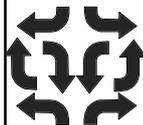


Chapter 10: CLP Systems

Where we examine how CLP systems work and introduce an important concept for constraint solvers: incrementality

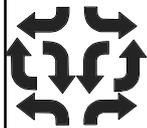
1



CLP Systems

- ▼ Simple Backtracking Goal Evaluation
- ▼ Incremental Constraint Solving
- ▼ Efficient Saving and Restoring of the Constraint Store
- ▼ Implementing If-Then-Else, Once and Negation
- ▼ Optimization
- ▼ Other Incremental Constraint Solvers

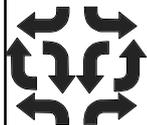
2



Backtracking Goal Evaln.

- ▼ Previously understood as depth-first left-right search through a derivation tree
- ▼ Specific algorithm: `simple_solve_goal`
 - ▼ parametric in `solv` and `simpl`
 - ▼ uses `defn(P,L)` which returns rules defining L in program P in the order they occur, renamed to not contain any previous variables
- ▼ `simple_solve_goal(G)`
 - ▼ **return** `simpl(vars(G), simple_backtrack(<G|true>))`

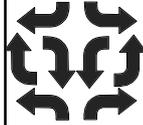
3



simple_backtrack

- ▼ `simple_backtrack(<G|C>)`
 - ▼ **if** G is empty **return** C
 - ▼ **let** G be of the form L, G'
 - ▼ **case** L is a primitive constraint
 - ▼ **if** `solv(C ∧ L) = false` **return** `false`
 - ▼ **return** `simple_backtrack(<G'|C ∧ L>)`
 - ▼ **case** L is an atom $p(s1, \dots, sn)$
 - ▼ **foreach** $p(t1, \dots, tn) :- B$ in `defn(P,L)`
 - ▼ $C1 = \text{simple_backtrack}(\langle s1=t1, \dots, sn=tn, B, G' | C \rangle)$
 - ▼ **if** $C1 \neq \text{false}$ **return** $C1$
 - ▼ **return** `false`

4



Example execution $sum(I, S)$

(S1) $sum(0, 0)$.

(S2) $sum(N, N+S) :- sum(N-1, S)$.

$simple_backtrack(<sum(I, S) | true>)$

$simple_backtrack(<I=0, S=0 | true>)$ rule S1

returns false

$simple_backtrack(<I=N', S=N'+S', sum(N'-I, S') | true>)$ rule S2

$simple_backtrack(<S=N'+S', sum(N'-I, S') | I=N'>)$

$simple_backtrack(<sum(N'-I, S') | I=N' \wedge S=N'+S' >)$

$simple_backtrack(<N'-I=0, S'=0 | I=N' \wedge S=N'+S' >)$ rule S1

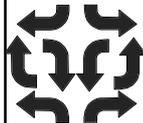
$simple_backtrack(<S'=0 | I=N' \wedge S=N'+S' \wedge N'-I=0 >)$

$simple_backtrack(<[] | I=N' \wedge S=N'+S' \wedge N'-I=0 \wedge S'=0 >)$

returns $I=N' \wedge S=N'+S' \wedge N'-I=0 \wedge S'=0$

5

$simpl(\{S\}, I=N' \wedge S=N'+S' \wedge N'-I=0 \wedge S'=0) = S = I$



Incremental Solving

- ▼ The simple backtracking evaluation is inefficient, consider calls to *solv*

- ▼ $solv(I=0)$

- ▼ $solv(I = N')$

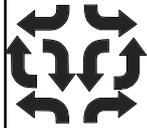
- ▼ $solv(I = N' \wedge S = N + S')$

- ▼ $solv(I = N' \wedge S = N + S' \wedge N'-I = 0)$

- ▼ $solv(I = N' \wedge S = N + S' \wedge N'-I = 0 \wedge S' = 0)$

- ▼ Repeated work

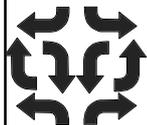
6



Incremental Constraint Solver

- ▼ An **incremental constraint solver** is a function *isolv* which takes a primitive constraint *c* and returns *true*, *false* or *unknown*. There is an implicit *constraint store S*
 - ▼ if $isolv(c) = true$ then $S \wedge c$ is satisfiable
 - ▼ if $isolv(c) = false$ then $S \wedge c$ is unsatisfiable
 - ▼ if $isolv(c) \neq false$ then store is updated to $S \wedge c$

