

Recursion Elimination in Dynamic Programming

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Introduction: levels of Recursion Elimination – interpreted and uninterpreted

Part 1

Recursive factorial

- Recursive program to compute the factorial function $F: \mathbf{N} \rightarrow \mathbf{N}$
 - $F(n) = \text{if } n = 0 \text{ then } 1 \text{ esle } n \cdot F(n - 1)$ (in the standard notation),
 - $F(n) = \text{if } p(n) \text{ then } c \text{ else } f(n, F(g(n)))$ (in a prefix notation),

where “known” functions are

- $p \equiv (\lambda x \in \mathbf{N}. (x = 0)) : \mathbf{N} \rightarrow \text{Boolean},$

- $c \equiv 1 : \rightarrow \mathbf{N}$ (i.e. a constant)

- $f \equiv (\lambda x, y \in \mathbf{N}. (x \cdot y)) : \mathbf{N} \times \mathbf{N} \rightarrow \mathbf{N},$

- $g \equiv (\lambda x \in \mathbf{N}. (\text{if } x = 0 \text{ then } 0 \text{ else } (x - 1))) : \mathbf{N} \rightarrow \mathbf{N}.$

Imperative factorial

Program 1

```
1. VAR  $x, y: \mathbf{N}$ ;  
2.  $y := 1$ ;  
3. while  $x \neq 0$  do  
4.    $y := x \cdot y$ ;  
5.    $x := x - 1$   
6. od
```

Program 2

```
1. VAR  $x, y, z: \mathbf{N}$ ;  
2.  $y := 1; z := 1$ ;  
3. while  $z \leq x$  do  
4.    $y := z \cdot y$ ;  
5.    $z := z + 1$   
6. od
```

What if *known* functions are *uninterpreted*?

Recursive schemata with a single available (not specified) data type T :

$$F(x) = \text{if } p(x) \text{ then } c \text{ else } f(x, F(g(x)))$$

Standard scheme 1

1. $VAR\ x, y: T;$
2. $y := c;$
3. $while\ \neg p(x)\ do$
4. $y := f(x, y);$
5. $x := g(x)$
6. od

Standard scheme 2

1. $VAR\ x, y, z: T;$
2. $y := c; z := c;$
3. $while\ q(x, z)\ do$
4. $y := f(z, y);$
5. $z := h(z)$
6. od

Herbrand models and structures

- To demonstrate that no two of program schemata from the previous slide are equivalent, it is sufficient to consider *Herbrand models* (also called *free models*).
- The domain of a Herbrand model comprises all terms constructed from the available functional symbols and input variables (while the domain of the Herbrand structures comprise the ground terms exclusively).

Why the schemata aren't equivalent?

- Let us consider a Herbrand model such that
 - q is always *TRUE*,
 - $p(g(g(x)))$ is *TRUE* while p is *FALSE* for all other terms.
- Then
 - $F(x) = f(x, F(g(x))) = f(x, f(g(x), F(g(g(x)))) = f(x, f(g(x), c))$,
 - the output value of y computed by scheme 1 is $f(g(x), f(x, c))$,
 - while scheme 2 does not halt at all.

Translation of the recursive scheme to a standard scheme (with equality)

1. $\forall AR\ x, y, u, v : T;$
2. $u := x;$
3. *while* $\neg p(u)$ *do*
4. $u := g(u)$
5. *od*
6. $y := c;$
7. *while* $u \neq x$ *do*
8. $v := x;$
9. *while* $g(v) \neq u$ *do*
 $Inv. 1: \exists m < n \in \mathbf{N} : v = g^m(x) \ \& \ u = g^n(x)$
 $v := g(v)$
10. $y := f(u, y); u := v$
11. *od*;
12. $y := \text{if } p(x) \text{ then } c \text{ else } f(x, y)$

How to rid of the equality

- Finally, the equality used in lines 7 and 9 of the scheme is easy to eliminate because it may be implemented as call of the following *tail-recursive* function *EQ* (easy to implement by an iterative program:

```
1  VAR x, y, u, v : D;  
2  u := x;  
3  while ¬p(u) do  
4    u := g(u)  
5  od  
6  y := c;  
7  while u ≠ x do  
8    v := x;  
9    while g(v) ≠ u do  
      //Invariant 1: ∃m < n ∈ ℕ : v = gm(x) & u = gn(x)  
      v := g(v)  
    od;  
    //Invariant 2: g(v) = u & y = F(u)  
10   y := f(u, y); u := v  
11 od
```

$$EQ(a, b) = \text{if } p(a) \vee p(b) \text{ then } p(a) \ \& \ p(b) \text{ else } EQ(g(a), g(b)).$$

