



# Program Verification

## Part 4 – Checking Well-Definedness of Functional Programs

René Thiemann

Department of Computer Science

### Type-Checking with Implicit Variables

### Overview

- recall: a functional program is well-defined if
  - it is pattern disjoint,
  - it is pattern complete, and
  - $\leftrightarrow$  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
  - missing: algorithm for **type-inference**
  - missing: algorithm for **deciding pattern disjointness**
  - missing: algorithm for **deciding pattern completeness**
  - missing: methods to **ensure termination**
- all of these missing parts will be covered in this chapter

### Type-Inference

Type-Checking with Implicit Variables

- 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 × List → List
append(Cons(x, y), z) = Cons(x, append(y, z))
append(Nil, x) = x
```

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

## Type-Checking for Single Equations

- known: type-checking algorithm (we omit the error message in `assert` and `failure`)
 

```
typeCheck :: Sig -> Vars -> Term -> CheckS Type
  type Sig = FSym -> CheckS ([Type], Type) -  $\Sigma$ 
  type Vars = Var -> CheckS Type -  $\mathcal{V}$ 
  typeCheck takes  $\Sigma$  and  $\mathcal{V}$  and delivers type of term
```
- we want a function that works in the other direction: it gets an intended **type as input**, and delivers a suitable type for the variables
 

```
inferType :: Sig -> Type -> Term -> CheckS [(Var,Type)]
```
- then type-checking an equation without explicit `Vars` is possible
 

```
typeCheckEqn :: Sig -> (Term, Term) -> CheckS ()
typeCheckEqn sigma (Var x, r) = failure
typeCheckEqn sigma (l @ (Fun f _), r) = do
  (_,ty) <- sigma f
  vars <- inferType sigma ty l
  tyR <- typeCheck sigma (\ x -> lookup x vars) r
  assert (ty == tyR)
```

## Type-Inference Algorithm

- note: upcoming algorithm only infers types of variables  
(in polymorphic setting often also type of function symbols is inferred)
- ```
inferType :: Sig -> Type -> Term -> CheckS [(Var,Type)]
inferType sigma ty (Var x) = return [(x,ty)]
inferType sigma ty (Fun f ts) = do
  (tysIn,tyOut) <- sigma f
  assert (length tysIn == length ts)
  assert (tyOut == ty)
  varsL <- mapM (\ (tyi, ti) -> inferType sigma tyi ti) (zip tysIn ts)
  let vars = nub (concat varsL) -- nub removes duplicates
      assert (distinct (map fst vars))
  return vars

distinct :: Eq a => [a] -> Bool
distinct xs = length (nub xs) == length xs
```

## Soundness of Type-Inference Algorithm

- properties
  - if  $t \in \mathcal{T}(\Sigma, \mathcal{V})_\tau$  then  $\text{inferType } \Sigma \tau t = \text{return } (\mathcal{V} \cap \text{Vars}(t))$
  - if  $\text{inferType } \Sigma \tau t = \text{return } \mathcal{V}$  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 2/37–44
- note that ‘if  $t \in \mathcal{T}(\Sigma, \mathcal{V})_\tau$  then  $\text{inferType } \Sigma \tau t \neq \text{failure}$ ’ is a property which is not strong enough when performing induction

## 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)
```

## 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
```

```
processDataDefinition ::
  ProgInfo -> DataDefinition -> CheckS ProgInfo
processDataDefinition = ... -- previous slide
```

```
processDataDefinitions ::
  ProgInfo -> [DataDefinition] -> CheckS ProgInfo
processDataDefinitions = foldM processDataDefinition
```

## Checking Function Definitions w.r.t. Slide 3/15

```
data FunctionDefinition = Function
  FSym          -- name of function
  FSymInfo      -- type of function
  [(Term,Term)] -- equations

processFunctionDefinition
  :: ProgInfo -> FunctionDefinition -> CheckS ProgInfo
processFunctionDefinition = ... -- exercise

processFunctionDefinitions ::
  ProgInfo -> [FunctionDefinition] -> CheckS ProgInfo
processFunctionDefinitions =
  foldM processFunctionDefinition
```

## Checking Functional Programs

```
initialProgInfo = ProgInfo [] [] [] []

processProgram :: FunctionalProg -> CheckS ProgInfo
processProgram (dataDefs, funDefs) = do
  pi <- processDataDefinitions initialProgInfo dataDefs
  processFunctionDefinitions pi funDefs
```

## Current State

- `processProgram :: FunctionalProg -> CheckS 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 us 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

## Deciding Pattern Disjointness

- program is pattern disjoint if for all  $f : \tau_1 \times \dots \times \tau_n \rightarrow \tau \in \mathcal{D}$  and all  $t_1 \in \mathcal{T}(\mathcal{C})_{\tau_1}, \dots, t_n \in \mathcal{T}(\mathcal{C})_{\tau_n}$  there is at most one equation  $\ell = r$  in the program, such that  $\ell$  matches  $f(t_1, \dots, t_n)$
- in proseminar: 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  $\exists \sigma. s\sigma = t\sigma$

## 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  $\forall i \in \{1, \dots, n\}. s_i\sigma = t_i\sigma$
- two different kinds of output:
  - unification problem in **resolved form**:  
 $\{x_1 \stackrel{?}{=} v_1, \dots, x_m \stackrel{?}{=} v_m\}$  with distinct  $x_j$ 's and  $x_j \notin \mathcal{V}(v_i)$   
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
- algorithm itself is build via one-step relation  $\rightsquigarrow$  which is applied as long as possible

## Unification Algorithm of Martelli and Montanari, continued

- 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

$$U \cup \{t \stackrel{?}{=} t\} \rightsquigarrow U \quad (\text{delete})$$

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

$$U \cup \{F(u_1, \dots, u_k) \stackrel{?}{=} G(v_1, \dots, v_\ell)\} \rightsquigarrow \perp, \text{ if } F \neq G \vee k \neq \ell \quad (\text{clash})$$

$$U \cup \{F(\dots) \stackrel{?}{=} x\} \rightsquigarrow U \cup \{x \stackrel{?}{=} F(\dots)\} \quad (\text{swap})$$

$$U \cup \{x \stackrel{?}{=} F(\dots)\} \rightsquigarrow \perp, \text{ if } x \in \text{Vars}(F(\dots)) \quad (\text{occurs check})$$

$$U \cup \{x \stackrel{?}{=} t\} \rightsquigarrow U\{x/t\} \cup \{x \stackrel{?}{=} t\}, \quad (\text{eliminate})$$

if  $x \notin \text{Vars}(t)$  and  $x$  occurs in  $U$

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

## Correctness of Unification Algorithm

- we only state properties (proofs: see term rewriting lecture)
  - $\rightsquigarrow$  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  $V = \perp$  then  $U$  is unsolvable
    - otherwise,  $V$  represents a substitution that is a solution to  $U$
- note that  $\rightsquigarrow$  is not confluent
  - $\{x \stackrel{?}{=} y, y \stackrel{?}{=} x\} \rightsquigarrow^{x/y} \{x \stackrel{?}{=} y, y \stackrel{?}{=} y\} \rightsquigarrow \{x \stackrel{?}{=} y\}$
  - $\{x \stackrel{?}{=} y, y \stackrel{?}{=} x\} \rightsquigarrow^{y/x} \{x \stackrel{?}{=} x, y \stackrel{?}{=} x\} \rightsquigarrow \{y \stackrel{?}{=} x\}$

## Correctness of an Implementation of a (Unification) Algorithm

- any concrete implementation will make choices
  - preference of rules
  - selection of pairs from  $U$
  - representation of sets  $U$
  - (pivot-selection in quicksort)
  - (order of edges in graph-/tree-traversals)
  - ...
- task: how to ensure that implementation is sound
- solution: **refinement** proof
  - aim: reuse correctness of abstract algorithm ( $\rightsquigarrow$ )
  - define relation between representations in concrete and abstract algorithm (this was called **alignment** before and done informally)
  - show that **concrete algorithm has less behaviour**, i.e., every result of concrete (deterministic) algorithm can be related to some result of (non-deterministic) abstract algorithm
  - benefit: clear **separation** between
    - soundness of abstract algorithm (solves unification problems)
    - soundness of implementation (implements abstract algorithm)

## A Concrete Implementing of the Unification Algorithm