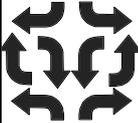
7



Incremental Gauss-Jordan

- ▼ $inc_gj(c)$
 - ▼ $c := eliminate(c, S)$
 - ▼ **if** c is of the form $0 = 0$ **return** *true*
 - ▼ **if** c is of the form $d = 0$ ($d \neq 0$) **return** *false*
 - ▼ rewrite c in the form $x = e$
 - ▼ $S := eliminate(S, x = e) \wedge x = e$
 - ▼ **return** *true*
- ▼ $eliminate(C, x1 = e1 \wedge \dots \wedge xn = en)$
 - ▼ **foreach** xi
 - ▼ replace xi by ei throughout C
 - ▼ **return** C

8

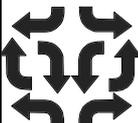


Incremental GJ Example

Solving $I = N' \wedge S = N + S' \wedge N' - I = 0 \wedge S' = 0$

	S	c	eliminate(c, S)
$isolv(N'=I)$	$true$	$1 = N'$	$1 = N'$
$isolv(S=N'+S')$	$N' = 1$	$S = N' + S'$	$S = 1 + S'$
$isolv(N'-I = 0)$	$N' = 1 \wedge S = 1 + S'$	$N' - 1 = 0$	$0 = 0$
$isolv(S' = 0)$	$N' = 1 \wedge S = 1 + S'$	$S' = 0$	$S' = 0$
	$N' = 1 \wedge S = 1 \wedge S' = 0$		

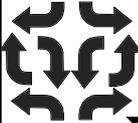
9



Incremental goal solver

- ▼ CLP systems use a global constraint store S and incremental solvers
- ▼ `inc_backtrack` similar to `simple_backtrack`
 - ▼ uses incremental solver
 - ▼ store is not part of argument
 - ▼ functions: `save_store`, `restore_store` for saving and restoring the implicit store

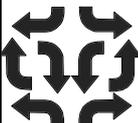
10



inc_backtrack

- ▼ *inc_backtrack*(G)
 - ▼ **if** *G* is empty **return** *true*
 - ▼ **let** *G* be of the form *L, G'*
 - ▼ **case** *L* is a primitive constraint
 - ▼ **if** *isolv*(*L*) = *false* **return** *false*
 - ▼ **return** *inc_backtrack*(*G'*)
 - ▼ **case** *L* is an atom *p*(*s1*,...,*sn*)
 - ▼ **foreach** *p*(*t1*,...,*tn*) :- *B* in *defn*(*P,L*)
 - ▼ *save_store*()
 - ▼ **if** *inc_backtrack*(*s1=t1*,...,*sn=tn,B,G'*) **then**
 - ▼ **return** *CI*
 - ▼ *restore_store*()
 - ▼ **return** *false*

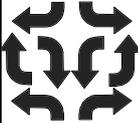
11



inc_solve_goal

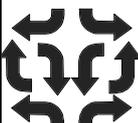
- ▼ The incremental goal solving algorithm, making use of auxiliary functions to initialize and get the constraint store
- ▼ *inc_solve_goal*(*G*)
 - ▼ *W* := *vars*(*G*)
 - ▼ *initialize_store*()
 - ▼ **if** *inc_backtrack*(*G*) **then**
 - ▼ **return** *simpl*(*W, get_store*())
 - ▼ **return** *false*

12



Example execution sum(1,S)

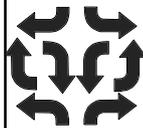
	constraint store stack
inc_backtrack(sum(1,S))	<empty>
inc_backtrack(I=0, S = 0)	true /
return false	<empty>
inc_backtrack(I=N', S = N' + S', sum(N' -1, S'))	true /
inc_backtrack(S = N' + S', sum(N' -1, S'))	true /
inc_backtrack(sum(N'-1, S'))	true /
inc_backtrack(N'-1 = 0, S' = 0)	true / N' = 1 ∧ S = 1 + S'
inc_backtrack(S' = 0)	true / N' = 1 ∧ S = 1 + S'
inc_backtrack([])	true / N' = 1 ∧ S = 1 + S'
<i>simpl</i> ({S}, N' = 1 ∧ S = 1 ∧ S' = 0) = S = 1	13