Translation of the recursive factorial to an iterative form

1. $V AR x, y, u, v : N;$
2. $u := x;$
3. $while u \neq 0 do$
4. $u := u - 1$
5. od
6. $y := 1;$
7. $while u \neq x do$
8. $v := x;$
9. $while (v - 1) \neq u do$
 $Inv. 1: \exists m < n \in N : v = x - m \ \& \ u = x - n$
 $v := v - 1$
- $od;$
- $Inv. 2: (v - 1) = u \ \& \ y = F(u)$
10. $y := u \cdot y; u := v$
11. $od;$
12. $y := (x = 0) \text{ then } 1 \text{ else } (x \cdot y)$

Extremely inefficient but semantic-independent

- Unfortunately, imperative factorial from the previous slide 10 is extremely inefficient – it runs in $O(n^2)$ time in contrast to both programs (1 and 2) from slide 4 that run in linear time $O(n)$.
- It worth to remark that Program 1 can be automatically constructed from the recursive factorial program using *co-recursion* and *tail-recursion*.
- This use of the co-recursion is semantic-dependent (since it is safe assuming commutativity of the function f), while our approach to recursion elimination is semantic-independent.

Co-recursion and Tail-recursion by example

- Recursive factorial $F(n) = \text{if } n = 0 \text{ then } 1 \text{ esle } n \cdot F(n - 1)$ is not in the tail-form (because has next call inside some function).
- But it is equivalent to the following recursive program in the tail-form:

$$\begin{cases} F(n) = P(n, 1) \\ P(n, m) = \text{if } n = 0 \text{ then } m \text{ esle } P((n - 1), (n \cdot m)) \end{cases}$$

- This program is in the tail-form because all calls are never inside other functions.
- Co-recursion is a “trick” that consists in converts result into another argument and use this argument in the recursion.

Teil-recursion elimination by example

- Tail-recursion $\begin{cases} F(n) = P(n, 1) \\ P(n, m) = \text{if } n = 0 \text{ then } m \text{ esle } P((n - 1), (n \cdot m)) \end{cases}$
is easy to eliminate (and compare with Program 1 from slide 4):

<i>start: VAR x, y: N goto 2</i>	<i>1. VAR x, y: N;</i>
<i>2: y := 1 goto 3</i>	<i>2. y := 1;</i>
<i>3: if x = 0 then goto <u>stop</u> else goto 4</i>	<i>3. while x ≠ 0 do</i>
<i>4: y := x · y goto 5</i>	<i>4. y := x · y;</i>
<i>5: x := x - 1 goto 3</i>	<i>5. x := x - 1</i>
<i><u>stop</u></i>	<i>6. od</i>

Recursive and iterative Dynamic Programming

Part 2

Warming-up Dropping Bricks Problem

- Define stability of “bricks” (cell phones) by dropping them from a tower of H meters. How many times do you need to drop bricks, if you have just 2 bricks?
- $G(n) = \text{if } n = 0 \text{ then } 0 \text{ else}$
 $1 + \min_{1 \leq k \leq n} \max\{(k - 1), G(n - k)\}.$



History of “Dynamic Programming”

- Dynamic Programming was introduced by Richard Bellman in the 1950s to tackle optimal planning problems.
- In 1950s the noun *programming* had nothing in common with more recent *computer programming* and meant *planning* (compare: *linear programming*).
- The adjective *dynamic* points out that Dynamic Programming is related to a *change of states* (compare: *dynamic logic, dynamic system*).

Bellman equation and optimality principle

- *Bellman equation* is a functional equality for the objective function that expresses the optimal solution at the *current* state in terms of the optimal solution at *next* (changed) states.
- It is conceptualized a so-called *Bellman Principle of Optimality*: an optimal plan (or program) should be optimal at every stage.

Descending (top-down) Dynamic Programming

- General pattern of Bellman equation may be formalised by the following *scheme of recursive descending Dynamic Programming*:

$G(x) = \text{if } p(x) \text{ then } f(x) \text{ else}$

$$g \left(x, \underbrace{\left\{ h_i \left(x, G(t_i(x)) \right) : i \in [1..n(x)] \right\}} \right);$$

the term is *linear in each branch*
w.r.t. the objective function G

Descending (top-down) Dynamic Programming – cont.

- In this scheme
 - $G: X \rightarrow Y$ is a symbol for the objective function,
 - $p: X \rightarrow Bool$ is a symbol for a known predicate,
 - $f: X \rightarrow Y$ is a symbol for a known function,
 - is a symbol for a known function with a variable (but finite) number of arguments,
 - all $h_i: X \times Z \rightarrow Y, i \in [1..n(x)]$ are symbols for known functions,
 - all $h_i: X \rightarrow X, i \in [1..n(x)]$ are symbols for known functions too.

