

Automated Approximate Recurrence Solving applied to Static Analysis of Energy Consumption

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Introduction

Team



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Why Energy?

IT's share of global carbon emissions has grown from 2.5% to around 5% in the last ten years.

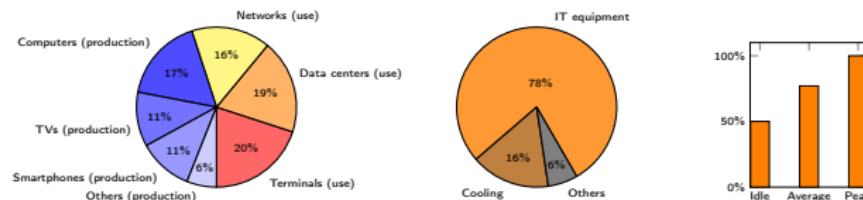
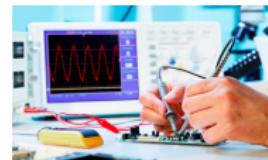


Figure: Energy Usage in IT. By subfield, and case studies on a data center.

- Energy consumption of programs is relevant and understudied by carbon audit experts.
- Also, potential applications to verification of embedded software or against side-channel attacks.



- Applicable to other resources than energy: time, memory, number of communications, ...

Why Recurrence Equations?

- There are both software and hardware components to such line of work. In this internship, we focused more on the problem of control flow analysis than on energy models.
- The hard software problem is recursivity.
Use Horn Clauses as Intermediate Representation.

"Systems of recurrence equations may be seen as programs stripped from information irrelevant to cost analysis"
- Here, we don't really care about exact solutions of equations: bounds on the solutions are satisfactory.
This is a research opportunity.

Contents of this talk

- ① Background (Ciao/CiaoPP, Logic Programming, Abstract Interpretation)
- ② Current pipeline for energy analysis of imperative programs
- ③ Implementation of classical recurrence solving techniques
- ④ Proposal of new order-theoretical recurrence solving techniques

Background

Ciao / CiaoPP

An extensible logic programming language making full use of analysis/verification/optimisation.

```

1 :- module(_, [app/3], [assertions]).  

2 :- assertions.  

3 :- entry app(A,B,C) : (list(A), list(B)).  

4 :- pred app(A,B,C) : (list(A), list(B)) => list(C).  

5 app([], Y,Y).  

6 app([X|Xs], Ys, [X|Zs]) :-  

    app(Xs,Ys,Zs).  

0 - 195 app_assert.pl Ciao ①@e@ unix | 8: 0 All  

(assertions checked in 2.498 msec.)  

yes  

ciaopp ?- output.  

(written file /Users/sabel.garcia/git/ciao-devel/ciaopp/doc/tutorials/fut_examples/app_assert_eterms_shfr_co.pl)  

yes  

ciaopp ?-  

0 * 7.7k *Ciao Preprocessor* Ciao Listener utf-8 | 181:10 Bottom  

1 :- module(_, [app/3], [assertions,regtypes,nativeprops]).  

2 :- assertions.  

3 :- 88 88 :- check pred app(A,B,C)  

4 :- 88 88 : ( list(A), list(B) )  

5 :- 88 88 => list(C).  

6 :-  

7 :- entry app(A,B,C)  

     : ( list(A), list(B) ).  

8 :-  

9 :-  

10 :- checked calls app(A,B,C)  

11   : ( list(A), list(B) ).  

12 :-  

13 :- true success app(A,B,C)  

14   : ( list(A), list(B) )  

15   => list(C).  

16 :-  

17 :- true pred app(A,B,C)  

18   : ( list(A), list(B), term(C) )  

19   => ( list(A), list(B), list(C) ).  

20 :-  

0 - 635 app_assert_eterms_shfr_co.pl Ciao unix |

```

Preprocessor Option Browser

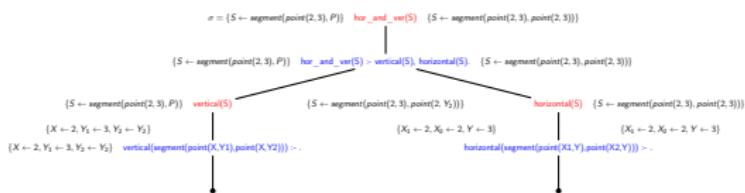
Option	Value
none	✓
naive	✓
check_assertions	✓
manual	✓
curr_mod	✓
on	✓
nf	✓
none	✓
none	✓
det	✓
entry	✓
off	✓
off	✓
warning	✓
off	✓
on	✓
on	✓
source	✓
off	✓
off	✓
on	✓

File Line Col Level ID Message (Checker)

revf.pl	1	error		Errors detected. Further preprocessing...
revf.pl	5	info		Verified assertion: :- check calls nrev(A,B) : list(A). (ciaopp-cost)
revf.pl	5	info		Verified assertion: :- check success nrev(A,B) : list(A) => list(B). (ciaopp-cost)
revf.pl	5	error		False assertion: :- check comp nrev(A,B) : list(A) + (not..fails, is..det, steps..oLength(A)). because the comp field is incompatible with [generic..comp] covered..is..det..mut..exec
revf.pl	12	info		Verified assertion: :- check calls conc(A,B,C). (ciaopp-cost)
revf.pl	12	info		Verified assertion: :- check comp conc(A,B,C) + (terminates, is..det, steps..oLength(A)).