Efficient saving and restoring

- ▼ Incremental solver requires saving/restoring the constraint store
- ▼ Dont need to save the entire store
- ▼ Save enough information to recreate store
 - ▼ **Trailing**: save modified parts of the constraint store in a trail and recover on backtracking
 - ▼ **Semantic backtracking**: store operations necessary to recover store

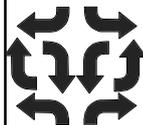
14



Trailing

- ▼ Associate a **timestamp** with each primitive constraint
- ▼ At a choicepoint
 - ▼ store the current timestamp
- ▼ Backtracking
 - ▼ remove all constraints with a later stamp
- ▼ Doesn't handle when an old primitive constraint is modified

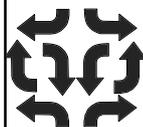
15



Trailing

- ▼ Whenever an old constraint (from before the last choicepoint) is modified
 - ▼ save the old value in the **trail**
- ▼ Note we don't have to trail the same constraint again if it is modified again before another choicepoint

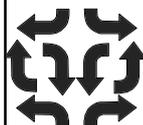
16



Trailing Gauss-Jordan

- ▼ Index each equation by arrival number
- ▼ Choicepoint saves:
 - ▼ index of last equation, *last*
 - ▼ trail of changes (initially empty)
- ▼ Whenever equation i is modified, if $i \leq last$ then each modified coefficient is added to trail $\langle i, x, a \rangle$ or $\langle i, constant, b \rangle$

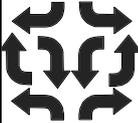
17



Semantic Backtracking

- ▼ Save high-level operations of how to restore the constraint store (domain dependent)
- ▼ For Gauss-Jordan
 - ▼ a new constraint only eliminates a variable x
 - ▼ remember the old coefficients of x and
 - ▼ **undo** the elimination on backtracking

18

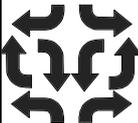


Semantic Backtracking Ex.

Imagine store is	1: $X = Y + 2Z + 4$	Removing constraint
	2: $U = 3Y + Z - 1$	
	3: $V = 3$	
Adding constraint	$Y + 2V + X = 2$	Add coefficient
Eliminate vars	$Y = -Z - 4$	* $(Y+Z+4)$ to eqns 1,2 and remove 4
Eliminate Y using equation and add.	1: $X = Z$	1: $X = Y + 2Z + 4$
Remember coeffs	2: $U = -2Z - 13$	2: $U = 3Y + Z - 1$
	3: $V = 3$	3: $V = 3$
	4: $Y = -Z - 4$	

$[\langle 1, Y, 1 \rangle, \langle 2, Y, 3 \rangle]$

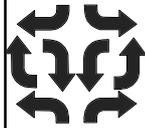
19



Extra Constructs

- ▼ So far “pure” programs (Chapter 4)
- ▼ Chapters 7 and 9 introduce
 - ▼ if-then-else
 - ▼ once
 - ▼ negation
 - ▼ optimization
- ▼ How are they implemented?

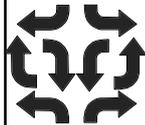
20



If-Then-Else, Once+Negation

- ▾ All three are implemented using a single construct, the **cut**, written !
- ▾ Cut prunes derivations from a tree
 - ▾ when reached: commit to this clause and remove any choices set up within this clause
- ▾ Very powerful, and dangerous
- ▾ Preferable to use if-then-else, once or negation rather than the lower level cut