More Examples: Factorial, Fibonacci Numbers and Words

- $F(n) = \text{if } n = 0 \text{ then } 1 \text{ else } n \cdot F(n - 1);$
- $Fib(n) = \text{if } 0 \leq n \leq 1 \text{ then } 1 \text{ else } Fib(n - 2) + Fib(n - 1);$
- $Wrd(n) = \text{if } n = 0 \text{ then } a$
 $\qquad \qquad \qquad \text{else if } n = 1 \text{ then } b$
 $\qquad \qquad \qquad \text{else } Wrd(n - 2) \circ Wrd(n - 1).$

Observations

- Factorial, Fibonacci Numbers and Words need static memory of a fixed size.
- Surprisingly, but Dropping Bricks Problem also needs just static memory of fix-size, since $G(n) = \arg \min k \in \mathbf{N}: \left(\frac{k(k+1)}{2} \geq n \right)$.

Problem under study

- It follows from Paterson M.S. and Hewitt C.T. paper *Comparative Schematology* (1970) that fix-size *static memory* is *not enough* for recursion elimination in Bellman equation.
- When one-time allocated
 - array (with integer indexes),
 - (fix-size) static memoryis sufficient to eliminate recursion in Bellman equation?

A Need of Dynamic Memory

- The following program scheme

$$F(x) = \text{if } p(x) \text{ then } x \text{ else } f\left(F(g(x)), F(h(x))\right)$$

is not equivalent to any standard program scheme:

for every $n > 0$

there exists an Herbrand model T_n

where any standard program scheme
needs n variables to compute F .

Support of the Objective Function

- If $G(x) = \text{if } p(x) \text{ then } f(x) \text{ else}$

$$g\left(x, \left\{h_i\left(x, G(t_i(x))\right) : i \in [1..n(x)]\right\}\right)$$

is defined for some value v , then it is possible to pre-compute the *support* $\text{spp}(v)$, the set of all values that occur in the computation of $G(v)$:

$$\text{spp}(x) = \text{if } p(x) \text{ then } \{x\} \text{ else } \{x\} \cup \left(\bigcup_{i \in [1..n(x)]} \text{spp}(t_i(x))\right).$$

- Remark, that for every v , if $G(v)$ is defined, then $\text{spp}(v)$ is finite (but not vice versa).

When an array suffices

- One-time allocated array with integer indexes suffices for computing

$G(x) = \text{if } p(x) \text{ then } f(x) \text{ else}$

$$g\left(x, \left\{h_i\left(x, G(t_i(x))\right) : i \in (1..n(x))\right\}\right)$$

if n is a constant and all $t_i, i \in (1..n(x))$, are interpreted by commutative functions.

When static memory suffices

- Fix-size static memory suffice for computing

$G(x) = \text{if } p(x) \text{ then } f(x) \text{ else}$

$$g\left(x, \left\{h_i\left(x, G(t_i(x))\right) : i \in (1..n(x))\right\}\right)$$

if $n(x) = n$ is a constant and there exists a known computable function t such that

- $t_i = t^i$ for all $i \in [1..n]$,
 - $p(u)$ implies $p(t(u))$ for all $u \in \text{spp}(x)$.
- Examples: Factorial, Fibonacci Numbers and Words.
 - Counter-example: Paterson-Hewitt scheme.

Design outlines and proof comments

Proof comments

- Proof idea – very same as for factorial function in Part 1.
- Scheme' design (with equality and invertible function t) is depicted to the right.

Design outlines

```
1  VAR  $x, x_1, \dots, x_n : X$ ;  
2  VAR  $y, y_1, \dots, y_n : Y$ ;  
3   $x := v$ ;  
4  if  $p(x)$  then  $y := f(x)$   
5     else { do  $x := t_1(x)$  until  $p(x)$ ;  
6              $x_1 := x; y_1 := f(x_1)$ ;  
               $x_2 := t(x_1); y_2 := f(x_2)$ ;  
              ... ..  
               $x_n := t(x_{n-1}); y_n := f(x_n)$ ;  
7             do  
8                  $x := t^{-}(x)$ ;  
//Invariant:  $x = t^{-}(x_1) \ \& \ \text{bas}(x) = \{x_1, \dots, x_n\} \ \&$   
//Invariant:  $\& y_1 = G(x_1) \ \& \dots \ \& y_n = G(x_n)$   
9                  $y := g(x, (h_1(x, y_1), \dots, h_n(x, y_n)))$ ;  
10                  $y_n := y_{n-1}; \dots y_3 := y_2; y_2 := y_1$ ;  
11                  $y_1 := y$ ;  
12                  $x_1 := t^{-}(x_1); \dots x_n := t^{-}(x_n)$   
13             until  $x = v$  }.
```

References, concluding remarks, and topics for further research

Part 3

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Concluding remarks

- A novelty of our study consists in use of templates (understood as semi-interpreted program schemata with symbol of a variable arity) and semantic sufficient conditions that allow recursive programs to be computed efficiently by iterative imperative programs (with either an associative or integer arrays or just with a finite fixed size static memory).

Further research topics

- All our sufficient conditions impose some constraints on interpretation of functional and predicate symbols. A very natural question is whether we can weaken these sufficient conditions?
- Computer-aided verification of the correctness of the translation of the descending dynamic programming template into iterative templates with arrays or fix-size static memory is a topic for further research.