Logic Programming

- Fact, rules and queries.
- Programming with Relations: nondeterminism and no fixed input/output status.
- Execution \leftrightarrow automated proof search. Computation happens by unification.
- Using Logic Programs as (Horn Clause) Intermediate Representation of imperative programs,
the only control structure is function call.
- AND-OR** trees, sets of substitutions, ...



```

vertical  (segment(point(X,_),point(X,_))).
horizontal(segment(point(_,Y),point(_,_))).

hor_and_ver(S) :- vertical(S), horizontal(S).

hor_or_ver(S) :- vertical(S).
hor_or_ver(S) :- horizontal(S).

?- hor_and_ver(segment(point(2,3), P)).
P = point(2, _1658)
P = point(_1364, 3)
  
```

Abstract Interpretation

- A general theory of sound abstractions of program semantics, based on order theory and Galois connections between lattices called *concrete* and *abstract domains*.



Figure: Abstract Interpretation was first developed by Patrick and Radhia Cousot in the 70's.

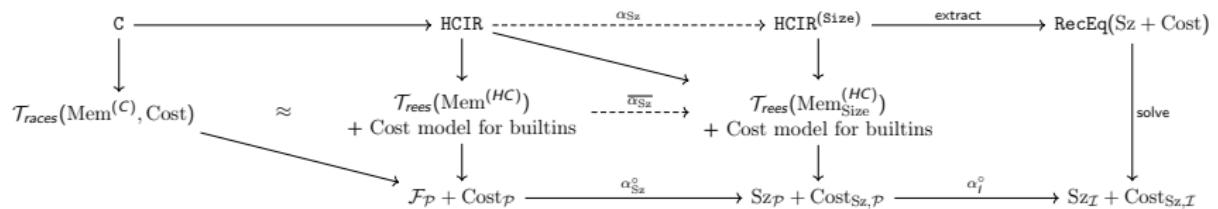
- Observation: program semantics can be viewed as the *least fixed point* of some monotone operator.



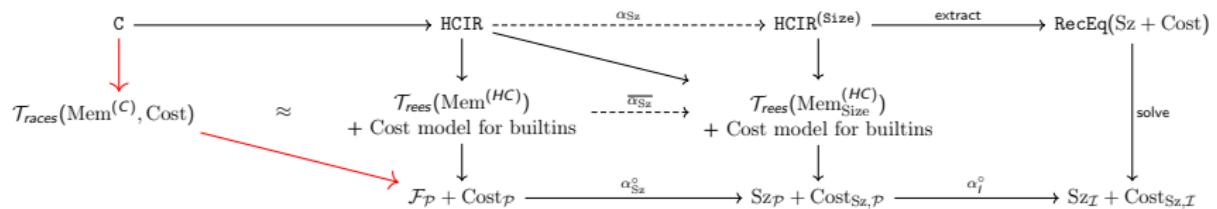
Figure: "Execute then Abstract" or "Abstract then Execute", or even "Execute Abstractly".