```
subst :: Var -> Term -> Term -> Term
subst x t = applySubst (\ y -> if y == x then t else Var y)

unify :: [(Term, Term)] -> CheckS [(Var, Term)]
unify u = unifyMain u []

unifyMain :: [(Term, Term)] -> [(Var,Term)] -> CheckS [(Var, Term)]
unifyMain [] v = return v -- return solved form
unifyMain ((Fun f ts, Fun g ss) : u) v = do
  assert (f == g && length ts == length ss) -- clash
  unifyMain (zip ts ss ++ u) v -- decompose
unifyMain ((Fun f ts, x) : u) v =
  unifyMain ((x, Fun f ts) : u) v -- swap
unifyMain ((Var x, t) : u) v =
  if Var x == t then unifyMain u v -- delete
  else do
    assert (not (elem x (varsTerm t))) -- occurs check
    unifyMain
      (map (\ (l,r) -> (subst x t l, subst x t r)) u)
      ((x,t) : map (\ (y, s) -> (y, subst x t s)) v)
```

## Notes on Implementation

- it is non-trivial to prove soundness of implementation, since there are several differences w.r.t.  $\rightsquigarrow$ 
  - unifyMain* 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.  $\rightsquigarrow$
  - 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  $v$** 
    - 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

## Soundness via Refinement: Setting up the Relation

- 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
  - $set\ (x : xs) = \{x\} \cup set\ xs$
  - $set\ (xs ++ ys) = set\ xs \cup set\ ys$
  - $set\ (zip\ [x_1, \dots, x_n]\ [y_1, \dots, y_n]) = \{(x_1, y_1), \dots, (x_n, y_n)\}$
  - $set\ (map\ f\ [x_1, \dots, x_n]) = \{f\ x_1, \dots, f\ x_n\}$
  - $subst\ x\ t\ s = s\{x/t\}$
  - ...

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

## Soundness via Refinement: Main Statement

- we need a function to align results in the algorithm to unification problems: check-to-unification problem (*ctu*)
 
$$ctu\ (return\ w) = set\ w \quad \text{and} \quad ctu\ (failure\ e) = \perp$$
- property**: whenever  $(u, v) \sim U$  and  $unifyMain\ u\ v = res$  then  $U \rightsquigarrow^! ctu\ res$
- 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,  $unifyMain\ u\ [] = res$
- moreover,  $(u, []) \sim set\ u$  by definition of  $\sim$
- via property conclude  $set\ u \rightsquigarrow^! ctu\ res$
- at this point apply correctness of  $\rightsquigarrow$ :  
 $ctu\ res$  is the correct answer to the unification problem  $set\ u$

## Proving the Refinement Property

- property  $P(u, v, U)$ :  $(u, v) \sim U \wedge unifyMain\ u\ v = res \longrightarrow U \rightsquigarrow^! ctu\ res$
- $(u, v) \sim U \iff U = set\ u \cup set\ v \wedge set\ v \in NF(\rightsquigarrow) \wedge \forall (x, t) \in set\ v. x \notin Vars(u)$
- 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
    - (\*\*)  $unifyMain\ []\ v = res$
  - from (\*) conclude  $U = set\ v$  and  $set\ v \in NF(\rightsquigarrow)$
  - from (\*\*) conclude  $res = return\ v$  and hence,  $ctu\ res = set\ v$
  - we have to show  $U \rightsquigarrow^! ctu\ res$ , i.e.,  $set\ v \rightsquigarrow^! set\ v$  which is satisfied since  $set\ v \in NF(\rightsquigarrow)$

- $P(u, v, U): (u, v) \sim U \wedge \text{unifyMain } u \ v = \text{res} \longrightarrow U \rightsquigarrow^1 \text{ctu res}$
- $(u, v) \sim U \iff U = \text{set } u \cup \text{set } v \wedge \text{set } v \in \text{NF}(\rightsquigarrow) \wedge \forall (x, t) \in \text{set } v. x \notin \text{Vars}(u)$

case 2 (arguments  $(F(ts), G(ss)) : u$  and  $v$ )

- we have to prove  $P((F(ts), G(ss)) : u, v, U)$ , so assume
  - (\*)  $((F(ts), G(ss)) : u, v) \sim U$  and
  - (\*\*)  $\text{unifyMain } ((F(ts), G(ss)) : u) \ v = \text{res}$
- consider sub-cases
  - $\neg(F = G \wedge \text{length } ts = \text{length } ss)$ :
    - from (\*\*) conclude  $\text{ctu res} = \perp$
    - from (\*) conclude  $F(ts) \stackrel{?}{=} G(ss) \in U$  and hence  $U \rightsquigarrow \perp$  by (clash)
    - consequently,  $U \rightsquigarrow^1 \text{ctu res}$
  - $F = G \wedge \text{length } ts = \text{length } ss$ :
    - from (\*\*) conclude  $\text{res} = \text{unifyMain } ((F(ts), G(ss)) : u) \ v = \text{unifyMain } (\text{zip } ts \ ss \ ++ \ u) \ v$
    - from (\*) and alignment for  $\text{zip}$  and  $\text{++}$  conclude  $U = \{F(ts) \stackrel{?}{=} G(ss)\} \cup \text{set } u \cup \text{set } v$  and hence  $U \rightsquigarrow \text{set } (\text{zip } ts \ ss \ ++ \ u) \cup \text{set } v =: V$  by (decompose)
    - we get  $P(\text{zip } ts \ ss \ ++ \ u, v, V)$  as IH;  $(\text{zip } ts \ ss \ ++ \ u, v) \sim V$  follows from (\*), so  $U \rightsquigarrow V \rightsquigarrow^1 \text{ctu res}$

- $P(u, v, U): (u, v) \sim U \wedge \text{unifyMain } u \ v = \text{res} \longrightarrow U \rightsquigarrow^1 \text{ctu res}$
- $(u, v) \sim U \iff U = \text{set } u \cup \text{set } v \wedge \text{set } v \in \text{NF}(\rightsquigarrow) \wedge \forall (x, t) \in \text{set } v. x \notin \text{Vars}(u)$

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
  - (\*\*)  $\text{unifyMain } ((x, t) : u) \ v = \text{res}$
- consider sub-cases (where the red part is not triggered by structure of algorithm)
  - $x \neq t \wedge x \notin \text{Vars}(t) \wedge x$  occurs in  $\text{set } u \cup \text{set } v$ :
    - define  $u' = \text{map } (\lambda(l, r). (\text{subst } x \ t \ l, \text{subst } x \ t \ r)) \ u$
    - define  $v' = \text{map } (\lambda(y, s). (y, \text{subst } x \ t \ s)) \ v$
    - define  $V = (\text{set } u \cup \text{set } v)\{x/t\} \cup \{x \stackrel{?}{=} t\}$
    - from (\*\*) conclude  $\text{res} = \text{unifyMain } ((x, t) : u) \ v = \text{unifyMain } u' \ ((x, t) : v')$
    - from IH conclude  $P(u', (x, t) : v', V)$  and hence,  $(u', (x, t) : v') \sim V \longrightarrow V \rightsquigarrow^1 \text{ctu res}$
    - for proving  $U \rightsquigarrow^1 \text{ctu res}$  it hence suffices to show  $(u', (x, t) : v') \sim V$  and  $U \rightsquigarrow V$
    - $U \stackrel{(*)}{=} \{x \stackrel{?}{=} t\} \cup \text{set } u \cup \text{set } v \rightsquigarrow (\text{set } u \cup \text{set } v)\{x/t\} \cup \{x \stackrel{?}{=} t\} = V$  by (eliminate) because of preconditions

- $(u, v) \sim U \iff U = \text{set } u \cup \text{set } v \wedge \text{set } v \in \text{NF}(\rightsquigarrow) \wedge \forall (x, t) \in \text{set } v. x \notin \text{Vars}(u)$

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  $x \neq t \wedge x \notin \text{Vars}(t) \wedge x$  occurs in  $\text{set } u \cup \text{set } v$ :
  - define  $u' = \text{map } (\lambda(l, r). (\text{subst } x \ t \ l, \text{subst } x \ t \ r)) \ u$
  - define  $v' = \text{map } (\lambda(y, s). (y, \text{subst } x \ t \ s)) \ v$
  - define  $V = (\text{set } u \cup \text{set } v)\{x/t\} \cup \{x \stackrel{?}{=} t\}$
  - we still need to show  $(u', (x, t) : v') \sim V$
  - since (\*) holds, we know  $\forall (y, s) \in \text{set } v. x \neq y$
  - hence,  $v' = \text{map } (\lambda(y, s). (\text{subst } x \ t \ y, \text{subst } x \ t \ s)) \ v$
  - so,  $V = (\text{set } u)\{x/t\} \cup \{x \stackrel{?}{=} t\} \cup (\text{set } v)\{x/t\} = \text{set } u' \cup \text{set } ((x, t) : v')$
  - we show  $\forall (y, s) \in \text{set } ((x, t) : v'). y \notin \text{Vars}(u')$  as follows:  $x \notin \text{Vars}(u')$  since  $x \notin \text{Vars}(t)$ ; and if  $(y, s) \in \text{set } v'$ , then  $(y, s') \in \text{set } v$  for some  $s'$  and by (\*) we conclude  $y \notin \text{Vars}((x, t) : u)$ ; thus,  $y \notin \text{Vars}((\text{set } u)\{x/t\}) = \text{Vars}(u')$
  - we finally show  $\text{set } ((x, t) : v') \in \text{NF}(\rightsquigarrow)$ : so, assume to the contrary that some step is applicable; by the shape of  $\text{set } ((x, t) : v')$  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

## 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  $\text{set } u \cup \text{set } v$ ”
    - most cases can easily be solved, after having identified **suitable invariants**
    - fully reuse correctness of  $\rightsquigarrow$
  - we only proved partial correctness
  - termination of implementation: consider lexicographic measure

$$\underbrace{(|\text{Vars}(\text{set } u)|)}_{(\text{eliminate})}, \quad \underbrace{|u|}_{(\text{decomp}), (\text{delete})}, \quad \underbrace{\text{length } [x \mid (t, \text{Var } x) \leftarrow u]}_{(\text{swap})}$$

## Checking Pattern Completeness

### Pattern Problems

- **reminder**: program is pattern complete, if for all  $f : \tau_1 \times \dots \times \tau_n \rightarrow \tau \in \mathcal{D}$  and all  $t_i \in \mathcal{T}(\mathcal{C})_{\tau_i}$  there is some lhs that matches  $f(t_1, \dots, t_n)$
- **algorithm** considers more generic shape
  - **matching problems**  $mp$  consist of pairs of terms  $(\ell, t)$  where
    - $\ell$  is (a subterm of) some lhs
    - $t$  is a term, representing the set of all its constructor ground instances, e.g.,  $t = f(x_1, \dots, x_n)$
    - semantics: find one substitution  $\sigma$  such that  $\ell\sigma = t\gamma$  for all  $(\ell, t) \in mp$  where  $\gamma$  is some constructor ground substitution
  - **pattern problems**  $pp$  consist of multiple matching problems
    - semantics: disjunction, i.e., find one suitable matching problem
    - reason for disjunction: a term  $t$  might be matched by arbitrary lhs
    - initially:  $pp = \{ \{(\ell_1, t)\}, \dots, \{(\ell_n, t)\} \}$  for lhs  $\ell_1, \dots, \ell_n$
  - **sets of pattern problems**  $P$  consist of several pattern problems
    - semantics: conjunction
    - reason: consider different ground instances and different defined function symbols
    - **initial set of pattern problems**:  $P_{init} = \{ \{(\ell, f(x_1, \dots, x_n))\} \mid \ell \text{ is lhs of } f\text{-eqn.} \} \mid f \in \mathcal{D} \}$
- overall semantics:  $P$  is **complete** iff

$$\forall pp \in P. \forall \gamma : \mathcal{V} \rightarrow \mathcal{T}(\mathcal{C}). \exists mp \in pp. \exists \sigma. \forall (\ell, t) \in mp. \ell\sigma = t\gamma$$

### Reformulation of Pattern Completeness of Programs

- definitions of previous slide (omitting types)
  - program is pattern complete iff for all  $f \in \mathcal{D}$  and all  $t_i \in \mathcal{T}(\mathcal{C})$  there is some lhs that matches  $f(t_1, \dots, t_n)$
  - $P_{init} = \{ \{(\ell, f(x_1, \dots, x_n))\} \mid \ell \text{ is lhs of } f\text{-equation} \} \mid f \in \mathcal{D} \}$
  - $P$  is complete iff  $\forall pp \in P. \forall \gamma : \mathcal{V} \rightarrow \mathcal{T}(\mathcal{C}). \exists mp \in pp. \exists \sigma. \forall (\ell, t) \in mp. \ell\sigma = t\gamma$
- **corollary**: program is pattern complete iff  $P_{init}$  is complete

### Task: determine completeness of pattern problems

- algorithm modifies matching problems and (sets of) pattern problems
- special problems:  $\perp$  represents a non-solvable matching problem and an incomplete set of pattern problems, and  $\top$  represents a complete pattern problem
- here: only consider **linear** pattern problems, i.e., problems where variables in lhs of programs occur at most once

### Transforming Matching and Pattern Problems

$$\begin{aligned} \{(F(\ell_1, \dots, \ell_n), F(t_1, \dots, t_n))\} \uplus mp &\rightarrow \{(\ell_1, t_1), \dots, (\ell_n, t_n)\} \cup mp && \text{(decompose)} \\ \{(x, t)\} \uplus mp &\rightarrow mp && \text{(match)} \\ \{(F(\dots), G(\dots))\} \uplus mp &\rightarrow \perp && \text{if } F \neq G \text{ (clash)} \\ \{mp\} \uplus pp &\rightarrow \{mp'\} \cup pp && \text{if } mp \rightarrow mp' \text{ (simp-mp)} \\ \{\perp\} \uplus pp &\rightarrow pp && \text{(remove-mp)} \\ \{\emptyset\} \uplus pp &\rightarrow \top && \text{(success)} \\ \{pp\} \uplus P &\rightarrow \{pp'\} \cup P && \text{if } pp \rightarrow pp' \text{ (simp-pp)} \\ \{\emptyset\} \uplus P &\rightarrow \perp && \text{(failure)} \\ \{\top\} \uplus P &\rightarrow P && \text{(remove-pp)} \\ \{pp\} \uplus P &\rightarrow Inst(pp, x) \cup P && \text{if } mp \in pp \text{ and } (F(\dots), x) \in mp \text{ (instantiate)} \end{aligned}$$

where  $Inst(pp, x)$  contains a pattern problem  $pp\sigma_{x,c}$  for each constructor  $c$  where

- $x : \tau$  and  $c : \tau_1 \times \dots \times \tau_n \rightarrow \tau$  and  $x_1 : \tau_1, \dots, x_n : \tau_n$  are fresh, and
- $pp\sigma_{x,c}$  is obtained from  $pp$  by replacing each pair  $(\ell, t)$  by  $(\ell, t\{x/c(x_1, \dots, x_n)\})$

## Example

consider

