

Efficient Deductive Methods for Program Analysis

Harald Ganzinger

Max-Planck-Institut für Informatik

- program analysis from high-level inference rules
- complexity analysis through general meta-complexity theorems
- logical aspects of fundamental algorithmic paradigms (dynamic programming, union-find, congruence closure)
- treatment of transitive relations: implication, **equivalence**, **congruence**, **quasi-orderings**
- avoiding the cubic-time bottleneck
- variable-free specializations of fundamental first-order methods: resolution, Knuth/Bendix-completion, **ordered chaining**
- closely related to McAllester's SAS'99 talk and paper

Linear-time analyses

Example: interprocedural reachability

Logic background: linear-time bottom-up deduction

Analyses for type congruences

Examples:

Steensgaard's pointer analysis ($O(n \log n)$)

Henglein's subtype analysis ($O(n^2)$)

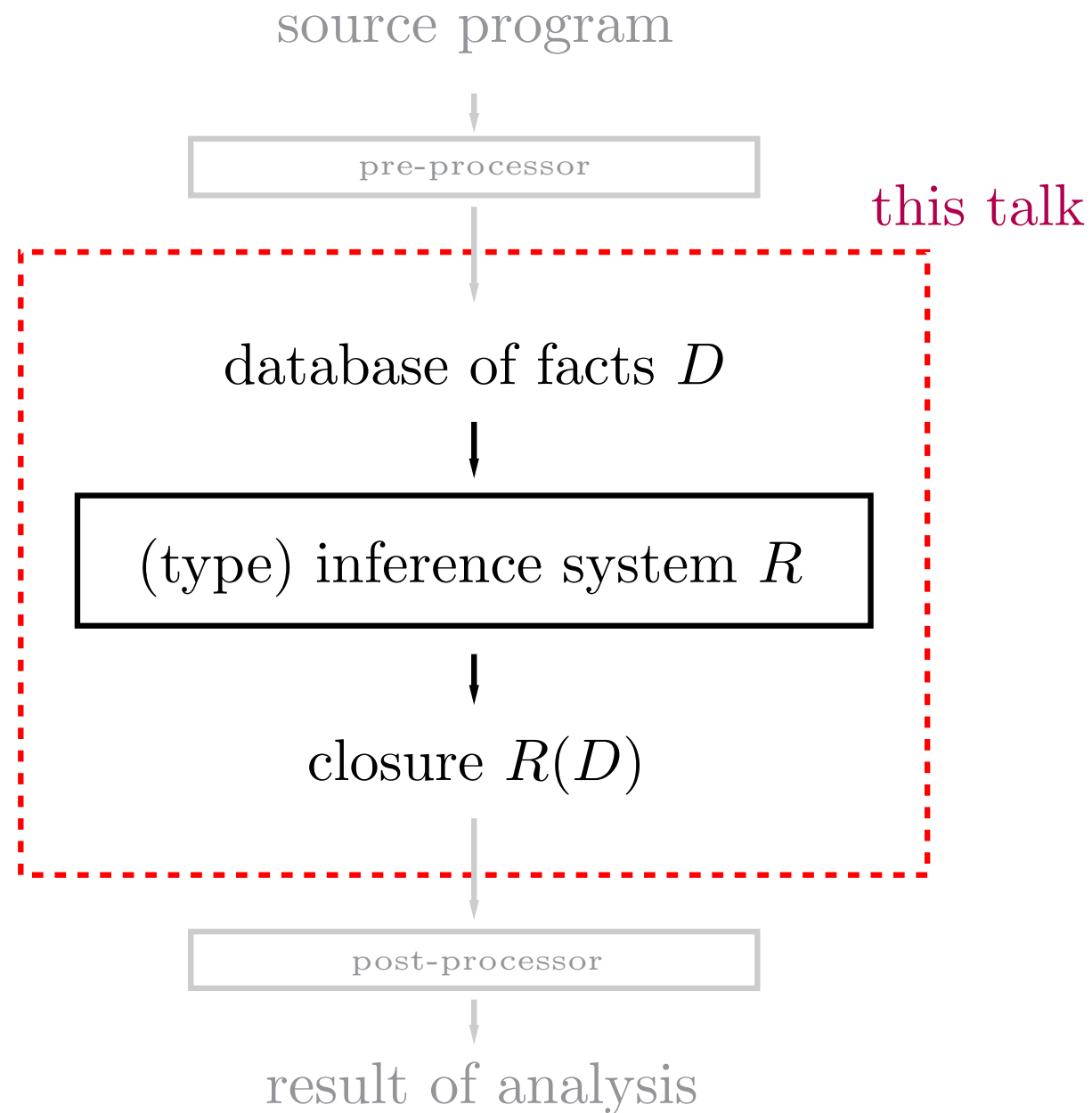
Logic background: congruence closure for Horn clauses

Dynamic transitive closure

Example: Andersen's pointer analysis via atomic set constraints

Logic background: ordered chaining

I. Linear-Time Analyses



Example

program

```
1 procedure main
2 begin
3   declare x: int
4   read(x)
5   call p(x)
6 end

7 procedure p(a:int)
8 begin
9   if a>0 then
10    read(g)
11    a:=a-g
12    call p(a)
13    print(a)
14  fi
15 end
```

facts

```
proc(main,2,6)
next(main,2,5)
call(main,p,5,6)

proc(p,8,15)
next(p,8,12)
call(p,p,12,13)
next(p,13,15)
next(p,8,15)
```