Pipeline

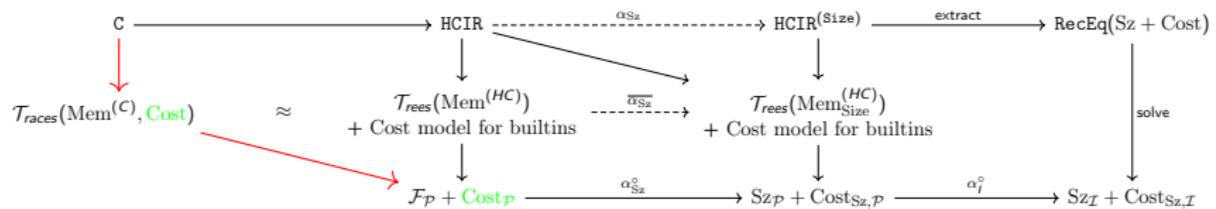
Pipeline



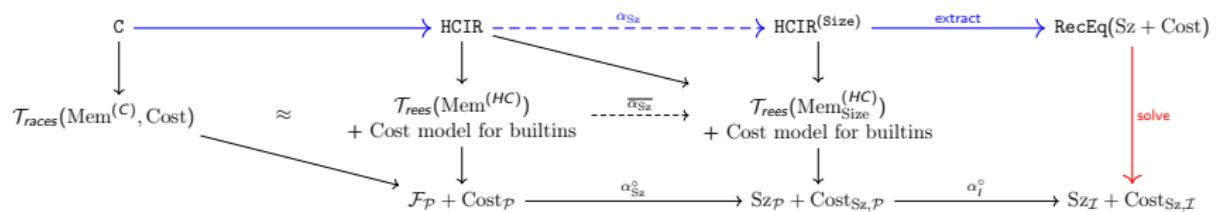
Pipeline



Pipeline



Pipeline



Code transformation (1)

```
int fact(int n){  
    if(n <= 0)  
        return 1;  
    return n*fact(n-1);  
}
```

```
<fact>:  
001: entsp 0x2  
002: stw r0, sp[0x1]  
003: ldw r1, sp[0x1]  
004: ldc r0, 0x0  
005: lss r0, r1  
006: bf <008>  
  
007: bu <010>  
010: ldw r0, sp[0x1]  
011: sub r0, r0, 0x1  
012: bl <fact>  
  
013: ldw r1, sp[0x1]  
014: mul r0, r1, r0  
015: retsp 0x2  
  
008: mkmsk r0, 0x1  
009: retsp 0x2
```

```
fact(R0, R0_3) :-  
    entsp(0x2),  
    stw(R0, Sp0x1),  
    ldw(R1, Sp0x1),  
    ldc(R0_1, 0x0),  
    lss(R0_2, R0_1, R1),  
    bf(R0_2, 0x8),  
    fact_aux(R0_2, Sp0x1, R0_3, R1_1).  
  
fact_aux(1, Sp0x1, R0_4, R1) :-  
    bu(0x0A),  
    ldw(R0_1, Sp0x1),  
    sub(R0_2, R0_1, 0x1),  
    bl(fact),  
    fact(R0_2, R0_3),  
    ldw(R1, Sp0x1),  
    mul(R0_4, R1, R0_3),  
    retsp(0x2).  
  
fact_aux(0, Sp0x1, R0, R1) :-  
    mkmsk(R0, 0x1),
```

Figure: C program (left) translated into ISA level (middle) and HCIR from ISA (right).

Code transformation (2)

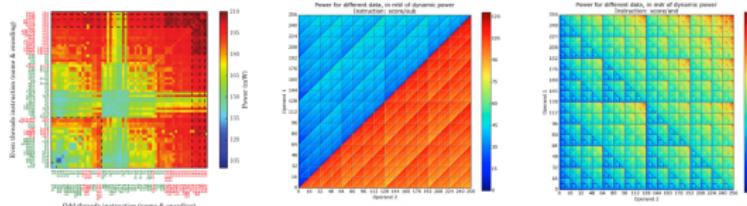
```
int fact(int n){  
    if(n <= 0)  
        return 1;  
    return n*fact(n-1);  
}
```

$$\Rightarrow \begin{cases} Sz_{\text{fact}}(n) = 1 & \text{when } n \leq 0, \\ Sz_{\text{fact}}(n) = n \times Sz_{\text{fact}}(n - 1) & \text{otherwise.} \end{cases}$$

- The recurrence structure of programs appears in the corresponding equations.
- This is a simple case. In general, size measure may introduce abstractions, and we use both size and cost functions.
- When translating imperative loops into recursive programs, ranking functions may have to be inferred.

The energy model problem

- Need to create models by measurements or simulation.
- A simple model might assign a *constant consumption* to each type of instruction, or a reasonably tight *interval*.
- More complex models might care about *history* of previous instructions, e.g. *pairs of instructions*, or about value of operands (*data-dependent consumption*).



- Can be important to deal with “hardware’s runtime policies”, e.g. *cache behaviour*. Static analysis techniques exist, but it is hard.
- Choosing the right level of granularity is hard.
May have to do compromises depending on the application.

Implementation

- Adding some classical techniques to the recurrence solver
 - “Dictionary lookup”
 - Rewriting

Adding classical recurrence solving techniques

- Second and third order linear recurrence equations with constant coefficients and a few options for the affine term. Classical method with particular solution + homogeneous solution using roots of characteristic polynomial.
- Required to add complex numbers to CiaoPP (and to its numerical expressions).

```

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Visit New File Open Directory Close Save Save As Undo ⌘C Cut ⌘V Copy ⌘P Paste ⌘F String Forward
p([X,Y], [X,Y]).  
p([_,Y|Z|R], T) :-  
    p([Y,Z|R], L1), p([Z|R], L2), p(R, L3),  
    append(L1, L2, L4), append(L4, L3, T).  
  
append([], X, X).  
append([H|X], Y, [H|Z]) :- append(X, Y, Z).

```

```

U:-*- third_ord.pl Bot L11 (Ciao)
:- true pred p(_A,T)
: ( list(gnd,_A), var(T) )
=> ( list(gnd,_A), list(gnd,T),
      size(ib,length,T,0) ).  
  
:- true pred p(_A,T)
: ( list(gnd,_A), var(T) )
=> ( list(gnd,_A), list(gnd,T),
      size(ub,length,T,inf) ).  
  
:-**- third_ord_eterms shfr_nf_resources.co.pl 5% L16 (Ciao)

```

Before

```

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Visit New File Open Directory Close Save Save As Undo ⌘C Cut ⌘V Copy ⌘P Paste ⌘F String Forward
p([X,Y], [X,Y]).  
p([_,Y|Z|R], T) :-  
    p([Y,Z|R], L1), p([Z|R], L2), p(R, L3),  
    append(L1, L2, L4), append(L4, L3, T).  
  
append([], X, X).  
append([H|X], Y, [H|Z]) :- append(X, Y, Z).