21



Cut Example

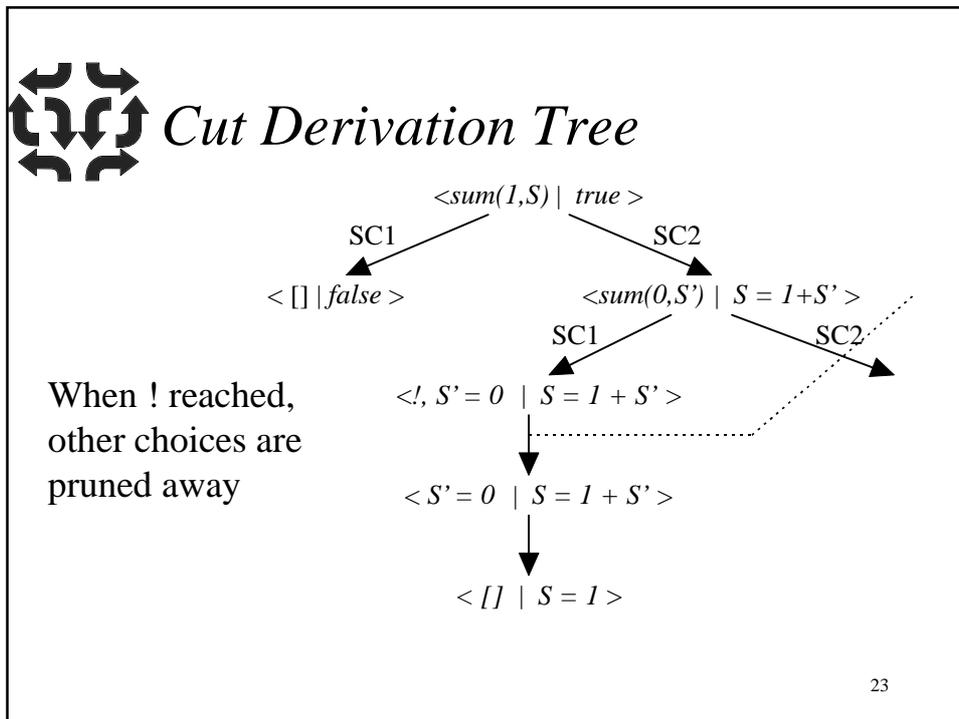
Sum program for mode of usage: first arg fixed

```
sum(N,SS) :-
    (N = 0 ->
        SS = 0
    ;
        N >= 1, SS = N + S,
        sum(N-1,S)
    ).
```

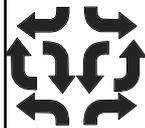
Equivalent version with cut

```
sum(N,SS) :- N = 0, !, SS = 0.
sum(N,SS) :- N >= 1, SS = N + S,
    sum(N-1, S).
```

22



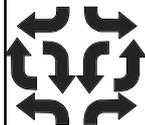
-
- Cut**
- ▼ Cut commits to all choices made since when the atom which was rewritten that introduced the cut
 - ▼ Assume rewriting atom A' using rule
 - ▼ $A \text{ :- } L1, \dots, Li, !, Li+1, \dots, Ln$
 - ▼ When ! reached all choices for rewriting A' and all choices in evaluation $L1, \dots, Li$ are removed
- 24



Implementing Cut

- ▼ Need `save_store` to return an index of the last store
- ▼ `remove_choicepoints(i)` removes all choicepoints with indexes $\geq i$
- ▼ simply modify `inc_backtrack` for case introducing a cut

25



Modifying `inc_backtrack`

- ▼ **case** L is an atom $p(s1, \dots, sn)$
 - ▼ **foreach** $p(t1, \dots, tn) :- L1, \dots, Ln$ in $defn(P, L)$
 - ▼ $i := \text{save_store}()$
 - ▼ **if** some $L_j = !$ **then**
 - ▼ **if** `inc_backtrack(s1=t1, ..., sn=tn, L1, ..., Lj-1)` **then**
 - ▼ `remove_choicepoints(i)`
 - ▼ **return** `inc_backtrack(Lj+1, ..., Ln, G')`
 - ▼ **elseif** `inc_backtrack(s1=t1, ..., sn=tn, G')` **then**
 - ▼ **return** `true`
 - ▼ `restore_store()`
 - ▼ **return** `false`

26

Cut Example 2

```

h(X) :- X > 0, p(X), q(X).
h(4).
p(X) :- X < 4, r(X), !.
p(3).
r(1).
r(2).
q(2).
q(3).
    
```