Read “ $L \Rightarrow L'$ in P ” as “ L' can be reached from L in procedure P ”.

$$\begin{array}{c}
 \text{next}(Q, L, L') \\
 X \Rightarrow L \text{ in } Q \\
 \hline
 X \Rightarrow L' \text{ in } Q
 \end{array}
 \qquad
 \begin{array}{c}
 \text{call}(Q, P, L_c, L_r) \\
 \text{proc}(P, L_0, L_f) \\
 L_0 \Rightarrow L_f \text{ in } P \\
 X \Rightarrow L_c \text{ in } Q \\
 \hline
 X \Rightarrow L_r \text{ in } Q
 \end{array}
 \qquad
 \begin{array}{c}
 \text{proc}(P, L_0, L_f) \\
 \hline
 L_0 \Rightarrow L_0 \text{ in } P
 \end{array}$$

THEOREM 1.1 $IPR(D)$ can be computed in time $O(|D|)$.

[$|D|$ = size of D = number of nodes in tree representation]

THEOREM 1.2 (MCALLESTER 1999) Let R be an inference system such that $R(D)$ is finite. Then $R(D)$ can be computed in time $O(|R(D)| + \text{pf}_R(R(D)))$.

$\text{pf}_R(R(D))$ is the number of **prefix firings** of R on $R(D)$:

$$\text{pf}_R(D) = \left| \left\{ (r, i, \sigma) \mid r = A_1 \wedge \dots \wedge A_i \wedge \dots \wedge A_n \supset A_0 \in R \right. \right. \\ \left. \left. A_j \sigma \in D, \text{ for } 1 \leq j \leq i \right\} \right|$$

COROLLARY 1.3 (DOWLING, GALLIER 1984) If R is ground, $R(D)$ can be computed in time $O(|D| + |R|)$.

Let $n = |D|$.

$$\frac{\text{proc}(P, L_0, L_f)}{L_0 \Rightarrow L_0 \text{ in } P}$$

has $O(n)$ (prefix) firings.^a

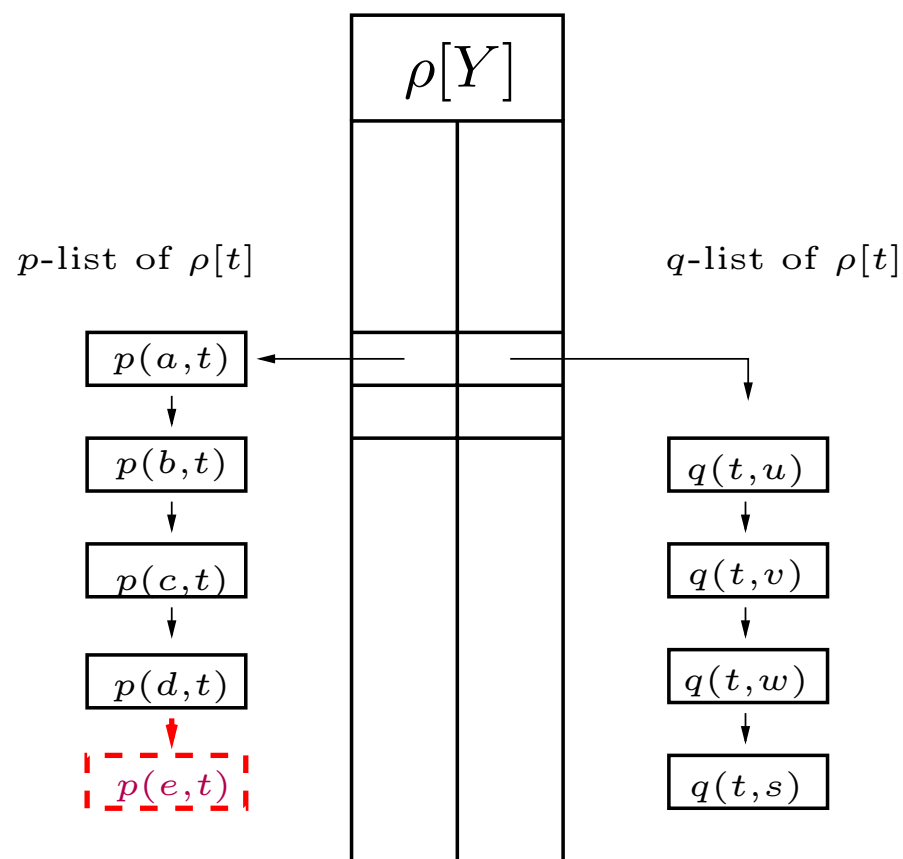
$\frac{\begin{array}{ll} \text{next}(Q, L, L') & O(n) * \\ X \Rightarrow L \text{ in } Q & O(1) \end{array}}{X \Rightarrow L' \text{ in } Q}$	$\frac{\begin{array}{ll} \text{call}(Q, P, L_c, R_r) & O(n) * \\ \text{proc}(P, L_0, L_f) & O(1) * \\ L_0 \Rightarrow L_f \text{ in } P & O(1) * \\ X \Rightarrow L_c \text{ in } Q & O(1) \end{array}}{X \Rightarrow L_r \text{ in } Q}$
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THEOREM 1.4 $IPR(D)$ can be computed in time $O(|D|)$.