```

```

U:-*- third_ord.pl Bot L11 (Ciao)
:- true pred p(_A,T)
: ( list(gnd,_A), var(T) )
=> ( list(gnd,_A), list(gnd,T),
      size(ib,length,T,complex(-0.25951582498161835,-0.142223987459299)) ).  
+ccomplex(-0.4196433776070884,-0.6062907292071993)*length(_A)+ccomplex(-0.25951582498161824,0.1422239874592993)*ccomplex(-0.41964337760708865,0.6862987292071993)*length(_A)+complex(0.5190316499632366,-1.7214688391283412e-17)*1.8392867552141612**length(_A)).  
+  
:- true pred p(_A,T)
: ( list(gnd,_A), var(T) )
=> ( list(gnd,_A), list(gnd,T),
      size(ub,length,T,complex(-0.25951582498161835,-0.142223987459299)) ).  
+ccomplex(-0.4196433776070884,-0.6062907292071993)*length(_A)+ccomplex(-0.25951582498161824,0.1422239874592993)*ccomplex(-0.41964337760708865,0.6862987292071993)*length(_A)+complex(0.5190316499632366,-1.7214688391283412e-17)*1.8392867552141612**length(_A)).  
:-**- third_ord_eterms shfr_nf_resources.co.pl 4% L16 (Ciao)

```

After

Irrelevant variables analysis – An analyse + rewrite pass

Example from cost analysis of a fact program with accumulator.

$$\begin{cases} f(n, a) = 1 + f(n - 1, (n - 1) \times a) & \text{if } n > 0, \\ f(n, a) = 0 & \text{if } n \leq 0. \end{cases}$$

Irrelevant variables analysis – An analyse + rewrite pass

Example from cost analysis of a fact program with accumulator.

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$$\begin{cases} \tilde{f}(n) = 1 + \tilde{f}(n - 1) & \text{if } n > 0, \\ \tilde{f}(n) = 0 & \text{if } n \leq 0. \end{cases}$$

Irrelevant variables analysis – Fixpoint formulation

$$\begin{cases} f(n, a) = 1 + f(n - 1, (n - 1) \times a) & \text{if } n > 0, \\ f(n, a) = 0 & \text{if } n \leq 0. \end{cases}$$

- In general, we do this by overapproximating the set of *relevant indices*, as the lfp of the following operator.

$$F : \mathcal{P}(\llbracket 1, k \rrbracket) \rightarrow \mathcal{P}(\llbracket 1, k \rrbracket)$$

$$I \mapsto I \cup \{i \mid n_i \text{ appears in a condition } \phi_j\}$$

$$\cup \left\{ i \mid \begin{array}{l} n_i \text{ appears in an expression } \Psi_j \\ \text{via a path going only through } f \text{ via indices } i' \in I \end{array} \right\}$$

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- In this example, only n is added, using the boundary conditions:

$$\text{lfp } F = \{1\}.$$

Irrelevant variables analysis – Fixpoint formulation (2)

$$\begin{cases} f(n_1, n_2, n_3, n_4, n_5) = 0 & \text{if } n_1 \leq 0 \\ f(n_1, n_2, n_3, n_4, n_5) = 1 + f(n_1 - n_3 - 1, n_1 \times n_2, n_2 \times n_3, n_3 \times n_4, n_4 \times n_5) & \text{if } n_1 > 0. \end{cases}$$

- Again, we overapproximate the set of *relevant indices*, as the lfp of F .

$$F : \mathcal{P}(\llbracket 1, k \rrbracket) \rightarrow \mathcal{P}(\llbracket 1, k \rrbracket)$$

$$I \mapsto I \cup \{i \mid n_i \text{ appears in a condition } \phi_j\}$$

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- First, n_1 is relevant because of the boundary conditions.

Irrelevant variables analysis – Fixpoint formulation (2)

$$\begin{cases} f(n_1, n_2, n_3, n_4, n_5) = 0 & \text{if } n_1 \leq 0 \\ f(n_1, n_2, n_3, n_4, n_5) = 1 + f(n_1 - n_3 - 1, n_1 \times n_2, n_2 \times n_3, n_3 \times n_4, n_4 \times n_5) & \text{if } n_1 > 0. \end{cases}$$

- Again, we overapproximate the set of *relevant indices*, as the lfp of F .

$$F : \mathcal{P}(\llbracket 1, k \rrbracket) \rightarrow \mathcal{P}(\llbracket 1, k \rrbracket)$$

$$I \mapsto I \cup \{i \mid n_i \text{ appears in a condition } \phi_j\}$$

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- First, n_1 is relevant because of the boundary conditions.
- To compute n_1 , we need n_3 ,

.

Irrelevant variables analysis – Fixpoint formulation (2)

$$\begin{cases} f(n_1, n_2, n_3, n_4, n_5) = 0 & \text{if } n_1 \leq 0 \\ f(n_1, n_2, n_3, n_4, n_5) = 1 + f(n_1 - n_3 - 1, n_1 \times n_2, n_2 \times n_3, n_3 \times n_4, n_4 \times n_5) & \text{if } n_1 > 0. \end{cases}$$

- Again, we overapproximate the set of *relevant indices*, as the lfp of F .

$$F : \mathcal{P}(\llbracket 1, k \rrbracket) \rightarrow \mathcal{P}(\llbracket 1, k \rrbracket)$$

$$I \mapsto I \cup \{i \mid n_i \text{ appears in a condition } \phi_j\}$$

$$\cup \left\{ i \mid \begin{array}{l} n_i \text{ appears in an expression } \Psi_j \\ \text{via a path going only through } f \text{ via indices } i' \in I \end{array} \right\}$$

- First, n_1 is relevant because of the boundary conditions.
- To compute n_1 , we need n_3 , and thus also n_2 .