```

data Bool = True : Bool | False : Bool
ℓ1 := conj(True, True) = ...
ℓ2 := conj(False, y) = ...
ℓ3 := conj(x, False) = ...

```

then we have

$$\begin{aligned}
P_{init} &= \{ \{ \{ (\ell_1, \text{conj}(x_1, x_2)) \}, \{ (\ell_2, \text{conj}(x_1, x_2)) \}, \{ (\ell_3, \text{conj}(x_1, x_2)) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_1), (\text{True}, x_2) \}, \{ (\text{False}, x_1), (y, x_2) \}, \{ (x, x_1), (\text{False}, x_2) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_1), (\text{True}, x_2) \}, \{ (\text{False}, x_1) \}, \{ (\text{False}, x_2) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, \text{True}), (\text{True}, x_2) \}, \{ (\text{False}, \text{True}) \}, \{ (\text{False}, x_2) \} \}, \\
&\quad \{ (\text{True}, \text{False}), (\text{True}, x_2) \}, \{ (\text{False}, \text{False}) \}, \{ (\text{False}, x_2) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_2) \}, \perp, \{ (\text{False}, x_2) \} \}, \{ \perp, \emptyset, \{ (\text{False}, x_2) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_2) \}, \{ (\text{False}, x_2) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, \text{True}) \}, \{ (\text{False}, \text{True}) \} \}, \{ \{ (\text{True}, \text{False}) \}, \{ (\text{False}, \text{False}) \} \} \} \multimap^* \emptyset
\end{aligned}$$

## Example

consider