Beweis. Both $|IPR(D)|$ and $\text{pf}_{IPR}(IPR(D))$ are in $O(|D|)$. \square

^aOnly facts $X \Rightarrow Y$ in P where X is the start label in P can be derived.

Data structure for rules ρ of the form $p(X, Y) \wedge q(Y, Z) \supset r(X, Y, Z)$



Upon adding a fact $p(e, t)$, fire all $r(e, t, z)$, for z on the *q*-list of $A[t]$.

The inference system can be transformed (maintaining **pf**) so that it contains unary rules and binary rules of the form ρ .

Problems

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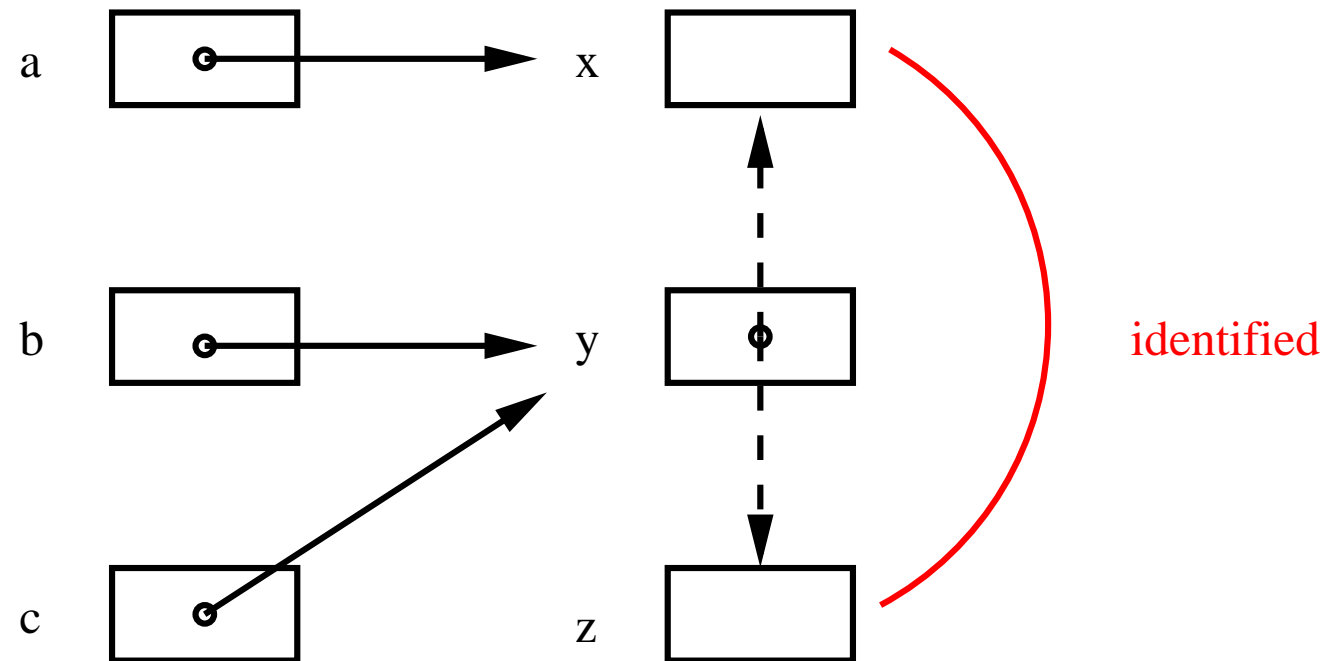
- if $R(D)$ infinite, consider $R(D) \cap \text{atoms}(\text{subterms}(D))$
 \Rightarrow concept of local inferences (Givan, McAllester 1993)
- in the presence of transitive relations, complexity is in $\Omega(n^3)$

II. Equivalence and Congruence

program

```
a = &x  
b = &y  
if ... then  
    y = &x;  
else  
    y = &z  
fi  
c = &y
```

shape graph



THEOREM 2.5 (STEENSGAARD 1996) Shape graphs can be computed in time $O(n\alpha(n, n))$.