Diagram illustrating the execution of the program with a cut. The search tree shows the following nodes and transitions:

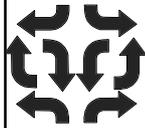
- Root node: $\langle h(X) \mid true \rangle$
- Transition to: $\langle p(X), q(X) \mid X > 0 \rangle$ (left) and $\langle \square \mid X = 4 \rangle$ (right)
- From $\langle p(X), q(X) \mid X > 0 \rangle$:
 - Transition to: $\langle r(X), !, q(X) \mid X > 0 \wedge X < 4 \rangle$ (left)
 - Transition to: $\langle \square \mid X = 4 \rangle$ (right)
- From $\langle r(X), !, q(X) \mid X > 0 \wedge X < 4 \rangle$:
 - Transition to: $\langle !, q(X) \mid X = 1 \rangle$ (left)
 - Transition to: $\langle \square \mid X = 4 \rangle$ (right)
- From $\langle !, q(X) \mid X = 1 \rangle$:
 - Transition to: $\langle q(X) \mid X = 1 \rangle$ (left)
 - Transition to: $\langle \square \mid X = 4 \rangle$ (right)
- From $\langle q(X) \mid X = 1 \rangle$:
 - Transition to: $\langle \square \mid false \rangle$ (left)
 - Transition to: $\langle \square \mid false \rangle$ (right)

27

Cut Example 2

	constraint storestack
inc_backtrack(h(X))	$\langle empty \rangle$
inc_backtrack(X > 0, p(X), q(X))	true /
inc_backtrack(p(X), q(X))	index 2 true / X > 0 /
inc_backtrack(X < 4, r(X)) (before cut)	true / X > 0 /
inc_backtrack(r(X))	true / X > 0 / X > 0 ∧ X < 4 /
return true	remove upto 2 true /
inc_backtrack(q(X)) (after cut)	true / X = 1 /
return false	restore 2 true /
return false	restore 1 $\langle empty \rangle$
inc_backtrack(X = 4)	true /
return true	Answer: X = 4

28

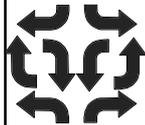


If-Then-Else, Once+Negation

All are implemented using the meta-programming facilities and the cut.

```
once(G) :- call(G), !.  
not(G) :- call(G)!, !, fail.  
not(G).  
G1 -> G2 ; G3 :- call(G1), !, call(G2).  
G1 -> G2 ; G3 :- call(G3).
```

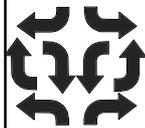
29



Optimization

- ▼ Implementing `minimize(G, E)`
 - ▼ `minimize_store(E)`: returns the minimal value of E wrt to current constraint store
 - ▼ search the derivation tree of G and collect minimum value m of E , then execute $E = m, G$
- ▼ Multiple approaches to search
 - ▼ retry search (restart after finding soln)
 - ▼ backtrack search (continue after finding)

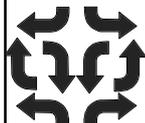
30



Retry Optimization

- ▼ **case** L is minimization literal $minimize(G, E)$
 - ▼ $i := save_store()$
 - ▼ $m := + \infty$
 - ▼ **while** $inc_backtrack(E < m, G)$ **do**
 - ▼ $m := minimize_store(E)$
 - ▼ $remove_choicepoints(i+1)$
 - ▼ $restore_store()$
 - ▼ $i := save_store()$
 - ▼ $restore_store()$
 - ▼ **return** $inc_backtrack(E = m, G, G')$

31



Retry Example

Evaluating $minimize(butterfly(S, P), -P)$

$inc_backtrack(-P < + \infty, butterfly(S, P))$

answer: $-P < + \infty \wedge P = -100 \wedge 0 \leq S \leq 1$

$m := 100$

$inc_backtrack(-P < 100, butterfly(S, P))$

answer: $-P < 100 \wedge P = 100S - 200 \wedge 1 \leq S \leq 3$

$m := -100$

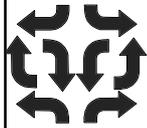
$inc_backtrack(-P < -100, butterfly(S, P))$

returns false

$inc_backtrack(-P = -100, butterfly(S, P))$

answers: $P = 100 \wedge S = 3$ (twice)