Irrelevant variables analysis – Fixpoint formulation (2)

$$\begin{cases} f(n_1, n_2, n_3, n_4, n_5) = 0 & \text{if } n_1 \leq 0 \\ f(n_1, n_2, n_3, n_4, n_5) = 1 + f(n_1 - n_3 - 1, n_1 \times n_2, n_2 \times n_3, n_3 \times n_4, n_4 \times n_5) & \text{if } n_1 > 0. \end{cases}$$

- Again, we overapproximate the set of *relevant indices*, as the lfp of F .

$$F : \mathcal{P}(\llbracket 1, k \rrbracket) \rightarrow \mathcal{P}(\llbracket 1, k \rrbracket)$$

$$I \mapsto I \cup \{i \mid n_i \text{ appears in a condition } \phi_j\}$$

$$\cup \left\{ i \mid \begin{array}{l} n_i \text{ appears in an expression } \Psi_j \\ \text{via a path going only through } f \text{ via indices } i' \in I \end{array} \right\}$$

- First, n_1 is relevant because of the boundary conditions.
- To compute n_1 , we need n_3 , and thus also n_2 .
- $\text{lfp } F = \{1, 2, 3\}$, we can rewrite the equation to

$$\begin{cases} \tilde{f}(n_1, n_2, n_3) = 0 & \text{if } n_1 \leq 0 \\ \tilde{f}(n_1, n_2, n_3) = 1 + \tilde{f}(n_1 - n_3 - 1, n_1 \times n_2, n_2 \times n_3) & \text{if } n_1 > 0. \end{cases} .$$

Irrelevant variables analysis – Before/After

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```
-- module(_, [fact/2], [assertions, regtypes, nativeprops, resdefs]).

-- resource fuel.
-- head_cost(ub, fuel, 0).
-- literal_cost(ub, fuel, 0).
-- default_cost(ub, fuel, 0).
-- trust_default+cost(ub, fuel, 0).

-- entry fact(N,R) : num * var.
fact(N,R) :-
    A is 1,
    fact_aux(N,A,R).

fact_aux(N,A,R) :-
    N > 0,
    mul(A,N,A1),
    N1 is N - 1,
    fact_aux(N1,A1,R).

fact_aux(0,A,R) :-
    R = A.

-- impl_defined([mul/3]).

-- trust pred mul(X,Y,Z)
--   : num * num * var
--   => num * num * num
--   + (not_fails, is_det, cost(ub,fuel,1), cost(lb,fuel,1)).

mul(X,Y,Z) :-
    Z is X * Y.
```

-- true pred fact(N,R)
-- : (num(N), var(R))
-- => (num(N), num(R),
-- size(lb,int,R,inf))
-- + cost(lb,fuel,0).

-- true pred fact(N,R)
-- : (num(N), var(R))
-- => (num(N), num(R),
-- size(ub,int,R,inf))
-- + cost(ub,fuel,inf).

fact(N,R) :-
 A is 1,
 fact_aux(N,A,R).

-- true pred fact_aux(N,A,R)
-- : (num(N), num(A), term(R))
-- => (num(N), num(A), num(R)).

-- true pred fact_aux(N,A,R)
-- : (mshare([|R|]),
-- var(R), ground([N,A]), num(N), num(A), term(R))
-- => (ground([N,A,R]), num(N), num(A), num(R))
-- + (not_fails, covered).

-- true pred fact_aux(N,A,R)
-- : (num(N), num(A), var(R))
-- => (num(N), num(A), num(R),
-- size(lb,int,R,inf))
-- + cost(lb,fuel,inf).

-- true pred fact_aux(N,A,R)
-- : (num(N), num(A), var(R))
-- => (num(N), num(A), num(R),
-- size(ub,int,R,inf))
-- + cost(ub,fuel,inf).

-*- fact.pl All L19 (Ciao)

-*- fact_termins_of_resources_co.pl 17% L65 (Ciao)

Irrelevant variables analysis – Before/After

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```
-- module(_, [fact/2], [assertions, regtypes, nativeprops, resdefs]).

-- resource fuel.
-- head_cost(ub, fuel, 0).
-- literal_cost(ub, fuel, 0).
-- default_cost(ub, fuel, 0).
-- trust_default+cost(ub, fuel, 0).

-- entry fact(N,R) : num * var.
fact(N,R) :-
    A is 1,
    fact_aux(N,A,R).

fact_aux(N,A,R) :-
    N > 0,
    mul(A,N,A1),
    N1 is N - 1,
    fact_aux(N1,A1,R).

fact_aux(0,A,R) :-
    R = A.

-- impl_defined([mul/3]).

-- trust pred mul(X,Y,Z)
--   : num * num * var
--   => num * num * num
--   + (not_fails, is_det, cost(ub,fuel,1), cost(lb,fuel,1)).

mul(X,Y,Z) :-
    Z is X * Y.
```

-- true pred fact(N,R)
-- : (num(N), var(R))
-- => (num(N), num(R),
-- size(lb,int,R,0))
-- + cost(lb,fuel,0).

-- true pred fact(N,R)
-- : (num(N), var(R))
-- => (num(N), num(R),
-- size(ub,int,R,inf))
-- + cost(ub,fuel,int(N)).

fact(N,R) :-
 A is 1,
 fact_aux(N,A,R).

-- true pred fact_aux(N,A,R)
-- : (num(N), num(A), term(R))
-- => (num(N), num(A), num(R)).

-- true pred fact_aux(N,A,R)
-- : (mshare([|R|]),
-- var(R), ground([N,A]), num(N), num(A), term(R))
-- => (ground([N,A,R]), num(N), num(A), num(R))
-- + (possibly_fails, possibly_not_covered).