```

data Bool = True : Bool | False : Bool
ℓ1 := conj(True, True) = ...
ℓ2 := conj(False, y) = ...

```

then we have

$$\begin{aligned}
P_{init} &= \{ \{ \{ (\ell_1, \text{conj}(x_1, x_2)) \}, \{ (\ell_2, \text{conj}(x_1, x_2)) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_1), (\text{True}, x_2) \}, \{ (\text{False}, x_1) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, \text{True}), (\text{True}, x_2) \}, \{ (\text{False}, \text{True}) \} \}, \\
&\quad \{ (\text{True}, \text{False}), (\text{True}, x_2) \}, \{ (\text{False}, \text{False}) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_2) \}, \perp \}, \{ \emptyset, \perp \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, x_2) \} \} \} \\
&\multimap^* \{ \{ \{ (\text{True}, \text{True}) \} \}, \{ \{ (\text{True}, \text{False}) \} \} \} \multimap^* \{ \top, \emptyset \} \multimap^* \perp
\end{aligned}$$
Partial Correctness of  $\multimap$ 

- **theorem:** whenever  $P \multimap Q$ , then  $P$  is complete iff  $Q$  is complete
- **corollary:** if  $P \multimap^* \emptyset$  then  $P$  is complete, and if  $P \multimap^* \perp$  then  $P$  is not complete

- **definition:**  $P$  is complete iff

$$\forall pp \in P. \forall \gamma : \mathcal{V} \rightarrow \mathcal{T}(\mathcal{C}). \underbrace{\exists mp \in pp. \exists \sigma. \forall (\ell, t) \in mp. \ell \sigma = t \gamma}_{=: \psi}$$

- **proof of theorem** by case analysis on the various rules

- (clash): first inline rule to  $\{ \{ \{ (F(\dots), G(\dots)) \} \} \} \uplus mp \} \uplus pp \} \uplus P \multimap \{ pp \} \cup P$ , if  $F \neq G$ 
  - by definition of completeness and structure of rule it suffices to show that completeness is preserved by rule

$$\underbrace{\{ \{ \{ (F(\dots), G(\dots)) \} \} \} \uplus mp \} \uplus pp}_{=: mp'}$$

- hence, it suffices to show that  $\psi$  is not satisfied when choosing  $mp'$  in the existential quantifier  $\exists mp \in pp. \dots$
- but this property is easy to see, since  $\ell \sigma = t \gamma$  is never satisfied if  $(\ell, t)$  is  $(F(\dots), G(\dots))$
- many other rules are similar, exceptions are (match) and (instantiate)

Partial Correctness of  $\multimap$ , continued

- **definition:**  $P$  is complete iff

$$\forall pp \in P. \forall \gamma : \mathcal{V} \rightarrow \mathcal{T}(\mathcal{C}). \exists mp \in pp. \exists \sigma. \forall (\ell, t) \in mp. \ell \sigma = t \gamma$$

- **proof continued**

- (instantiate):  $\{ pp \} \uplus P \multimap \text{Inst}(pp, x) \cup P$ , where  $x : \tau$ ,  $\tau$  has constructors  $c_1, \dots, c_n$ , and  $\sigma_i = \{ x / c_i(x_1, \dots, x_k) \}$  for fresh  $x_i$ , and  $\text{Inst}(pp, x) = \{ pp \sigma_i \mid 1 \leq i \leq n \}$ 
  - we only consider one direction of the proof: we assume that  $\text{Inst}(pp, x)$  is complete and prove that  $pp$  is complete
  - to this end, consider an arbitrary constructor ground substitution  $\gamma$
  - since  $\gamma$  is constructor ground, we know  $\gamma(x) = c_i(t_1, \dots, t_k)$  for some constructor  $c_i$  and constructor ground terms  $t_1, \dots, t_k$
  - define  $\gamma'(y) = \begin{cases} t_j, & \text{if } y = x_j \\ \gamma(y), & \text{otherwise} \end{cases}$
  - $\gamma'$  is well-defined since the  $x_j$ 's are distinct, and  $\gamma'$  is a constructor ground substitution
  - note that  $t \gamma = t \gamma_i \gamma'$  for all terms  $t$  that occur in  $pp$  since the  $x_j$ 's are fresh
  - by completeness of  $\text{Inst}(pp, x)$  there must be some  $mp \in pp \sigma_i$  and  $\sigma$  such that  $\forall (\ell, t) \in mp. \ell \sigma = t \gamma'$
  - hence, there is some  $mp \in pp$  and  $\sigma$  such that  $\forall (\ell, t) \in mp. \ell \sigma = t \gamma_i \gamma'$
  - together with  $t \gamma = t \gamma_i \gamma'$  we conclude that  $pp$  is complete

## Correctness of $\multimap$ , Missing Parts

- already proven
  - if  $P \multimap^* \emptyset$  then  $P$  is complete
  - if  $P \multimap^* \perp$  then  $P$  is not complete
- open: termination of  $\multimap$
- open: can  $\multimap$  get stuck?

## Termination of $\multimap$

$$\begin{aligned} \{pp\} \uplus P \multimap \{pp'\} \cup P & \quad \text{if } pp \multimap pp' && \text{(simp-pp)} \\ \{\emptyset\} \uplus P \multimap \perp & && \text{(failure)} \\ \{\top\} \uplus P \multimap P & && \text{(remove-pp)} \\ \{pp\} \uplus P \multimap \text{Inst}(pp, x) \cup P & \quad \text{if } mp \in pp \text{ and } (F(\dots), x) \in mp && \text{(instantiate)} \end{aligned}$$

- define  $|\ell - t|$  as a measure of difference of  $\ell$  and  $t$ 
  - $|\ell - x|$  = number of function symbols in  $\ell$
  - $|F(\ell_1, \dots, \ell_n) - F(t_1, \dots, t_n)| = \sum_i |\ell_i - t_i|$
  - $|\ell - t| = 0$ , in all other cases
- map each pattern problem  $pp$  to number  $|pp| = \sum_{mp \in pp, (\ell, t) \in mp} |\ell - t|$
- map each set of pattern problem  $P$  to multiset  $\{|pp| \mid pp \in P\}$
- this multiset decreases in (instantiate) and is not increased in the other  $\multimap$ -rules (multiset decrease:  $M \cup N \succ^{mul} M \cup N'$  if  $N \neq \emptyset$  and  $\forall y \in N'. \exists x \in N. x > y$ )
- hence (instantiate) cannot be applied infinitely often
- since the remaining rules also terminate,  $\multimap$  must terminate

## $\multimap$ Cannot Get Stuck

$$\begin{aligned} \{(F(\ell_1, \dots, \ell_n), F(t_1, \dots, t_n))\} \uplus mp \multimap \{(\ell_1, t_1), \dots, (\ell_n, t_n)\} \cup mp & \quad \text{(decompose)} \\ \{(x, t)\} \uplus mp \multimap mp & \quad \text{(match)} \\ \{(F(\dots), G(\dots))\} \uplus mp \multimap \perp & \quad \text{if } F \neq G && \text{(clash)} \\ \{mp\} \uplus pp \multimap \{mp'\} \cup pp & \quad \text{if } mp \multimap mp' && \text{(simp-mp)} \\ \{\perp\} \uplus pp \multimap pp & && \text{(remove-mp)} \\ \{\emptyset\} \uplus pp \multimap \top & && \text{(success)} \\ \{pp\} \uplus P \multimap \{pp'\} \cup P & \quad \text{if } pp \multimap pp' && \text{(simp-pp)} \\ \{\emptyset\} \uplus P \multimap \perp & && \text{(failure)} \\ \{\top\} \uplus P \multimap P & && \text{(remove-pp)} \\ \{pp\} \uplus P \multimap \text{Inst}(pp, x) \cup P & \quad \text{if } mp \in pp \text{ and } (F(\dots), x) \in mp && \text{(instantiate)} \end{aligned}$$

- **lemma:** whenever  $P$  is well-typed and in normal form w.r.t.  $\multimap$ , then  $P \in \{\emptyset, \perp\}$
- **proof:** by a large case-analysis

## Implementation and Complexity of $\multimap$

- clearly,  $\multimap$  is formulated abstractly
- a concrete implementation has to use a concrete representation for matching- and pattern problems; it has to resolve non-determinism, e.g., order of rules, selection of instantiation variables, etc.
