"A Source-Level Estimation and Optimization Methodology for the Execution Time and Energy Consumption of the Embedded Software"



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This thesis at a glance:

• Estimation:

- designers frequently need to estimate the {time, energy} consumption of significant clusters of operations;
- current approaches (ISS, STA, SLI)
 do not solve the problem effectively;
- we propose a new method (SLE)
 which is flexible, fast, accurate

• Optimization:

- exploring source-level optimizing transformation is a slow task
- many approaches involve ISS
- we propose a new flow which is short-loop, scalable, modular

Estimation

Previous approaches are inadequate

•	Static Timing Analys (STA) cannot deal with dynamism: - its main objective is the determination of the WCET - cannot deal with dynamic features: unbounded loops, recursion, dynamic fn ref; - unfortunately, code is becoming more and more dynamic (e.g. object based video coding, wireless ad-hoc networks,)	[Puschner89,, Chen01]
•	 Instruction-Set Simulation (ISS) is slow and at a low level: it is 10k-100k times slower than application execution; provides estimate at assembly level whereas developer works at source level; estimates are difficult to interpret: not much helpful for optimization: (deep pipelines, superscalarity, wide-issue, speculation, branch prediction,) 	[Brooks00, Sinha01, Qin03]
•	ISS + gprof provide estimates only at a function level	[Simunic01]
•	Atomium/PowerEscape is source-level, but only for memory aspects	[Bormans99, Arnout05]
•	SoftExplorer is a static techniqueuser interaction required to determine loop iterations: unthinkable for real sized projects	[Senn02]
•	Compilation-based approaches do not provide link to source level	[Lajolo99]
•	SIT is source level (good!) but still unable to resolve chosen clusters	[Ravasi03]
•	Black-box techniques do not provide any link with code	[Muttreja04]

What we do, and others can't

• Motivational example: we consider a sample fragment of real code (FFT implementation, [Guthaus01])

```
74 for (i=rev=0; i < NumBits; i++)
75 {
76   rev = (rev << 1) | (index & 1);
77   index >>= 1;
78 }
```

• After the analysis, we provide estimates for the individual operator instances

Line	Time	Time(%)	Energy	Energy(%)	Code
74	2.030 ms		980.357 uJ		for(i=rev=O; i< NumBits; i++)
75	0.000 s		0.000 J		{
76	3.796 ms		2.137 mJ		rev = (rev << 1) (index & 1);
77	1.265 ms		712.279 uJ		index >>= 1;
78	0.000 s		0.000 J		}

- Currently, no other method can provide this detailed results
- Estimation at the source-level is 10,000 x faster than an ISS

