Automatic Granularity-Aware Parallelization of Programs with Predicates, Functions, and Constraints

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Objectives

- Parallelism (*finally*!) becoming mainstream thanks to *multicore* –even on laptops!
- Our objective herein is *automatic parallelization* of programs with predicates, functions, and constraints.
- We concentrate on detecting and-parallelism (corresponds to, e.g., loop parallelization, task parallelism, divide and conquer, etc.):

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- Parallelism (*finally*!) becoming mainstream thanks to *multicore* –even on laptops!
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- We concentrate on detecting and-parallelism (corresponds to, e.g., loop parallelization, task parallelism, divide and conquer, etc.):

```
fib(0) := 0.
fib(1) := 1.
fib(N) := fib(N-1)+fib(N-2)
      :- N>1.
fib(0, 0).
fib(1, 1).
fib(N, F) :-
      N>1,
      ( N1 is N-1,
      fib(N1, F1) ) &
      ( N2 is N-2,
      fib(N2, F2) ),
      F1+F2.
```

 \rightarrow Need to detect *independent* tasks.

What is Independence? (for Functions, Predicates, Constraints, ...)

- Correctness: "same" solutions as sequential execution.
- Efficiency: execution time < than seq. program (or, at least, no-slowdown: <). (We assume parallel execution has no overhead in this first stage.)

0	Running	s_1	//	s_2 :
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	Imperative	Functions	Constraints	
s_1	Y := W+2;	(+ W 2)	Y = W+2,	
s_2	X := Y+Z;	(+ Z)	X = Y + Z,	
	read-write deps	strictness	cost!	

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T	$1101111119 \ s_1 \ n \ s_2.$	s_2	X := Y+Z;	(+ Z)	X = Y + Z,	
			read-write deps	strictness	cost!	
	For <i>Predicates</i> (multiple procedure definitions):					
	main:-		p(X) :- X=a.			
	s_1 p(X), s_2 q(X), write(X).		q(X) :- X=b, large computation. q(X) :- X=a.			
	Again, cost issue: if p affects q (prunes its choices) then q ahead of p is speculative.					

Independence: condition that guarantees correctness and efficiency.

Independence

- Strict independence (suff. condition): no "pointers" shared at run-time:
- Non-strict independence: only one thread accesses each shared variable.
 - Requires global analysis.
 - Required in programs using "incomplete structures" (difference lists, etc.).

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Sufficient a-priori condition: given $g_1(\bar{x})$ and $g_2(\bar{y})$, c state just before them:

$$(\bar{x} \cap \bar{y} \subseteq def(c)) \ and \ (\exists_{-\bar{x}}c \land \exists_{-\bar{y}}c \to \exists_{-\bar{y}\cup\bar{x}}c))$$

(def(c) = set of variables constrained to a unique value in c)

• For
$$c = \{x > y, z > y\}$$

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• $\overline{\exists}_{\{x,z\}}c = x > z$

Approximation: presence of "links" through the store.

Parallelization Process

- Conditional dependency graph (of some code segment, e.g., a clause):
 - Vertices: possible tasks (statements, calls,...),
 - Edges: possible dependencies (labels: conditions needed for independence).
- Local or global analysis used to reduce/remove checks in the edges.
- Annotation process converts graph back to parallel expressions in source.



- One of the popular Prolog/CLP systems (supports ISO-Prolog fully).
- At the same time, new-generation multi-paradigm language/prog.env. with:
 - Predicates, constraints, functions (including lazyness), higher-order, ... (And Prolog impure features only present as compatibility libraries.)

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 - Automatic parallelization.
 - Automatic granularity and resource control.

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 - Automatic parallelization.
 - Automatic granularity and resource control.
 - + several control rules (e.g., bf, id, Andorra), objects, syntactic/semantic extensibility, LGPL, ...

Some Speedups (for different analysis abstract domains)



The parallelizer, self-parallelized

- Replace parallel with sequential execution based on task size and overheads.
- Cannot be done completely at compile-time: cost often depends on input (hard to approximate at compile time, even w/abstract interpretation).
 main :- read(X), read(Z), inc_all(X,Y) & r(Z,M), ...

Granularity Control

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- Cannot be done completely at compile-time: cost often depends on input (hard to approximate at compile time, even w/abstract interpretation).
 main :- read(X), read(Z), inc_all(X,Y) & r(Z,M), ...
- Our approach:
 - Derive at compile-time cost *functions* (to be evaluated at run-time) that efficiently bound task size (lower, upper *bounds*).
 - Transform programs to carry out run-time granularity control.



For inc_all, (assuming "threshold" is 100 units):

Inference of Bounds on Argument Sizes and Procedure Cost in CiaoPP

- 1. Perform type/mode inference: :- true inc_all(X,Y) : list(X,int), var(Y) => list(Y,int).
- 2. Infer size measures: list length.
- 3. Use data dependency graphs to determine the relative sizes of structures that variables point to at different program points infer argument size relations:

$$\begin{split} & \texttt{Size}_{\texttt{inc_all}}^2(0) = 0 \text{ (boundary condition from base case),} \\ & \texttt{Size}_{\texttt{inc_all}}^2(n) = 1 + \texttt{Size}_{\texttt{inc_all}}^2(n-1). \end{split}$$

$$\mathsf{Sol} = \mathtt{Size}_{\mathtt{inc_all}}^2(n) = n.$$

4. Use this, set up recurrence equations for the computational cost of procedures:

$$\begin{split} & \operatorname{Cost}_{\mathtt{inc_all}}^{\mathrm{L}}(0) = 1 \text{ (boundary condition from base case),} \\ & \operatorname{Cost}_{\mathtt{inc_all}}^{\mathrm{L}}(n) = 2 + \operatorname{Cost}_{\mathtt{inc_all}}^{\mathrm{L}}(n-1). \end{split}$$

$$\mathsf{Sol} = \mathsf{Cost}_{\texttt{inc_all}}^\mathsf{L}(n) = 2 \; n+1.$$

- We obtain lower/upper bounds on task granularities.
- Non-failure (absence of exceptions) analysis needed for lower bounds.