- theorem: deciding pattern completeness is co-NP-hard
- consequence: worst-case complexity on required number of  $\multimap$ -steps unlikely to be sub-exponential
- fully verified implementation exists
- currently fastest known algorithm for pattern completeness, developed for this lecture

## Summary on Pattern Completeness

- pattern completeness of functional programs is decidable:
  - program is pattern complete iff  $P_{init} \not\rightarrow^! \emptyset$
- two possible extensions
  - generation of counter-examples
  - handling of non-linear pattern problems
- partial correctness was proven via invariant of  $\rightarrow^*$
- 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

## Termination – Dependency Pairs

## Termination of Programs

- 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

- 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  $f(t_1, \dots, t_n)$
- example without confluence

$$\begin{aligned}
 f(\text{True}, \text{False}, x) &= f(x, x, x) \\
 f(\dots, \dots, x) &= x \quad (\text{all other cases}) \\
 \text{coin} &= \text{True} \\
 \text{coin} &= \text{False}
 \end{aligned}$$

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

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

## Subterm Relation and Innermost Evaluation

- define  $\triangleright$  as the strict **subterm relation** and  $\trianglerighteq$  as its reflexive closure

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

- innermost evaluation**  $\overset{\circ}{\hookrightarrow}$  is defined similar to one-step evaluation  $\hookrightarrow$

$$\frac{\frac{s_i \overset{\circ}{\hookrightarrow} t_i}{F(s_1, \dots, s_i, \dots, s_n) \overset{\circ}{\hookrightarrow} F(s_1, \dots, t_i, \dots, s_n)} \text{ rewrite in context}}{\ell = r \text{ is equation in program } \forall s \triangleleft \ell \sigma. s \in NF(\overset{\circ}{\hookrightarrow})} \text{ root step} \frac{}{\ell \sigma \overset{\circ}{\hookrightarrow} r \sigma}$$

- example

$$f(\text{True}, \text{False}, \text{coin}) \not\overset{\circ}{\hookrightarrow} f(\text{coin}, \text{coin}, \text{coin})$$

since  $\text{coin} \triangleleft f(\text{True}, \text{False}, \text{coin})$  and  $\text{coin} \notin NF(\overset{\circ}{\hookrightarrow})$

## 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  $SN(\overset{\circ}{\hookrightarrow})$
- only reason for potential non-termination: recursive calls
- for each recursive call of equation  $f(t_1, \dots, t_n) = \ell = r \trianglerighteq f(s_1, \dots, s_n)$  build one **dependency pair** with fresh (constructor) symbol  $f^\sharp$ :

$$f^\sharp(t_1, \dots, t_n) \rightarrow f^\sharp(s_1, \dots, s_n)$$

define  $DP$  as the set of all dependency pairs

- example program for Ackermann function has three dependency pairs

$$\text{ack}(\text{Zero}, y) = \text{Succ}(y)$$

$$\text{ack}(\text{Succ}(x), \text{Zero}) = \text{ack}(x, \text{Succ}(\text{Zero}))$$

$$\text{ack}(\text{Succ}(x), \text{Succ}(y)) = \text{ack}(x, \text{ack}(\text{Succ}(x), y))$$

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

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

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

## Termination Analysis with Dependency Pairs, continued

- dependency pairs provide characterization of termination
- definition: let  $P \subseteq DP$ ; a **P-chain** is a possible infinite sequence

$$s_1 \sigma_1 \rightarrow t_1 \sigma_1 \overset{\circ}{\hookrightarrow}^* s_2 \sigma_2 \rightarrow t_2 \sigma_2 \overset{\circ}{\hookrightarrow}^* s_3 \sigma_3 \rightarrow t_3 \sigma_3 \overset{\circ}{\hookrightarrow}^* \dots$$

such that all  $s_i \rightarrow t_i \in P$  and all  $s_i \sigma_i \in NF(\overset{\circ}{\hookrightarrow})$

- $s_i \sigma_i \rightarrow t_i \sigma_i$  represent the “main” recursive calls that may lead to non-termination
- $t_i \sigma_i \overset{\circ}{\hookrightarrow}^* s_{i+1} \sigma_{i+1}$  corresponds to evaluation of arguments of recursive calls
- theorem:**  $SN(\overset{\circ}{\hookrightarrow})$  iff there is no infinite  $DP$ -chain
- advantage of dependency pairs
  - in infinite chain, non-terminating recursive calls are always applied at the root
  - simplifies termination analysis

## Example of Evaluation and Chain

$$\begin{aligned} \text{minus}(x, \text{Zero}) &= x \\ \text{minus}(\text{Succ}(x), \text{Succ}(y)) &= \text{minus}(x, y) \\ \text{div}(\text{Zero}, \text{Succ}(y)) &= \text{Zero} \\ \text{div}(\text{Succ}(x), \text{Succ}(y)) &= \text{Succ}(\text{div}(\text{minus}(x, y), \text{Succ}(y))) \\ \text{minus}^\#(\text{Succ}(x), \text{Succ}(y)) &\rightarrow \text{minus}^\#(x, y) \\ \text{div}^\#(\text{Succ}(x), \text{Succ}(y)) &\rightarrow \text{div}^\#(\text{minus}(x, y), \text{Succ}(y)) \end{aligned}$$

- example innermost evaluation

$$\begin{aligned} &\text{div}(\text{Succ}(\text{Zero}), \text{Succ}(\text{Zero})) \\ &\stackrel{i}{\rightarrow} \text{Succ}(\text{div}(\text{minus}(\text{Zero}, \text{Zero}), \text{Succ}(\text{Zero}))) \\ &\stackrel{i}{\rightarrow} \text{Succ}(\text{div}(\text{Zero}, \text{Succ}(\text{Zero}))) \\ &\stackrel{i}{\rightarrow} \text{Succ}(\text{Zero}) \end{aligned}$$

and its (partial) representation as *DP*-chain

$$\begin{aligned} &\text{div}^\#(\text{Succ}(\text{Zero}), \text{Succ}(\text{Zero})) \\ &\rightarrow \text{div}^\#(\text{minus}(\text{Zero}, \text{Zero}), \text{Succ}(\text{Zero})) \\ &\stackrel{i}{\rightarrow}^* \text{div}^\#(\text{Zero}, \text{Succ}(\text{Zero})) \end{aligned}$$

## 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 48

## Termination – Subterm Criterion

### A First Termination Technique – The Subterm Criterion

- the **subterm criterion** works as follows
  - let  $P \subseteq DP$
  - choose  $f^\#$ , a symbol of arity  $n$
  - choose some argument position  $i \in \{1, \dots, n\}$
  - demand  $s_i \triangleright t_i$  for all  $f^\#(s_1, \dots, s_n) \rightarrow f^\#(t_1, \dots, t_n) \in P$
  - define  $P_{\triangleright} = \{f^\#(s_1, \dots, s_n) \rightarrow f^\#(t_1, \dots, t_n) \in P \mid s_i \triangleright t_i\}$
  - 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
  - easy to test: just find argument position  $i$  such that each  $i$ -th argument of all  $f^\#$ -dependency pairs decreases w.r.t.  $\triangleright$  and then remove all strictly decreasing pairs
  - incremental method: adding other dependency pairs for  $g^\#$  later on will have no impact
  - can be applied iteratively
  - fast, but limited power

## Subterm Criterion – Example

- consider a program with the following set of dependency pairs

$$\text{ack}^\#(\text{Succ}(x), \text{Zero}) \rightarrow \text{ack}^\#(x, \text{Succ}(\text{Zero})) \quad (1)$$

$$\text{ack}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{ack}^\#(x, \text{ack}(\text{Succ}(x), y)) \quad (2)$$