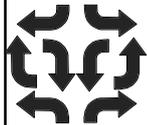
32



Backtracking Optimization

- ▼ `minimize(G, E)`
- ▼ Search the derivation tree for G
- ▼ At each success update the minimal value m of E found (handled by a *catch* literal)
- ▼ Then execute $E=m, G$

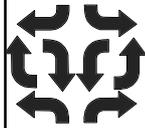
33



Backtracking Optimization

- ▼ **case** L is minimization literal $minimize(G, E)$
 - ▼ $i := save_store()$
 - ▼ $m := + \infty$
 - ▼ `inc_backtrack(G, catch(m, E))`
 - ▼ `restore_store(i)`
 - ▼ **return** `inc_backtrack(E=m, G, G')`
- ▼ **case** L is a catch subgoal $catch(m, E)$
 - ▼ **if** `isolv(E < m) != false` **then**
 - ▼ $m := minimize_store(E)$
 - ▼ **return** `false`

34



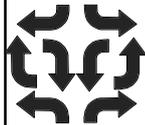
Backtracking Example

```

inc_backtrack(butterfly(S,P),catch(+ , -P))
  inc_backtrack(catch(+ , -P))
    store:  $P = -100 \wedge 0 \leq S \leq 1$  sets  $m := 100$ 
  inc_backtrack(catch(100, -P))
    store:  $P = 100S - 200 \wedge 1 \leq S \leq 3$  sets  $m := -100$ 
  inc_backtrack(catch(-100, -P))
    store:  $P = -100S + 400 \wedge 3 \leq S \leq 5$ 
    isolv( $-P < -100$ ) fails no update
  inc_backtrack(catch(100, -P))
    store:  $P = -100 S \geq 5$  fails no update
inc_backtrack(-P = -100, butterfly(S,P))
  answers:  $P = 100 \wedge S = 3$  (twice)

```

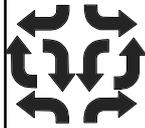
35



Other Incremental Solvers

- ▼ Incremental Tree Solving
 - ▼ Use the store to eliminate variables and solve remainder as before, then use it to eliminate
 - ▼ *inc_tree_solve*(c)
 - ▼ $c := \text{eliminate}(c, S)$
 - ▼ $R := \text{unify}(c)$
 - ▼ **if** $R = \text{false}$ **then return false**
 - ▼ $S := \text{eliminate}(S, R) \wedge R$
 - ▼ **return true**

36



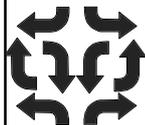
Incremental Tree Solving Ex.

Constraints collected by goal `append([a],[b,c],L)`

$[a] = [F|R], [b,c] = Y, L = [F|Z], R = [], Y = Z$

c	S	$\text{elim}(c)$	$\text{unify}(c)$
$[a] = [F R]$	true	$[a] = [F R]$	$F = a \wedge R = []$
$[b,c] = Y$	$F = a \wedge R = []$	$[b,c] = Y$	$Y = [b,c]$
$L = [F Z]$	$F = a \wedge R = [] \wedge Y = [b,c]$	$L = [a Z]$	$L = [a Z]$
$R = []$	$F = a \wedge R = [] \wedge Y = [b,c] \wedge L = [a Z]$	$[] = []$	true
$Y = Z$	$F = a \wedge R = [] \wedge Y = [b,c] \wedge L = [a Z]$	$[b,c] = Z$	$Z = [b,c]$
	$F = a \wedge R = [] \wedge Y = [b,c] \wedge L = [a,b,c] \wedge Z = [b,c]$		