assignments

$$\begin{array}{c}
 \text{input}(X = \&Y) \\
 X : \text{ref}(T_x) \\
 Y : T_y \\
 \hline
 T_x \doteq T_y
 \end{array}
 \qquad
 \begin{array}{c}
 \text{input}(X = Y) \\
 X : \text{ref}(T_x) \\
 Y : \text{ref}(T_y) \\
 \hline
 T_y \leq T_x
 \end{array}$$

subtyping rules

$$\begin{array}{c}
 \hline
 \perp \leq T
 \end{array}
 \qquad
 \begin{array}{c}
 \text{ref}(T) \leq T' \\
 \hline
 \text{ref}(T) \doteq T'
 \end{array}
 \qquad
 \begin{array}{c}
 \text{ref}(T) \doteq \text{ref}(T') \\
 \hline
 T \doteq T'
 \end{array}$$

type equality

$$\begin{array}{c}
 \hline
 T \doteq T
 \end{array}
 \qquad
 \begin{array}{c}
 T \doteq T' \quad T \doteq T'' \\
 \hline
 T'' \doteq T'
 \end{array}
 \qquad
 \begin{array}{c}
 T \doteq T' \quad T' \leq T'' \quad T'' \doteq T''' \\
 \hline
 T \leq T'''
 \end{array}$$

facts from the program

$$\begin{array}{lll} a : \text{ref}(\tau_a) & b : \text{ref}(\tau_b) & c : \text{ref}(\tau_c) \\ x : \text{ref}(\tau_x) & y : \text{ref}(\tau_y) & z : \text{ref}(\tau_z) \end{array}$$

derived equations from the assignments

$$\begin{array}{lll} \tau_a \doteq \text{ref}(\tau_x) & \tau_b \doteq \text{ref}(\tau_y) & \tau_y \doteq \text{ref}(\tau_z) \\ \tau_y \doteq \text{ref}(\tau_x) & \tau_c \doteq \text{ref}(\tau_y) & \end{array}$$

additionally, after computing the closure

$$\text{ref}(\tau_z) \doteq \text{ref}(\tau_x) \quad \tau_z \doteq \tau_x$$

Meta-Complexity Theorem for Horn Clauses with Equality₁₆

THEOREM 2.6 (DOWNEY, SETHI, TARJAN 1980) Let \mathcal{E} be a set of ground equations over terms in \mathcal{T} . Then \mathcal{T}/\mathcal{E} is computable in time $O(n + m \log m)$, with $n = |\mathcal{E}|$ and $m = |\mathcal{T}|$.

THEOREM 2.7 (G, McALLESTER 2001) Let \mathcal{E} be a set of ground **Horn clauses with equality**^a over terms in \mathcal{T} . Then \mathcal{T}/\mathcal{E} is computable in time $O(n + \min(n \log m, m^2))$, with $n = |\mathcal{E}|$ and $m = |\mathcal{T}|$.

COROLLARY 2.8 $SPA(D)$ can be computed in time $O(|D|^2)$.

With some more work we can get it down to $O(n \log n)$.

^aequivalences with some/all compatibility axioms

Language with record types

$$\sigma = [l_1 : \sigma_1; \dots ; l_n : \sigma_n]$$

and subtyping $\sigma \leq \tau$.

Main requirement to check: if $\sigma \leq \tau$ and τ accepts l , then σ accepts l .

Data base contains facts

- $accepts(\sigma, l)$ giving the field labels
- equations $\sigma.l_i \doteq \sigma_i$ for describing component types
- subtype facts of the form $\sigma \leq \tau$

Typing rules:

$$\begin{array}{c}
 \hline
 \sigma \sqsubseteq \sigma
 \end{array}
 \quad
 \begin{array}{c}
 \sigma \leq \tau \\
 \tau \sqsubseteq \rho \\
 \hline
 \sigma \sqsubseteq \rho
 \end{array}
 \quad
 \begin{array}{c}
 \text{accepts}(\sigma, l) \quad \text{accepts}(\tau, l) \\
 \sigma \sqsubseteq \tau \\
 \hline
 \sigma.l \doteq \tau.l
 \end{array}$$

Type equality is an equivalence, plus compatibility axioms:

$$\begin{array}{c}
 \sigma \doteq \tau \\
 \hline
 \sigma.l \doteq \tau.l
 \end{array}
 \quad
 \begin{array}{c}
 \sigma \doteq \sigma' \quad \sigma' \sqsubseteq \tau' \quad \tau' \doteq \tau \\
 \hline
 \sigma \sqsubseteq \tau
 \end{array}$$

THEOREM 2.9 (HENGLEIN 1997) Subtype constraints can be checked in quadratic time.

Beweis. $STA(D)$ can be computed in time $O(|D|^2)$. \square

- extend the Downey, Sethi, Tarjan (1980) algorithm
- alternatively,
 - extend the first meta-complexity theorem to **inference systems with priorities and deletion**

THEOREM 2.10 (G, McALLESTER 2001) Let R be an inference system with priorities and deletion such that **all** closures $R(D)$ are finite. Then **one** closure $R(D)$ can be computed in time $O(|R(D)| + \text{pf}_R(R(D)))$.

- define conditional congruence closure by inferences with priorities and deletion based on ideas by (Bachmair, Tiwari 2000)

Inference system UF (priorities from left to right; premises in $[\dots]$ are deleted after the rule has fired)^a:

$$\begin{array}{c}
 \frac{[x \doteq x]}{\top} \qquad \frac{[x \rightarrow y] \quad y \rightarrow z}{x \rightarrow z} \qquad \frac{[x \doteq y] \quad x \rightarrow z}{x \doteq z} \qquad \frac{
 \begin{array}{c}
 [x \doteq y] \\
 [weight(x, w_1)] \\
 weight(y, w_2) \\
 w_1 \geq w_2
 \end{array}
 }{(y \rightarrow x) \wedge weight(x, w_1 + w_2)}
 \end{array}$$

THEOREM 2.11 Let \mathcal{E} be a set of ground equations over terms in \mathcal{T} . Then $\text{pf}_{UF}(UF(\mathcal{E}))$ is in $O(n \log m)$, with $n = |\mathcal{E}|$ and $m = |\mathcal{T}|$.