$$\text{ack}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{ack}^\#(\text{Succ}(x), y) \quad (3)$$

$$\text{minus}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{minus}^\#(x, y) \quad (4)$$

$$\text{div}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{div}^\#(\text{minus}(x, y), \text{Succ}(y)) \quad (5)$$

$$\text{plus}^\#(\text{Succ}(x), y) \rightarrow \text{plus}^\#(y, x) \quad (6)$$

- it is easy to remove (4) by choosing any argument of  $\text{minus}^\#$
- we can remove (1) and (2) by choosing argument 1 of  $\text{ack}^\#$
- afterwards we can remove (3) by choosing argument 2 of  $\text{ack}^\#$
- it is not possible to remove any of the remaining dependency pairs (5) and (6) by the subterm criterion

## Subterm Criterion – Soundness Proof

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

$$s_1\sigma_1 \rightarrow t_1\sigma_1 \xrightarrow{c_i^*} s_2\sigma_2 \rightarrow t_2\sigma_2 \xrightarrow{c_i^*} s_3\sigma_3 \rightarrow t_3\sigma_3 \xrightarrow{c_i^*} \dots$$

such that all  $s_j \rightarrow t_j \in P$ , all  $s_j$  and  $t_j$  have  $f^\#$  as root and there are infinitely many  $s_j \rightarrow t_j \in P_{\triangleright}$ ; perform proof by contradiction

- hence all  $s_j \rightarrow t_j$  are of the form  $f^\#(s_{j,1}, \dots, s_{j,n}) \rightarrow f^\#(t_{j,1}, \dots, t_{j,n})$
- from condition  $s_{j,i} \triangleright t_{j,i}$  of criterion conclude  $s_{j,i}\sigma_j \triangleright t_{j,i}\sigma_j$  and if  $s_j \rightarrow t_j \in P_{\triangleright}$  then  $s_{j,i} \triangleright t_{j,i}$  and thus  $s_{j,i}\sigma_j \triangleright t_{j,i}\sigma_j$
- we further know  $t_{j,i}\sigma_j \xrightarrow{c_i^*} s_{j+1,i}\sigma_{j+1}$  since  $f^\#$  is a constructor
- this implies  $t_{j,i}\sigma_j = s_{j+1,i}\sigma_{j+1}$  since  $t_{j,i}\sigma_j \in NF(\leftarrow)$  as  $t_{j,i}\sigma_j \triangleleft s_{j,i}\sigma_j \triangleleft f^\#(s_{j,1}\sigma_j, \dots, s_{j,n}\sigma_j) = s_j\sigma_j \in NF(\leftarrow)$
- obtain an infinite sequence with infinitely many  $\triangleright$ ; this is a contradiction to  $SN(\triangleright)$

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

## Termination – Size-Change Principle

### The Size-Change Principle

- the size-change principle abstracts decreases of arguments into size-change graphs
- size-change graph
  - let  $f^\#$  be a symbol of arity  $n$
  - a size-change graph for  $f^\#$  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  $\succsim$
  - the size-change graph  $G$  of a dependency pair  $f^\#(s_1, \dots, s_n) \rightarrow f^\#(t_1, \dots, t_n)$  defines  $E$  as follows
    - $i_{in} \xrightarrow{\succ} j_{out} \in E$  whenever  $s_i \triangleright t_j$  (strict decrease)
    - $i_{in} \xrightarrow{\succsim} j_{out} \in E$  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

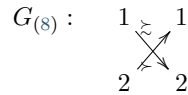
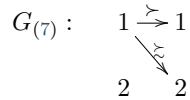
### Example – Size-Change Graphs

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

$$f^\sharp(\text{Succ}(x), y) \rightarrow f^\sharp(x, \text{Succ}(x)) \quad (7)$$

$$f^\sharp(x, \text{Succ}(y)) \rightarrow f^\sharp(y, x) \quad (8)$$

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

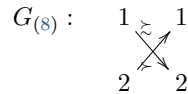
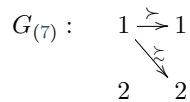


### Multigraphs and Concatenation

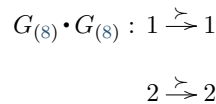
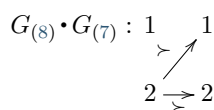
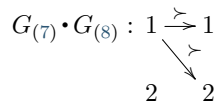
- graphs can be glued together, tracing size-changes in chains, i.e., subsequent dependency pairs
- definition: let  $\mathcal{G}$  be a set of size-change graphs for the same symbol  $f^\sharp$ ; then the set of **multigraphs** for  $f^\sharp$  is defined as follows
  - every  $G \in \mathcal{G}$  is a multigraph
  - whenever there are multigraphs  $G_1$  and  $G_2$  with edges  $E_1$  and  $E_2$  then also the **concatenated graph**  $G = G_1 \cdot G_2$  is a multigraph; here, the edges of  $E$  of  $G$  are defined as
    - if  $i \rightarrow j \in E_1$  and  $j \rightarrow k \in E_2$ , then  $i \rightarrow k \in E$
    - if at least one of the edges  $i \rightarrow j$  and  $j \rightarrow k$  is labeled with  $\succ$  then  $i \rightarrow k$  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

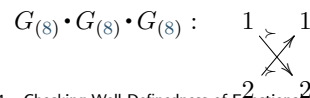
- consider size-change graphs



- this leads to three maximal multigraphs



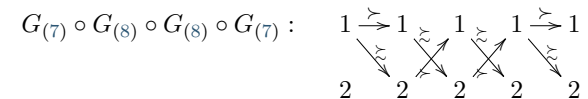
- and a non-maximal multigraph



### 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  $G_1 \circ G_2 \circ G_3 \circ \dots$

- example:



- 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

## Deciding Size-Change Termination

- 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  $\overset{\sim}{\rightarrow}$ -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  $i \overset{\sim}{\rightarrow} i$
- 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

- 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

## 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  $G = G_1 \cdot \dots \cdot G_n$  with each  $G_i \in \mathcal{G}$
- consider infinite graph  $G_1 \circ \dots \circ G_n \circ G_1 \circ \dots \circ G_n \circ \dots$
- because of size-change termination, this graph contains path with infinitely many  $\overset{\sim}{\rightarrow}$ -edges
- hence  $G \circ G \circ \dots$  also has a path with infinitely many  $\overset{\sim}{\rightarrow}$ -edges
- on this path some index  $i$  must be visited infinitely often
- hence there is a path of length  $k$  such that  $G \circ G \circ \dots \circ G$  ( $k$ -times) contains a path from the leftmost argument  $i$  to the rightmost argument  $i$  with at least one  $\overset{\sim}{\rightarrow}$ -edge
- consequently  $G \cdot G \cdot \dots \cdot G$  ( $k$ -times) contains an edge  $i \overset{\sim}{\rightarrow} i$
- by maximality,  $G = G \cdot G \cdot \dots \cdot G$ , and thus  $G$  contains an edge  $i \overset{\sim}{\rightarrow} i$

## 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:

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

- **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.,  $c : X^{(n)} \rightarrow C$
  - **theorem**: there exists an infinite subset  $Y \subseteq X$  such that all size  $n$  sets of  $Y$  have the same color

## Proof of Theorem: Hard Direction (2. implies 1.)

- consider some arbitrary infinite graph  $G_0 \circ G_1 \circ G_2 \circ \dots$
- for  $n < m$  define  $G_{n,m} = G_n \cdot \dots \cdot G_{m-1}$
- 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  $n < m$   
( $n = 2$ ,  $C = \text{multigraphs}$ ,  $X = \mathbb{N}$ ,  $c(\{n, m\}) = G_{\min\{n,m\}, \max\{n,m\}}$ )
- $G$  is maximal: for  $n_1 < n_2 < n_3$  with  $\{n_1, n_2, n_3\} \subseteq I$ , we have  