Simplification of cost functions:



```
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..., (length(X) > 50 -> inc_all(X,Y) & r(Z,M)
; inc_all(X,Y), r(Z,M)), ...
..., (length_gt(LX,50) -> inc_all(X,Y) & r(Z,M)
; inc_all(X,Y), r(Z,M)), ...
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• Complex thresholds: use also communication cost functions, load, ... [Example:] Assume $CommCost(inc_all(X)) = 0.1$ (length(X) + length(Y)). We know $ub_length(Y)$ (actually, exact size) = length(X); thus:

$$2 \ length(X) + 1 > 0.1 \ (length(X) + length(X)) \cong$$
$$2 \ length(X) > 0.2 \ length(X) \equiv$$
$$2 > 0.2$$

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- Checking of data sizes can be stopped once under threshold.
- Data size computations can often be done on-the-fly.
- Static task clustering (loop unrolling), static placement, etc.

Granularity Control System Output Example

```
g_qsort([], []).
g_qsort([First|L1], L2) :-
  partition3o4o(First, L1, Ls, Lg, Size_Ls, Size_Lg),
  Size_Ls > 20 -> (Size_Lg > 20 -> g_qsort(Ls, Ls2) & g_qsort(Lg, Lg2)
                                ; g_qsort(Ls, Ls2) , s_qsort(Lg, Lg2))
               ; (Size_Lg > 20 -> s_qsort(Ls, Ls2) , g_qsort(Lg, Lg2)
                                ; s_qsort(Ls, Ls2) , s_qsort(Lg, Lg2))),
  append(Ls2, [First|Lg2], L2).
partition3o4o(F, [], [], [], 0, 0).
partition3o4o(F, [X|Y], [X|Y1], Y2, SL, SG) :-
  X =< F, partition3o4o(F, Y, Y1, Y2, SL1, SG), SL is SL1 + 1.
partition3o4o(F, [X|Y], Y1, [X|Y2], SL, SG) :-
   X > F, partition3o4o(F, Y, Y1, Y2, SL, SG1), SG is SG1 + 1.
```

Refinements (2): Granularity-Aware Annotation

With classic annotators (MEL, UDG, CDG, ...) we applied granularity control after parallelization:



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With classic annotators (MEL, UDG, CDG, ...) we applied granularity control after parallelization:



- Developed new annotation algorithm that takes task granularity into account:
 - Annotation is a heuristic process (several alternatives possible).
 - Taking task granularity into account during annotation can help make better choices and speed up annotation process.
 - Tasks with larger cost bounds given priority, small ones not parallelized.



Granularity-Aware Annotation: Concrete Example

- Consider the clause: p := a, b, c, d, e.
- Assume that the dependencies detected between the subgoals of p are given by:



Assume also that:

T(a) < T(c) < T(e) < T(b) < T(d),

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where T(i) < T(j) means: cost of subgoal i is smaller than the cost of j.

MEL annotator: (a, b & c, d & e)
UDG annotator: (c & (a, b, e), d)
Granularity-aware: (a, c, (b & d), e)

Refinements (3): Using Execution Time Bounds/Estimates

- Use estimations/bounds on *execution time* for controlling granularity (instead of steps/reductions).
- Execution time generally dependent on platform characteristics (\approx constants) and input data sizes (unknowns).
- Platform-dependent, one-time calibration using fixed set of programs:
 - Obtains value of the platform-dependent constants (costs of basic operations).
- Platform-independent, compile-time analysis:
 - Infers cost functions (using modification of previous method), which return count of *basic operations* given input data sizes.
 - Incorporate the constants from the calibration.
 - \rightarrow we obtain functions yielding *execution times* depending on size of input.
- Predicts execution times with reasonable accuracy (challenging!).
- Improving by taking into account lower level factors (current work).

Execution Time Estimation: Concrete Example

Consider nrev with mode:

```
:- pred nrev/2 : list(int) * var.
```

Estimation of execution time for a concrete input —consider:

 $A = [1,2,3,4,5], \overline{n} = \text{length}(A) = 5$

	Once	Static Analysis	Application	
component	K_{ω_i}	$Cost_{p}(I(\omega_{i}),\overline{n}) = C_{i}(\overline{n})$	$C_i(5)$	$K_{\omega_i} \times C_i(5)$
step	21.27	$0.5 \times n^2 + 1.5 \times n + 1$	21	446.7
nargs	9.96	$1.5 \times n^2 + 3.5 \times n + 2$	57	567.7
giunif	10.30	$0.5 \times n^2 + 3.5 \times n + 1$	31	319.3
gounif	8.23	$0.5 \times n^2 + 0.5 \times n + 1$	16	131.7
viunif	6.46	$1.5 \times n^2 + 1.5 \times n + 1$	45	290.7
vounif	5.69	$n^2 + n$	30	170.7
Execution time $\overline{K}_{\Omega} \bullet \overline{\text{Cost}_{p}}(\overline{I(\Omega)}, \overline{n})$:				1926.8

Visualization of And-parallelism - (small) qsort, 4 processors



Fib 15, 1 processor



Fib 15, 8 processors (same scale)



Fib 15, 8 processors (full scale)



Fib 15, 8 processors, with granularity control (same scale)