With a slightly more sophisticated system we obtain $O(n + m \log m)$.

^aWe also need the symmetric variants of the last two rules, and we assume that initial data bases initialize *weight* by 1.

III. Dynamic Transitive Closure

Basic axioms QO

$$\overline{x \Rightarrow x} \quad \frac{x \Rightarrow x' \quad x' \Rightarrow x''}{x \Rightarrow x''} \quad \frac{x \Rightarrow x'}{f(x) \Rightarrow f(x')} \quad \text{for certain } f$$

optionally exploiting the induced congruence

$$\frac{x \Rightarrow y \quad y \Rightarrow x}{x \dot{=} y}$$

additionally, for atomic set constraints (Melski, Reps 1997):

$$\frac{f(x) \Rightarrow f(y)}{x \Rightarrow y}$$

additionally, from pointer analysis:

$$\frac{\text{input}(X = Y) \quad X : \text{ref}(T) \quad Y : \text{ref}(T')}{T' \Rightarrow T}$$

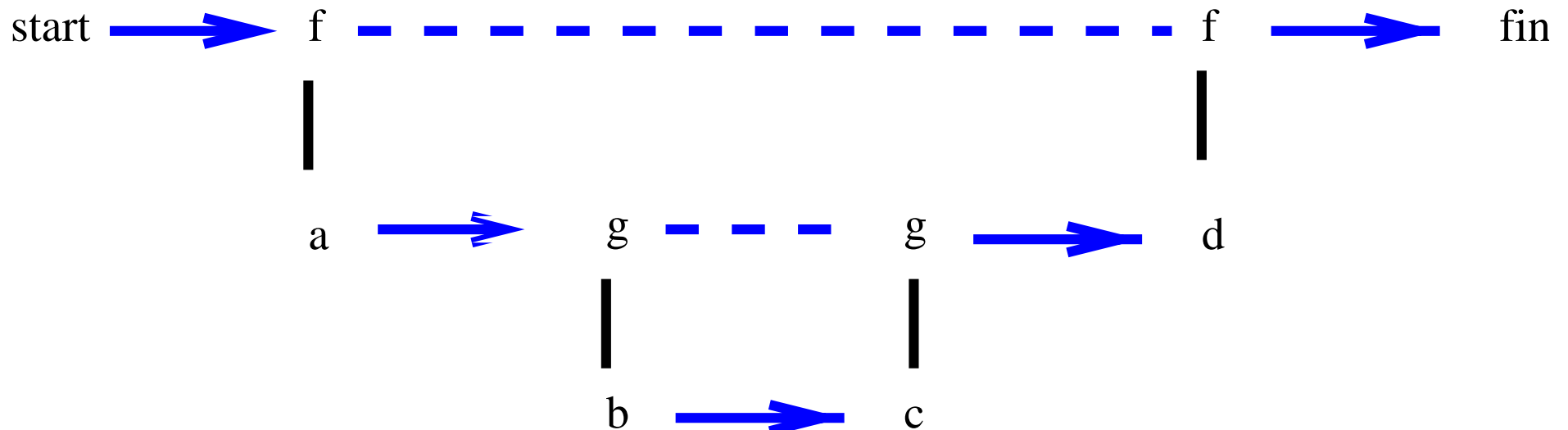
Decision problem:

$$QO \models (s_1 \Rightarrow t_1) \wedge \dots \wedge (s_n \Rightarrow t_n) \supset (s_0 \Rightarrow t_0) \quad (s_i, t_i \text{ ground})$$

Example:

$$(\text{start} \Rightarrow fa) \wedge (a \Rightarrow gb) \wedge (b \Rightarrow c) \wedge (gc \Rightarrow d) \wedge (fd \Rightarrow \text{fin}) \supset (\text{start} \Rightarrow \text{fin})$$

Graphically:



- GMR is 2NPDA-complete (Neal 1989)^a
- 2NPDA acceptance is in $O(n^3)$ (Aho, Hopcroft, Ullman 1968)
- no subcubic algorithm known
- QO (also non-monadic) is a local theory, that is,
 $QO \models C$ iff $QO[\text{subterms in } C] \models C$,
 thus in $O(n^3)$ by (Dowling, Gallier 1980)

$$\begin{array}{c}
 \frac{a \Rightarrow gb}{\frac{\frac{b \Rightarrow c}{gb \Rightarrow gc} \quad gc \Rightarrow d}{gb \Rightarrow d}} \\
 \frac{\text{start} \Rightarrow fa \quad \frac{a \Rightarrow d}{fa \Rightarrow fd}}{\text{start} \Rightarrow fd} \quad fd \Rightarrow \text{fin} \\
 \hline
 \text{start} \Rightarrow \text{fin}
 \end{array}$$

^aThis holds for flat terms already.