 $G_{n_1, n_3} = G_{n_1} \cdot \dots \cdot G_{n_2-1} \cdot G_{n_2} \cdot \dots \cdot G_{n_3-1} = G_{n_1, n_2} \cdot G_{n_2, n_3}$ , and thus  $G = G \cdot G$
- by assumption,  $G$  contains edge  $i \xrightarrow{\succ} i$
- let  $I = \{n_1, n_2, \dots\}$  with  $n_1 < n_2 < \dots$  and obtain

$$\begin{aligned} & G_0 \circ G_1 \circ \dots \\ & = G_0 \circ \dots \circ G_{n_1-1} \circ G_{n_1} \circ \dots \circ G_{n_2-1} \circ G_{n_2} \circ \dots \circ G_{n_3-1} \circ \dots \\ & \sim G_0 \circ \dots \circ G_{n_1-1} \circ G \qquad \qquad \qquad \circ G \qquad \qquad \qquad \circ \dots \end{aligned}$$

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

## 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.,  $c : X^{(n)} \rightarrow C$
- 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  $c' : Y_0^{(m)} \rightarrow C$  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  $a_0, a_1, a_2, \dots$ , colors  $c_0, c_1, c_2 \dots$  and sets  $X_0, X_1, X_2, \dots$  satisfying the above properties
- since  $C$  is finite there must be some color  $d$  in the infinite list  $c_0, c_1, \dots$  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  $\{a_{i_1}, \dots, a_{i_n}\}$  with  $i_1 < \dots < i_n$ ; hence,  $Z = \{a_{i_1}\} \cup Z'$  with  $Z' \in X_{i_1+1}^{(m)}$ , i.e.,  $c(Z) = c_{i_1} = d$

## 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 in the following sense:
  - if **all** DPs can be deleted by subterm criterion, then also size-change principle is successful
- still no handling of defined symbols in dependency pairs as in

$$\text{div}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{div}^\#(\text{minus}(x, y), \text{Succ}(y))$$

## Termination – Reduction Pairs

### Reduction Pairs

- recall definition:  $P$ -chain is sequence

$$s_1\sigma_1 \rightarrow t_1\sigma_1 \xrightarrow{i}^* s_2\sigma_2 \rightarrow t_2\sigma_2 \xrightarrow{i}^* s_3\sigma_3 \rightarrow t_3\sigma_3 \xrightarrow{i}^* \dots$$

such that all  $s_i \rightarrow t_i \in P$  and all  $s_i\sigma_i \in NF(\leftrightarrow)$

- 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  $s_i\sigma \in NF(\leftrightarrow)$  and  $\triangleright$  to turn  $\xrightarrow{i}^*$  into  $=$
- now generalize  $\triangleright$  to strongly normalizing relation  $\succ$
- now demand  $\ell \succ r$  for equations to ensure decrease  $t_i\sigma_i \succ s_{i+1}\sigma_{i+1}$
- definition: **reduction pair**  $(\succ, \succsim)$  is pair of relations such that
  - $SN(\succ)$
  - $\succsim$  is transitive
  - $\succ$  and  $\succsim$  are compatible:  $\succ \circ \succsim \subseteq \succ$
  - both  $\succ$  and  $\succsim$  are closed under substitutions:  $s \succsim t \rightarrow s\sigma \succsim t\sigma$
  - $\succ$  is closed under contexts:  $s \succ t \rightarrow F(\dots, s, \dots) \succ F(\dots, t, \dots)$
  - note:  $\succ$  does not have to be closed under contexts

### Applying Reduction Pairs

- recall definition:  $P$ -chain is sequence

$$s_1\sigma_1 \rightarrow t_1\sigma_1 \xrightarrow{i}^* s_2\sigma_2 \rightarrow t_2\sigma_2 \xrightarrow{i}^* s_3\sigma_3 \rightarrow t_3\sigma_3 \xrightarrow{i}^* \dots$$

such that all  $s_i \rightarrow t_i \in P$  and all  $s_i\sigma \in NF(\leftrightarrow)$

- demand  $s \succ t$  for all  $s \rightarrow t \in P$  to ensure  $s_i\sigma_i \succ t_i\sigma_i$
- demand  $\ell \succ r$  for all equations to ensure  $t_i\sigma_i \succ s_{i+1}\sigma_{i+1}$
- define  $P_\succ = \{s \rightarrow t \in P \mid s \succ t\}$
- 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

### Example

- remaining termination problem

$$\text{minus}(x, \text{Zero}) = x$$

$$\text{minus}(\text{Succ}(x), \text{Succ}(y)) = \text{minus}(x, y)$$

$$\text{div}(\text{Zero}, \text{Succ}(y)) = \text{Zero}$$

$$\text{div}(\text{Succ}(x), \text{Succ}(y)) = \text{Succ}(\text{div}(\text{minus}(x, y), \text{Succ}(y)))$$

$$\text{div}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{div}^\#(\text{minus}(x, y), \text{Succ}(y))$$

- constraints

$$\text{minus}(x, \text{Zero}) \succ x$$

$$\text{minus}(\text{Succ}(x), \text{Succ}(y)) \succ \text{minus}(x, y)$$

$$\text{div}(\text{Zero}, \text{Succ}(y)) \succ \text{Zero}$$

$$\text{div}(\text{Succ}(x), \text{Succ}(y)) \succ \text{Succ}(\text{div}(\text{minus}(x, y), \text{Succ}(y)))$$

$$\text{div}^\#(\text{Succ}(x), \text{Succ}(y)) \succ \text{div}^\#(\text{minus}(x, y), \text{Succ}(y))$$

$$\text{div}^\#(\text{Succ}(x), \text{Succ}(y)) \rightarrow \text{div}^\#(\text{minus}(x, y), \text{Succ}(y))$$

- requiring  $\ell \succsim r$  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 DP only requires **minus**
- definition: the **usable symbols**  $US$  w.r.t. a set  $P$  are those defined symbols that occur in rhss of  $P$  (no  $\#$ -symbols(!)), or that occur in rhss of defining equations of usable symbols;

- $US(x) = \emptyset$
- $US(F(t_1, \dots, t_n)) = US(t_1) \cup \dots \cup US(t_n) \cup \begin{cases} \{F\}, & \text{if } F \in \mathcal{D} \\ \emptyset, & \text{otherwise} \end{cases}$
- assuming  $\mathcal{E}$  is the set of equations of a program, then  $US(P)$  is defined by

$$\frac{s \rightarrow t \in P}{US(t) \subseteq US(P)} \qquad \frac{f \in US(P) \quad f(\dots) = r \in \mathcal{E}}{US(r) \subseteq US(P)}$$

- the **usable equations** of  $P$  are  $\mathcal{U}(P) = \{f(\dots) = r \in \mathcal{E} \mid f \in US(P)\}$
- theorem: whenever  $t_i \sigma_i \xrightarrow{\text{ci}^*} s_{i+1} \sigma_{i+1}$  in chain, then only  $\mathcal{U}(\{s_i \rightarrow t_i\})$  can be used

Applying Reduction Pairs with Usable Equations

- recall definition:  $P$ -chain is sequence

$$s_1 \sigma_1 \rightarrow t_1 \sigma_1 \xrightarrow{\text{ci}^*} s_2 \sigma_2 \rightarrow t_2 \sigma_2 \xrightarrow{\text{ci}^*} s_3 \sigma_3 \rightarrow t_3 \sigma_3 \xrightarrow{\text{ci}^*} \dots$$

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

- choose a symbol  $f^\#$  and define  $P_{f^\#} = \{s \rightarrow t \in P \mid \text{root } s = f^\#\}$
- demand  $s \succsim t$  for all  $s \rightarrow t \in P_{f^\#}$
- demand  $\ell \succsim r$  for all  $\ell = r \in \mathcal{U}(P_{f^\#})$
- define  $P_{\succ} = \{s \rightarrow t \in P_{f^\#} \mid s \succ t\}$
- 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

Example with Usable Equations

- remaining termination problem

$$\begin{aligned} \text{minus}(x, \text{Zero}) &= x \\ \text{minus}(\text{Succ}(x), \text{Succ}(y)) &= \text{minus}(x, y) \\ \text{div}(\text{Zero}, \text{Succ}(y)) &= \text{Zero} \\ \text{div}(\text{Succ}(x), \text{Succ}(y)) &= \text{Succ}(\text{div}(\text{minus}(x, y), \text{Succ}(y))) \\ \text{div}^\#(\text{Succ}(x), \text{Succ}(y)) &\rightarrow \text{div}^\#(\text{minus}(x, y), \text{Succ}(y)) \end{aligned}$$