-- true pred fact_aux(N,A,R)
-- : (num(N), num(A), var(R))
-- => (num(N), num(A), num(R),
-- size(lb,int,R,0))
-- + cost(lb,fuel,0).

-- true pred fact_aux(N,A,R)
-- : (num(N), num(A), var(R))
-- => (num(N), num(A), num(R),
-- size(ub,int,R,inf))
-- + cost(ub,fuel,int(N)).

fact.pl All L19 (Ciao)
Mark set

fact_termins_of_resources.co.pl 17% L46 (Ciao)

Irrelevant variables analysis – Before/After

The screenshot shows the Ciao Prolog IDE interface with two windows side-by-side, illustrating the results of irrelevant variables analysis.

Left Window (fact.pl):

```
-- module(...,[fact/2],[assertions, regtypes, nativeprops, resdefs]).  
  
:- resource fuel.  
:- head_cost(ub,fuel,0).  
:- literal_cost(ub,fuel,0).  
:- default_cost(ub,fuel,0).  
:- trust_default+cost(ub,fuel,0).  
  
:- entry fact(N,R) : num * var.  
fact(N,R) :-  
    A is 1,  
    fact_aux(N,A,R).  
  
fact_aux(N,A,R) :-  
    N > 0,  
    mul(A,N,A1),  
    N1 is N - 1,  
    fact_aux(N1,A1,R).  
  
fact_aux(0,A,R) :-  
    R = A.  
  
:- impl_defined([mul/3]).  
  
:- trust pred mul(X,Y,Z)  
    : num * num * var  
    => num * num * num  
    + (not_fails, is_det, cost(ub,fuel,1), cost(lb,fuel,1)).  
  
mul(X,Y,Z) :-  
    Z is X * Y.
```

Right Window (fact_termin_nf_resources.co.pl):

```
-- true pred fact(N,R)  
  : ( num(N), var(R) )  
=> ( num(N), num(R),  
      size(lb,int,R,0) )  
  + cost(lb,fuel,0).  
  
-- true pred fact(N,R)  
  : ( num(N), var(R) )  
=> ( num(N), num(R),  
      size(ub,int,R,inf) )  
  + cost(ub,fuel,int(N)).  
  
fact(N,R) :-  
    A is 1,  
    fact_aux(N,A,R).  
  
:- true pred fact_aux(N,A,R)  
  : ( num(N), num(A), term(R) )  
=> ( num(N), num(A), num(R) ).  
  
:- true pred fact_aux(N,A,R)  
  : ( mshare([[N|R]]),  
      var(R), ground([N|A]), num(N), num(A), term(R) )  
=> ( ground([N,A,R]), num(N), num(A), num(R) )  
  + (possibly_fails, possibly_not_covered).  
  
:- true pred fact_aux(N,A,R)  
  : ( num(N), num(A), var(R) )  
=> ( num(N), num(A), num(R),  
      size(lb,int,R,0) )  
  + cost(lb,fuel,0).  
  
:- true pred fact_aux(N,A,R)  
  : ( num(N), num(A), var(R) )  
=> ( num(N), num(A), num(R),  
      size(ub,int,R,inf) )  
  + cost(ub,fuel,int(N)).
```

Both windows show the same code, but the right window's code has been annotated with analysis results and costs, demonstrating the transformation from the original code to one that is more efficient or easier to analyze.

Other analyse + rewrite passes could be implemented,
e.g. using change of variables or usage of inferred ranking functions.

Order

Automated Approximate Recurrence Solving

- Design a *generic, approximate* recurrence solver.

Automated Approximate Recurrence Solving

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- Observation: just like program semantics, solutions of recurrence equations may be seen as (least) fixed points of a well-chosen operator.

$$\begin{cases} f(0) = a \\ f(n) = f(f(n - 1)) + 1, \quad \forall n \in \mathbb{N}^* \end{cases} \iff$$

$$\boxed{\begin{aligned} S : (\mathbb{N} \rightarrow \mathbb{N}) &\rightarrow (\mathbb{N} \rightarrow \mathbb{N}) \\ g &\mapsto \left(n \mapsto \begin{cases} a & \text{if } n = 0 \\ g(g(n - 1)) + 1 & \text{otherwise} \end{cases} \right) \end{aligned}}$$

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- To make the theory work well, restrict ourselves to *monotone* recurrence equations, and use $D \stackrel{\Delta}{=} \mathbb{N}_\infty \rightarrow \mathbb{N}_\infty$, which is a *complete lattice*, as the *domain of sequences*.

First idea: Abstract interpretation of Recurrence equations

- With $S : D \rightarrow D$ an equation/operator on the concrete domain of sequences, $f_{\text{sol}} = \text{lfp } S = \sup_{\alpha \in \text{Ord}} S^{(\alpha)}(\perp)$.
- Use an abstract domain D^\sharp and abstract operator S^\sharp , $f_{\text{sol}} \leq \gamma(\text{lfp } S^\sharp)$.
- Using subsets of D (e.g. all affine sequences) as D^\sharp doesn't work well, so we use the power trick and use the domain of *sets of abstract bounds*.

Definition (Domain of abstract bounds)

Let A be a domain of abstract sequences, e.g.

$A = \text{all affine sequences} \cong \mathbb{N}_\infty \times \mathbb{N} + \{\top_A\}$, with $\phi : A \rightarrow D$ a concretisation.

We set $D^\sharp \triangleq (\mathcal{P}(A), \supseteq)$.