Many Data Flow Problems are Equivalent with GMR

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- atomic set constraints (Melski, Reps 1997)
- interprocedural reachability for higher-order languages (Heintze, McAllester 1997)
- Amadio/Cardelli typability (Heintze, McAllester 1997)
- Andersen's (1994) pointer analysis (Aiken et al 1998)

Issue: better balancing of forward and backward computation

History: • Bledsoe, Kunen, Shostak (1985), Hines (1992):

limes theorems, set theory

- Levy, Agustí (1993): bi-rewriting for distributive lattices
- Bachmair, G (1996): ordered chaining for binary relations

Assumption: ground terms are ordered by \succ (total, well-founded, ...)

Ordered Chaining *OC*:

$$\frac{y \Rightarrow x \quad u[x] \Rightarrow v}{u[y] \Rightarrow v} \text{ if } x \succ y \text{ and } u \succ v$$

(Ground) reachability through rewrite proofs: ^a

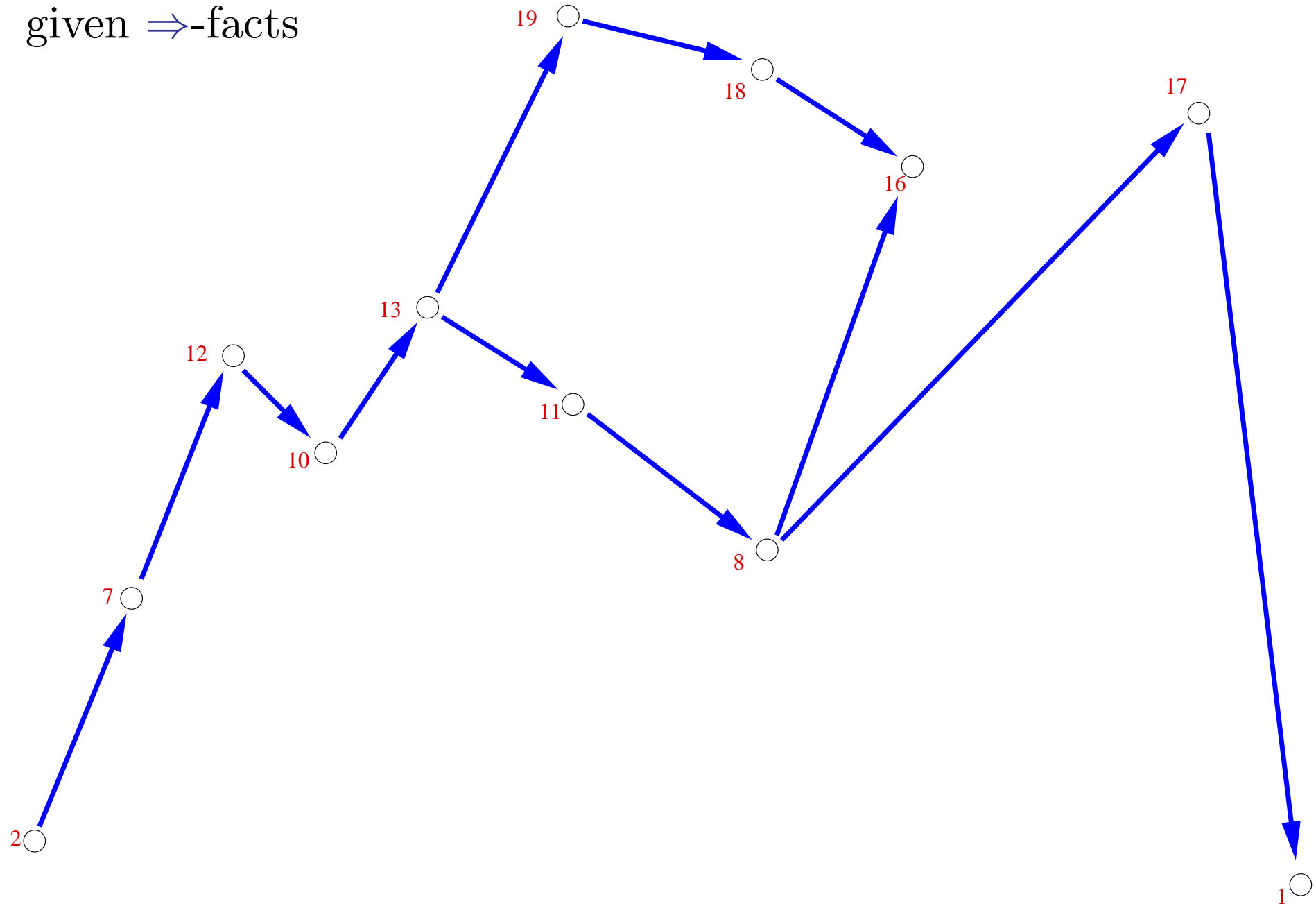
$QO \models D \supset (s \Rightarrow t)$ iff $s \overset{\vee}{\Rightarrow} t$ in $OC(D)$, that is,

$$s \underset{\succ}{\Rightarrow} \dots \underset{\succ}{\Rightarrow} w \underset{\succ}{\Rightarrow} \dots \underset{\succ}{\Rightarrow} t$$

