\ (\mathsf{d}_1\mathsf{S}_0 - \mathsf{d}_1\mathsf{m}_0 - 1) + (\mathsf{d}_1\mathsf{S}_1 - \mathsf{d}_1\mathsf{m}_1)x + (-\mathsf{d}_1\mathsf{m}_2)y &\geq 0\\ \text{Part 4 - Checking Well-Definedness of Functional Programs} \end{aligned}$$

Absolute Positiveness – Symbolic Example
 on symbolic polynomial constraints

$$(\mathsf{m}_0 + \mathsf{m}_2 \mathsf{Z}_0) + (\mathsf{m}_1 - 1)x \ge 0$$
  
$$(\mathsf{m}_1\mathsf{S}_0 + \mathsf{m}_2\mathsf{S}_0) + (\mathsf{m}_1\mathsf{S}_1 - \mathsf{m}_1)x + (\mathsf{m}_2\mathsf{S}_1 - \mathsf{m}_2)y \ge 0$$
  
$$(\mathsf{d}_1\mathsf{S}_0 - \mathsf{d}_1\mathsf{m}_0 - 1) + (\mathsf{d}_1\mathsf{S}_1 - \mathsf{d}_1\mathsf{m}_1)x + (-\mathsf{d}_1\mathsf{m}_2)y \ge 0$$

absolute positiveness works as before; obtain constraints

$m_0 + m_2 Z_0 \geq 0$	$m_1-1 \geq 0$	
$m_1S_0+m_2S_0\geq 0$	$m_1S_1-m_1\geq 0$	$m_2S_1-m_2\geq 0$
$d_1S_0-d_1m_0-1\geq 0$	$d_1S_1-d_1m_1\geq 0$	$-d_1m_2 \ge 0$

- at this point, use solver for integer arithmetic to find suitable coefficients (in  $\mathbb{N}$ )
- popular choice: SMT solver for integer arithmetic where one has to add constraints  $m_0 \ge 0, m_1 \ge 0, m_2 \ge 0, S_0 \ge 0, S_1 \ge 0, Z_0 \ge 0, \dots$

Constraint Solvi • original constr	ing by Hand – Ex aints	ample	Termination – Rec	duction Pairs	Constraint Solving <ul> <li>original constraints</li> </ul>	-	er – Example	Termination – Reduction Pairs	
	$m_0 + m_2 Z_0 \ge 0$	$m_1-1\geq 0$			<b>m</b> 0 -	$+ \mathbf{m}_2 \mathbf{Z}_0 \ge 0$	$m_1 - 1 \ge 0$		
m	$_1S_0 + m_2S_0 \ge 0$	$m_1S_1-m_1\geq 0$	$m_2S_1-m_2\geq 0$		$m_1S_0$	$+ \mathbf{m}_2 \mathbf{S}_0 \ge 0$	$m_1S_1-m_1\geq 0$	$m_2S_1-m_2\geq 0$	
$d_1S_0$	$-d_1m_0 - 1 \ge 0$	$d_1S_1-d_1m_1\geq 0$	$-d_1m_2 \ge 0$		$d_1S_0-d_1$	$\mathbf{m}_0 - 1 \ge 0$	$d_1S_1-d_1m_1\geq 0$	$-d_1m_2\geq 0$	
• delete trivial c	onstraints				<ul> <li>encode as SMT pr</li> </ul>	oblem in file div	vision.smt2		
		$m_1-1\geq 0$			(set-logic QF_N	IIA)			
		$m_1S_1 - m_1 \ge 0$	$m_2S_1-m_2\geq 0$		•		(declare-fun d2 () I	nt)	
$d_1S_0$	$-d_1m_0 - 1 \ge 0$	$d_1S_1-d_1m_1\geq 0$	$-d_1m_2\geq 0$		(assert (>= m0 0)) (assert (>= d2 0)) (assert (>= (+ m0 (* m2 Z0)) 0))				
<ul> <li>conclusions</li> </ul>					 (assert (>= (*	(- 1) d1 m2)	0))		
	$m_1 \geq 1$	$d_1 \geq 1$			(check-sat)	( 1) (1	0))		
	$S_0 \ge 1$	$S_1 \ge 1$			(get-model)				
	$m_2 = 0$	$S_1 \ge m_1$	$m_0 = 0$		(exit)				
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Constraint Solving by SMT-Solver - Example Continued

• invoke SMT solver, e.g., Microsoft's open source solver Z3

cmd> z3 division.smt2 sat (model (define-fun d1 () Int 8)

- (define-fun S1 () Int 15)
- (define-fun SO () Int 8)
- (define-fun ZO () Int O) (define-fun m2 () Int 0)
- (define-fun m1 () Int 12)
- (define-fun m0 () Int 4)
- (define-fun d2 () Int 0)
- (define-fun d0 () Int 0)
- )

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parse result to obtain polynomial interpretation

Constraint Solving by SMT-Solver - Scepticism

- polynomial interpretation found by SMT solving approach is generated by complex (potentially buggy) tool
- however, termination is essential for well-defined programs, i.e., in particular to derive correct theorems
- solution: certification
  - search for interpretation can be done in arbitrary untrusted way
  - write simple trusted checker that certifies whether concrete interpretation indeed satisfies all constraints
  - like solving NP-complete problem: positive answer can easily be verified
- in fact, this approach is heavily used in termination proving
  - untrusted tools: AProVE, T<sub>T</sub>T<sub>2</sub>, Terminator, ...
  - trusted checker: CeTA; soundness formally proven in Isabelle/HOL

Termination - Reduction Pairs

### Summary

- pattern-completeness and pattern-disjointness are decidable
- termination proving can be done via
  - dependency pairs
  - subterm criterion
  - size-change termination
  - polynomial interpretation
- termination proving often performed with help of SMT solvers
- increase reliability via certification: checking of generated proofs

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