- $US(P_{\text{div}^\#}) = \{\text{minus}\}$ , so constraints are

$$\begin{aligned} \text{minus}(x, \text{Zero}) &\succsim x \\ \text{minus}(\text{Succ}(x), \text{Succ}(y)) &\succsim \text{minus}(x, y) \\ \text{div}^\#(\text{Succ}(x), \text{Succ}(y)) &\succ \text{div}^\#(\text{minus}(x, y), \text{Succ}(y)) \end{aligned}$$

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

Remaining Problem

- given constraints

$$\begin{aligned} \text{minus}(x, \text{Zero}) &\succsim x \\ \text{minus}(\text{Succ}(x), \text{Succ}(y)) &\succsim \text{minus}(x, y) \\ \text{div}^\#(\text{Succ}(x), \text{Succ}(y)) &\succ \text{div}^\#(\text{minus}(x, y), \text{Succ}(y)) \end{aligned}$$

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)

## Polynomial Interpretation

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

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

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

- interpretation is lifted to terms by composing polynomials

$$\begin{aligned} \llbracket x \rrbracket &= x \\ \llbracket F(t_1, \dots, t_n) \rrbracket &= p_F(\llbracket t_1 \rrbracket, \dots, \llbracket t_n \rrbracket) \end{aligned}$$

- $(\succ, \succsim)$  is defined as
 
$$s \succsim t \text{ iff } \forall \vec{x} \in \mathbb{N}^*. \llbracket s \rrbracket (\geq) \llbracket t \rrbracket$$
- $(\succ, \succsim)$  is a reduction pair, e.g.,
  - $SN(\succ)$  follows from strong-normalization of  $>$  on  $\mathbb{N}$
  - $\succsim$  is closed under contexts since each  $p_F$  is weakly monotone

## Example – Polynomial Interpretation

- given constraints

$$\begin{aligned} \text{minus}(x, \text{Zero}) &\succsim x \\ \text{minus}(\text{Succ}(x), \text{Succ}(y)) &\succsim \text{minus}(x, y) \\ \text{div}^\#(\text{Succ}(x), \text{Succ}(y)) &\succ \text{div}^\#(\text{minus}(x, y), \text{Succ}(y)) \end{aligned}$$

and polynomial interpretation

$$\begin{aligned} p_{\text{minus}}(x_1, x_2) &= x_1 \\ p_{\text{Zero}} &= 2 \\ p_{\text{Succ}}(x_1) &= 1 + x_1 \\ p_{\text{div}^\#}(x_1, x_2) &= x_1 + 3x_2 \end{aligned}$$

we obtain polynomial constraints

$$\begin{aligned} \llbracket \text{minus}(x, \text{Zero}) \rrbracket &= x \geq x = \llbracket x \rrbracket \\ \llbracket \text{minus}(\text{Succ}(x), \text{Succ}(y)) \rrbracket &= 1 + x \geq x = \llbracket \text{minus}(x, y) \rrbracket \\ \llbracket \text{div}^\#(\text{Succ} \dots) \rrbracket &= 4 + x + 3y > 3 + x + 3y = \llbracket \text{div}^\#(\text{minus} \dots) \rrbracket \end{aligned}$$

## Solving Polynomial Constraints

- each polynomial constraint over  $\mathbb{N}$  can be brought into simple form “ $p \geq 0$ ” for some polynomial  $p$ 
  - replace  $p_1 > p_2$  by  $p_1 \geq p_2 + 1$
  - replace  $p_1 \geq p_2$  by  $p_1 - p_2 \geq 0$
- the question of “ $p \geq 0$ ” 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 |
|---------------------------|------------|
| $x \geq x$                | $0 \geq 0$ |
| $1 + x \geq x$            | $1 \geq 0$ |
| $4 + x + 3y > 3 + x + 3y$ | $0 \geq 0$ |

## Finding Polynomial Interpretations

- in division example, interpretation was given on slides
- aim: search for suitable interpretation
- approach: perform everything symbolically

## Symbolic Polynomial Interpretations

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

$$p_F(x_1, \dots, x_n) = F_0 + F_1x_1 + \dots + F_nx_n$$

where the  $F_i$  are symbolic coefficients

- 

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

$$p_{\text{Zero}} = 2$$

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

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

concrete interpretation above becomes symbolic

$$p_{\text{minus}}(x_1, x_2) = m_0 + m_1x_1 + m_2x_2$$

$$p_{\text{Zero}} = Z_0$$

$$p_{\text{Succ}}(x_1) = S_0 + S_1x_1$$

$$p_{\text{div}^\#}(x_1, x_2) = d_0 + d_1x_1 + d_2x_2$$

## Symbolic Polynomial Constraints

- given constraints

$$\text{minus}(x, \text{Zero}) \succsim x$$

$$\text{minus}(\text{Succ}(x), \text{Succ}(y)) \succsim \text{minus}(x, y)$$

$$\text{div}^\#(\text{Succ}(x), \text{Succ}(y)) \succ \text{div}^\#(\text{minus}(x, y), \text{Succ}(y))$$

- obtain symbolic polynomial constraints

$$m_0 + m_1x + m_2Z_0 \geq x$$

$$m_0 + m_1(S_0 + S_1x) + m_2(S_0 + S_1y) \geq m_0 + m_1x + m_2y$$

$$d_0 + d_1(S_0 + S_1x) + d_2(S_0 + S_1y) > d_0 + d_1(m_0 + m_1x + m_2y) + d_2(S_0 + S_1y)$$

- and simplify to

$$(m_0 + m_2Z_0) + (m_1 - 1)x \geq 0$$

$$(m_1S_0 + m_2S_0) + (m_1S_1 - m_1)x + (m_2S_1 - m_2)y \geq 0$$

$$(d_1S_0 - d_1m_0 - 1) + (d_1S_1 - d_1m_1)x + (-d_1m_2)y \geq 0$$

## Absolute Positiveness – Symbolic Example

- on symbolic polynomial constraints

$$(m_0 + m_2Z_0) + (m_1 - 1)x \geq 0$$

$$(m_1S_0 + m_2S_0) + (m_1S_1 - m_1)x + (m_2S_1 - m_2)y \geq 0$$

$$(d_1S_0 - d_1m_0 - 1) + (d_1S_1 - d_1m_1)x + (-d_1m_2)y \geq 0$$

absolute positiveness works as before; obtain constraints

$$m_0 + m_2Z_0 \geq 0 \quad m_1 - 1 \geq 0$$

$$m_1S_0 + m_2S_0 \geq 0 \quad m_1S_1 - m_1 \geq 0 \quad m_2S_1 - m_2 \geq 0$$

$$d_1S_0 - d_1m_0 - 1 \geq 0 \quad d_1S_1 - d_1m_1 \geq 0 \quad -d_1m_2 \geq 0$$

- 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  $m_0 \geq 0, m_1 \geq 0, m_2 \geq 0, S_0 \geq 0, S_1 \geq 0, Z_0 \geq 0, \dots$

## Constraint Solving by Hand – Example

- original constraints

$$m_0 + m_2Z_0 \geq 0$$

$$m_1 - 1 \geq 0$$

$$m_1S_0 + m_2S_0 \geq 0$$

$$m_1S_1 - m_1 \geq 0$$

$$m_2S_1 - m_2 \geq 0$$

$$d_1S_0 - d_1m_0 - 1 \geq 0$$

$$d_1S_1 - d_1m_1 \geq 0$$

$$-d_1m_2 \geq 0$$

- delete trivial constraints

$$m_1 - 1 \geq 0$$

$$m_1S_1 - m_1 \geq 0$$

$$m_2S_1 - m_2 \geq 0$$

$$d_1S_0 - d_1m_0 - 1 \geq 0$$

$$d_1S_1 - d_1m_1 \geq 0$$

$$-d_1m_2 \geq 0$$

- conclusions

$$m_1 \geq 1$$

$$d_1 \geq 1$$

$$S_0 \geq 1$$

$$S_1 \geq 1$$

$$m_2 = 0$$

$$S_1 \geq m_1$$

$$m_0 = 0$$

## Constraint Solving by SMT-Solver – Example

- original constraints

$$\begin{array}{lll}
 m_0 + m_2 Z_0 \geq 0 & m_1 - 1 \geq 0 & \\
 m_1 S_0 + m_2 S_0 \geq 0 & m_1 S_1 - m_1 \geq 0 & m_2 S_1 - m_2 \geq 0 \\
 d_1 S_0 - d_1 m_0 - 1 \geq 0 & d_1 S_1 - d_1 m_1 \geq 0 & -d_1 m_2 \geq 0
 \end{array}$$

- encode as SMT problem in file `division.smt2`

```

(set-logic QF_NIA)
(declare-fun m0 () Int) ... (declare-fun d2 () Int)
(assert (>= m0 0)) ... (assert (>= d2 0))
(assert (>= (+ m0 (* m2 Z0)) 0))
...
(assert (>= (* (- 1) d1 m2) 0))
(check-sat)
(get-model)
(exit)

```

## 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 S0 () Int 8)
  (define-fun Z0 () Int 0)
  (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)
)

```

- 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
  - certification will be handled in more detail in Part 7 of this lecture

## 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