Proposition

We have a Galois connection $D \xrightleftharpoons[\alpha]{\gamma} D^\sharp$, defined by

$$\alpha : D \rightarrow D^\sharp$$

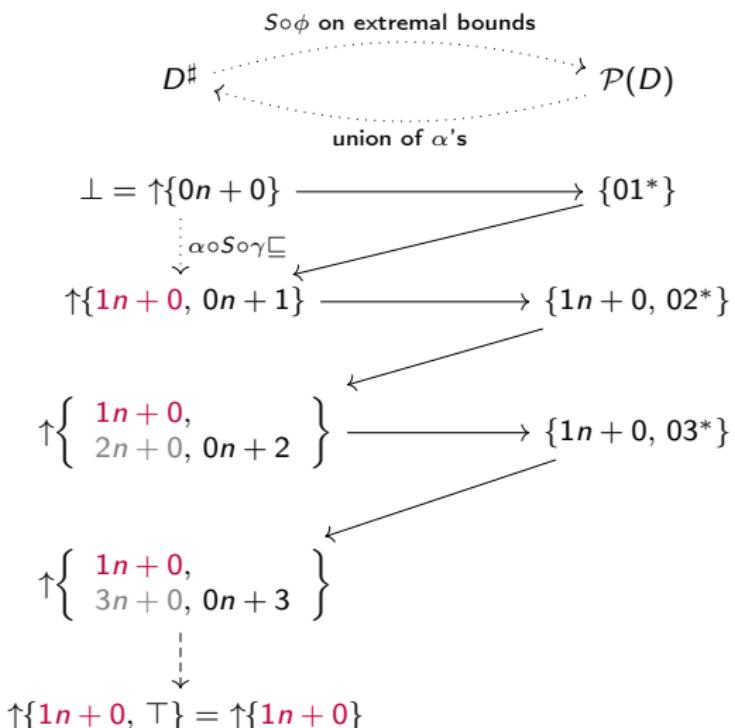
$$\gamma : D^\sharp \rightarrow D$$

$$f \mapsto \{f^\sharp \in A \mid f \leq_D \phi(f^\sharp)\}, \quad U \mapsto (n \mapsto \min_{f^\sharp \in U} \phi(f^\sharp)(n)).$$

Moreover, for any U , $\gamma \circ \alpha(U) = \uparrow U$. In particular, $\alpha(U) = \alpha(\uparrow U)$.

First idea: Abstract interpretation of Recurrence equations

In the ideal case, for an equation like $f(0) = 0$, $f(n) = f(f(n - 1)) + 1$ when $n > 0$, we could do

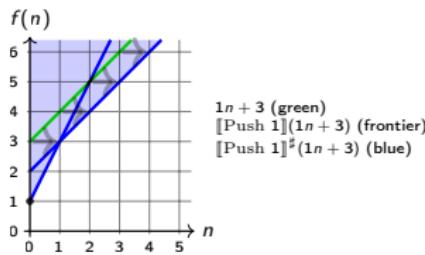


First idea: Abstract interpretation of Recurrence equations

In practice, we don't have direct access to S , γ and α , so we need to design *transfer functions*. An abstract semantic $[\![\cdot]\!]^\sharp : \text{Eqs} \rightarrow (\mathcal{A} \rightarrow \mathcal{P}(\mathcal{A}))$ is defined and extended to $\text{Eqs} \rightarrow (\mathcal{P}(\mathcal{A}) \rightarrow \mathcal{P}(\mathcal{A}))$. For example, with affine functions,

$$\begin{aligned}
 [\![\text{Cst } c]\!]^\sharp(f^\sharp) &= \{(0, c)\} \\
 [\![n]\!]^\sharp(f^\sharp) &= \{(1, 0)\} \\
 [\![f]\!]^\sharp(f^\sharp) &= \{f^\sharp\} \\
 [\![\Diamond]\!]^\sharp(\top_A, g^\sharp) &= [\![\Diamond]\!]^\sharp(f^\sharp, \top_A) = \{\top_A\} \\
 \text{for } \Diamond \in \{+, \times, \circ, -\} \\
 [\![+]\!]^\sharp((a_1, b_1), (a_2, b_2)) &= \{(a_1 + a_2, b_1 + b_2)\} \\
 [\![\times]\!]^\sharp((a_1, b_1), (a_2, b_2)) &= \{(a_1 a_2 \infty + a_1 b_2 + a_2 b_1, b_1 b_2)\} \\
 \text{where we set } 0 \times \infty = \infty \\
 [\![\circ]\!]^\sharp((a_1, b_1), (a_2, b_2)) &= \{(a_1 a_2, a_1 b_2 + b_1)\} \\
 [\![-]\!]^\sharp((a_1, b_1), (a_2, b_2)) &= \{(a_1 - a_2, b_1 - b_2)\} \\
 [\![\text{Mult}_{in} c]\!]^\sharp((a, b)) &= \{(ca, b)\} \\
 [\![\text{Div}_{in} c]\!]^\sharp((a, b)) &= \left\{ \left(\left\lceil \frac{a}{c} \right\rceil, b \right) \right\}
 \end{aligned}$$