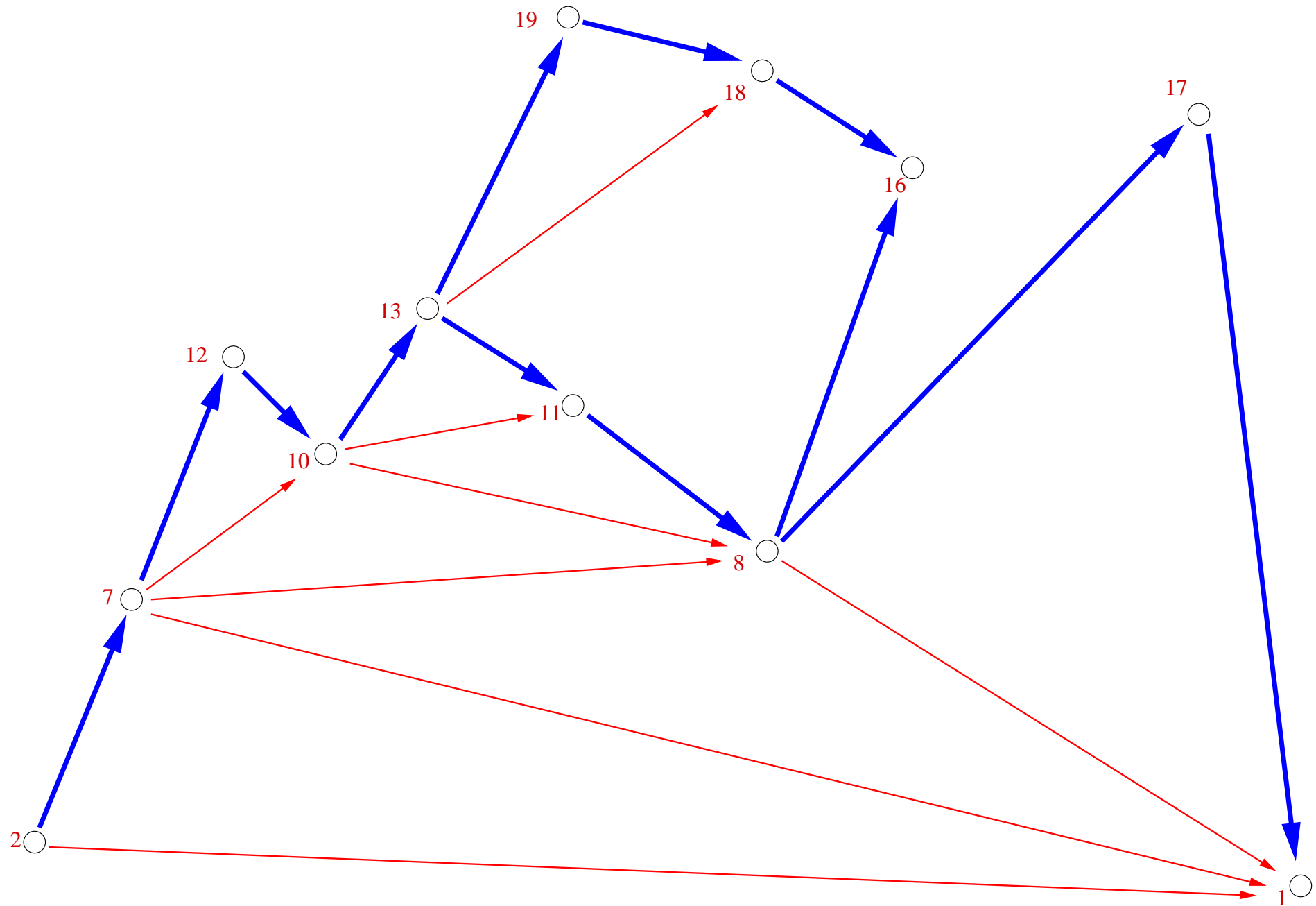
^afor flat terms decidable in $O(|D|^2)$ since $|OC(D)|$ is in $O(|D|^2)$.