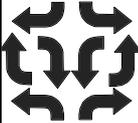
37



Data Structures for Trees

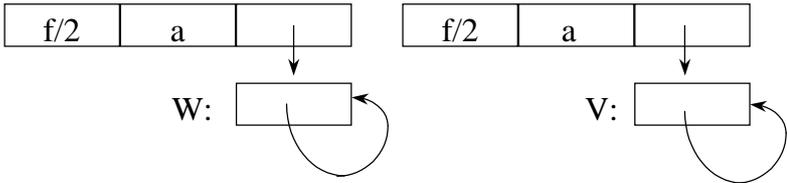
- ▼ Tree constraints are stored/manipulated as dynamic data structures
 - ▼ **variable**: unique memory cell (pointer)
 - ▼ unconstrained: self-pointer
 - ▼ equated to term: pointer at term rep
 - ▼ **term** $f(t1, \dots, tn)$: $n+1$ memory cells
 - ▼ first: constructor info f/n
 - ▼ rest: pointers to $t1, \dots, tn$
 - ▼ optimization: store subtrees with no children directly

38

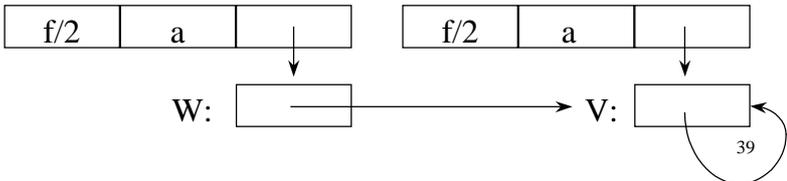


Data Structures for Trees

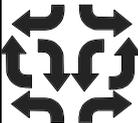
Handling the equation $f(a, W) = f(a, V)$



Match constructor/arities and each arg. Eqn $W = V$ binds W to V

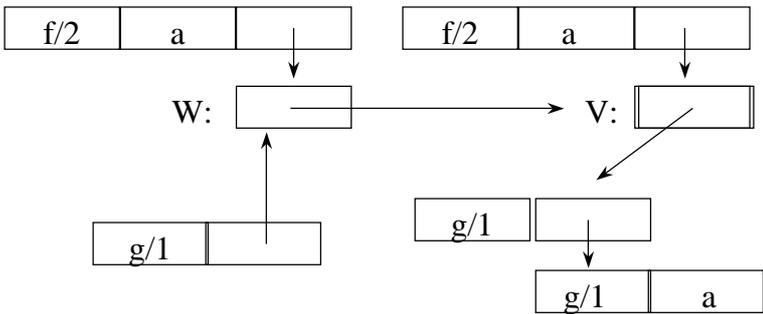


39



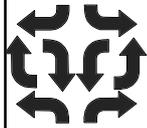
Data Structures for Trees

Incrementally adding $g(W) = g(g(a))$



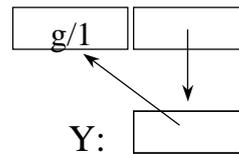
Represents solved form: $V = g(a) \wedge W = g(a)$

40

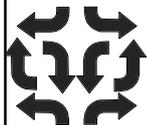


Occurs Check Revisited

- ▼ Most implementations ignore the occurs check!
- ▼ Problems: e.g. $Y = g(Y)$
- ▼ Builds cyclic structures
- ▼ Infinite computation
- ▼ e.g. $Y = g(Y), Z = g(Z), Y = Z$



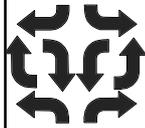
41



Incremental Bounds Cons.

- ▼ Propagation is essentially incremental
- ▼ incremental bounds consistency:
 - ▼ Add new prim. constraint to store and queue
 - ▼ Pick prim. constraint from queue
 - ▼ Enforce its bounds consistency
 - ▼ Add prim. constraint with modified variables to queue
 - ▼ Repeat until queue is empty, or empty domain

42



Incremental Bounds Ex.

Smugglers knapsack, no whiskey

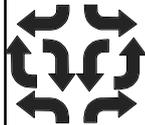
$$\boxed{4W + 3P + 2C \leq 9} \wedge \boxed{15W + 10P + 7C \geq 30}$$

$$D(W) = [0..0], D(P) = [1..3], D(C) = [0..3]$$

Add first constraint

Add second constraint

43



CLP Systems Summary

- ▼ Incremental constraint solving
 - ▼ essential for efficiency
- ▼ Global constraint store
 - ▼ require efficient save and restore
- ▼ The Cut!
 - ▼ implements if-then-else, once + negation
- ▼ Minimization
 - ▼ many possible implementations

44