$$\begin{aligned}
 [\![\text{Push } c]\!]^\sharp(f^\sharp) &= \begin{cases} \{(\infty, c)\} & \text{if } f^\sharp = \top_A \\ \{(a, c)\} & \text{if } f^\sharp = (a, b) \text{ and } b \leq a + c \\ \{(a, b - a), (b - c, c)\} & \text{if } f^\sharp = (a, b) \text{ and } b > a + c \end{cases} \\
 [\![\text{Pop}]\!]^\sharp((a, b)) &= \begin{cases} \{\top_A\} & \text{if } f^\sharp = \top_A \\ \{(a, b + a)\} & \text{if } f^\sharp = (a, b) \end{cases} \\
 [\![\text{Shift } \delta]\!]^\sharp(f^\sharp) &= \begin{cases} ([\![\text{Push } 0]\!]^\sharp)^\delta(f^\sharp) & \text{if } \delta \geq 0 \\ \{\top_A\} & \text{if } \delta \leq 0 \text{ and } f^\sharp = \top_A \\ \{(a, b + |\delta|a)\} & \text{if } \delta \leq 0 \text{ and } f^\sharp = (a, b) \end{cases} \\
 [\![\text{Set}_0 c]\!]^\sharp(f^\sharp) &= [\![\text{Push } c]\!]^\sharp \circ [\![\text{Pop}]\!]^\sharp(f^\sharp) \\
 &= \begin{cases} \{(\infty, c)\} & \text{if } f^\sharp = \top_A \\ \{(a, c)\} & \text{if } f^\sharp = (a, b) \text{ and } b \leq c \\ \{(a, b), ((a + b) - c, c)\} & \text{if } f^\sharp = (a, b) \text{ and } b > c \end{cases}
 \end{aligned}$$



Second idea: Pre-/Post-fixpoints as easily checkable bounds

- By the properties of lattices, for a monotone equation/operator S , if $f \geq S(f)$ (prefix), then $f \geq \text{lfp } S$.
Similarly, if $f \leq S(f)$ (postfix), then $f \leq \text{gfp } S$.
- In particular, when we can prove that there is only one solution to $f = S(f)$, we get $f_{\text{sol}} \leq f$ for any prefixpoint and $f \leq f_{\text{sol}}$ for any postfixpoint.
- This gives a new simple proof method for complexity analysis equations.

Example

Consider the following equation, written as a sequence operator. Such equation may arise while doing worst-case analysis of a quicksort program. S is indeed monotone.

$$S : f \mapsto n \mapsto \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ \max_{1 \leq k \leq n-1} f(k) + f(n-k) + n & \text{otherwise} \end{cases}$$

- $g : n \mapsto n^2$?

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- $g : n \mapsto n^2$?
- $(Sg)(0) = 0$, $(Sg)(1) = 1$, and $(Sg)(n) = 1^2 + (n-1)^2 + n = n^2 - n + 2$ when $n \geq 2$, thus $Sg \leq g$, thus $f_{\text{sol}}(n) \leq n^2$.

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- $g : n \mapsto \frac{1}{2}n^2$?
- $(Sg)(0) \geq g(0)$, $(Sg)(1) \geq g(1)$ and
 $(Sg)(n) = \frac{1}{2}(1^2 + (n-1)^2) + n = \frac{1}{2}n^2 + 1 \geq g(n)$, thus $g \leq Sg$, thus
 $\frac{1}{2}n^2 \leq f_{\text{sol}}(n)$.

Second idea: Pre-/Post-fixpoints as easily checkable bounds

- By the properties of lattices, if we can prove that there is only one solution to $f = S(f)$, we get $f_{\text{sol}} \leq f$ for any prefixpoint and $f \leq f_{\text{sol}}$ for any postfixpoint.
- This gives a new simple proof method for complexity analysis equations.
- Because it is simple, it is amenable to automation.
 - Use some sort of divide-and-conquer heuristics ?
Recent papers have improved dichotomy to deal with the lattice $([1, n])^d$ in logarithmic time, but their technique to deal with the case $f \not\leq Sf$ isn't easy to translate to general lattices.
 - Use some some of guess-and-check method ?
A current technique (Lasso + SMT) tries to *fit* the exact solution using common complexity functions, and then check equality.
Guessing *bounds* and then checking *inequality* seems more likely to succeed.
 - Note that automatic check of $f_{\text{sol}} \leq g$ is much harder than $f_{\text{sol}} = g$ given only the equation, and our method provides a way to do so in many cases.

Conclusion

Summary and Next Steps

Contributions

- Implementation – Classical Recurrence Solver
 - Small additions to the “table lookup” side of the solver,
 - New analyse + rewrite pass for irrelevant variable analysis,
 - Various small bugfixes.
- Theory – New order-theoretical point of view
 - Abstract iteration, with a few domains,
 - Pre-/Post-fixpoints as “easily checkable bounds”.

Possible Next Steps

- Implementation of algorithms, with their extensions.
- Design of new domains, e.g. using results of experiments with SMT solvers.
- Exploration of the new abstract point of views on “ideal size measures” as optimal abstractions, on $HCIR^{Sz} \leftrightarrow \text{RecEq}$ for rewritings, etc.
- Finish benchmark work on “the recurrence structure of real programs”.
- Improvement in translation of imperative programs to equations (inference of ranking functions, output/output size relations, preservation of type properties, ...).
- Usage of analysis for optimisation.
- Improvement in energy models (VHDL, low-level runtime policies, etc.).
- Go further in contacts with industry/associations.
- At some point, include our cost analysis as a plugin of other tools? (Clang, Frama-C, ...).

Thank you !