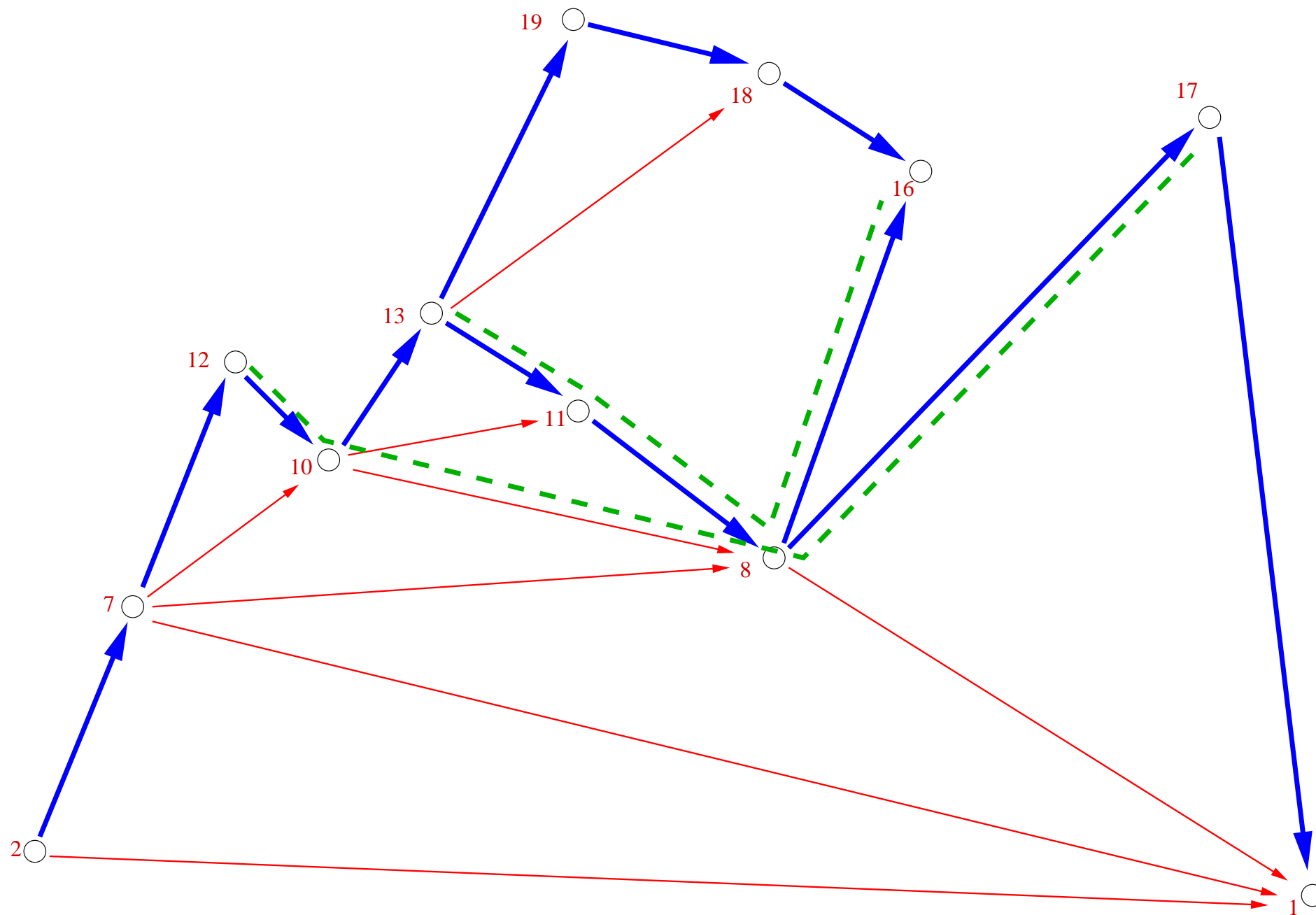
Chaining Diagram (Terms Ordered by Number)

given \Rightarrow -facts



Adding Peak Facts





Deriving equations from inequations is optional. Using them for simplification collapses cycles. **Premises in parenthesis become redundant and can be deleted.**

$$\frac{[x \overset{\vee}{\Rightarrow} y] \quad [y \overset{\vee}{\Rightarrow} x]}{x \doteq y} \quad (\text{whenever you like}) \quad \frac{x \doteq y \quad [A(x)]}{A(y)} \quad (\text{if } x \succ y)$$

Negative inequations in inference rules have to be replaced by rewrite provability, e.g., for set constraints we may add:

$$\frac{f(x) \overset{\vee}{\Rightarrow} f(y)}{x \Rightarrow y}$$

Theoretical Results and Open Questions

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- completeness
- worst-case complexity not better than $O(n^3)$
- for which classes of data bases quadratic?
- how to choose a good ordering?

Encouraging results by Aiken, Fähndrich, Foster, Su (1998, 2000) for Andersen's pointer analysis via atomic set constraints:

- **flat inequations** $\mathcal{X} \Rightarrow \mathcal{Y}$, $\text{ref}(\mathcal{X}) \Rightarrow \mathcal{Y}$, and $\mathcal{X} \Rightarrow \text{ref}(\mathcal{Y})$
- **ref(\mathcal{X}) minimal** in \succ , therefore, $O(1)$ test for injectivity
- if \succ on set variables is **random**, then relatively few variable-variable edges are added
- **partial cycle elimination** according to

$$\frac{x \underset{\succ}{\Rightarrow} \dots \underset{\succ}{\Rightarrow} y \quad y \underset{\prec}{\Rightarrow} x}{x \doteq y}$$

- **analytical model**: $O(1)$ for partial cycle test; ordered chaining adds only 40% of the transitive edges
- transformation to **delay peak computation** that eventually collapse

Very long programs can be analysed in reasonable time

Fundamental problem: efficient deduction for transitive relations in algebraic structures

Logical view: clarifies the issues and provides general efficient methods

Advice to the PL community: adopt that view and obtain almost optimal complexity results and prototype implementations for free

Advice to the ATP community: • make first-order provers work well on these near-propositional cases

- find more meta-complexity theorems for the general case
- implement the algorithms behind the meta-complexity theorems
- analytical models for ordered chaining: when is GMR sub-cubic?