How we perform estimation

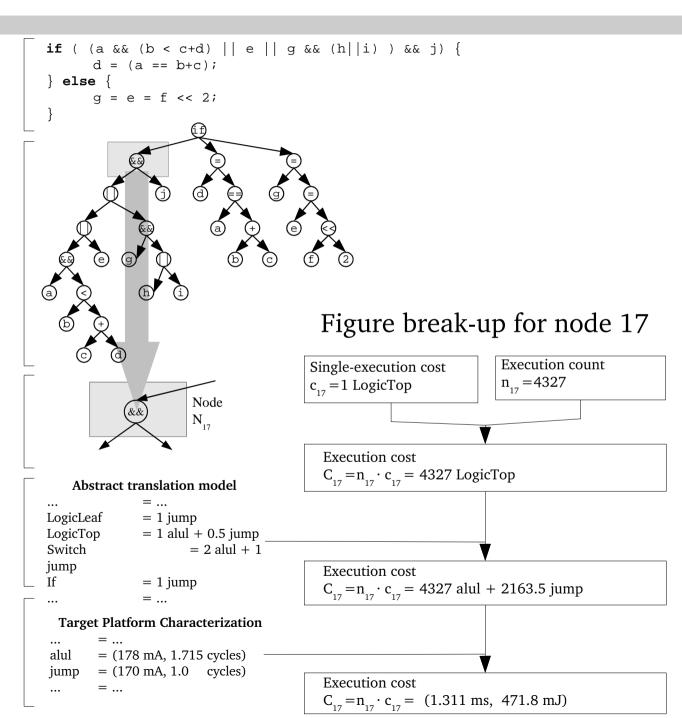
Input source code

Abstract syntax tree

Atoms

Abstract instructions

Time and energy

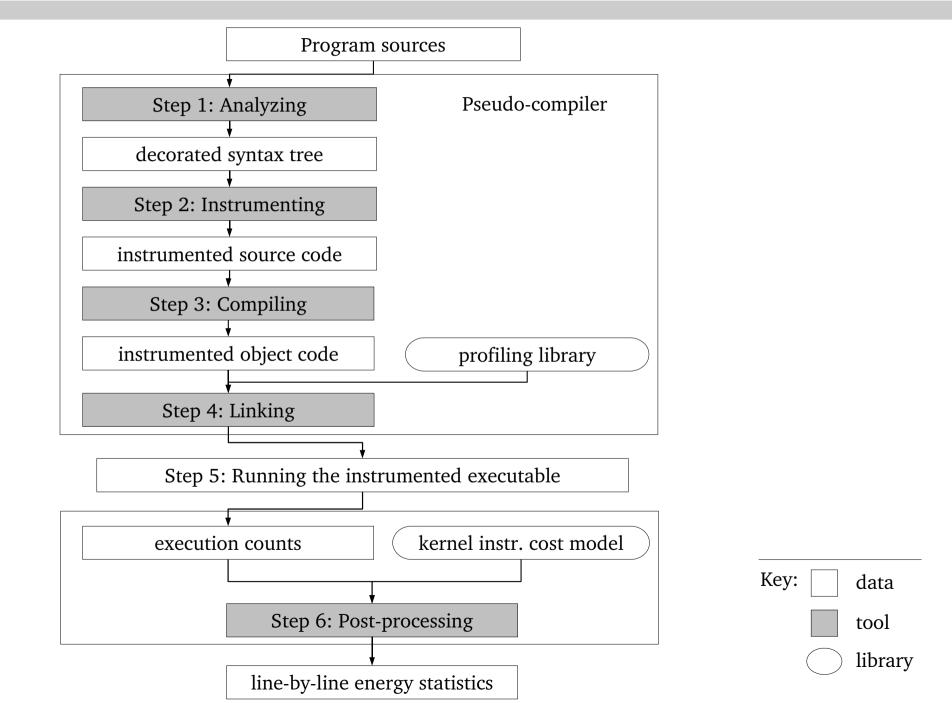


The cost of syntax elements

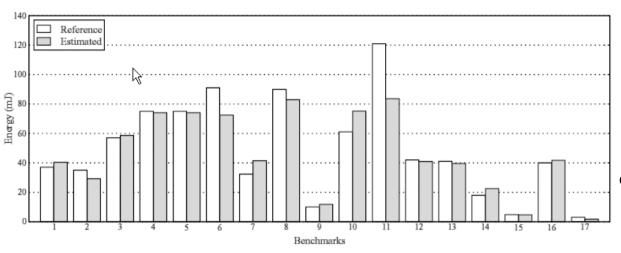
- Step 1 (Analysis) associates a single-execution cost c(i) to each syntax node, expressed as sum of atoms
- the cost is due to 3 contributions: c(i) = ci(i) + cf(i) + cc(i)
- contributions are calculated by an attribute grammar over the AST;

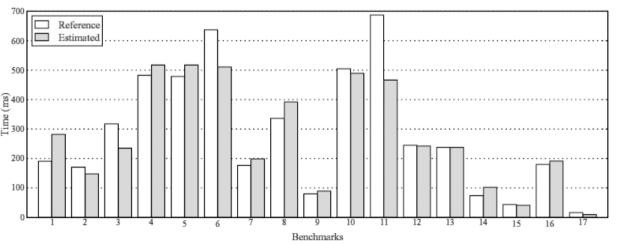
Attribute		Name	Defined for		
k	synthesized	constancy	expressions		
e	synthesized	constant value	expressions		
t	synthesized	real result type	expressions		
ν	inherited	valueness	expressions		
r	inherited	restricted result type	expressions		
f	inherited	translation flavor	expressions and statements		
ci	synthesized	inherent cost	expressions and statements		
cc	synthesized	conversion cost	expressions and statements		
cf	inherited	flow control cost	expressions and statements		
С	synthesized	total cost	expressions and statements		

Estimation: the tool flow



Results: accuracy and speed





Experimental Setup

- comparison against SimIt-Arm (cycles) [Qin03]
- current figures from JouleTrack (energy) [Sinha01]
- modelling for SA1100, 206 MHz, 1.5 V
- 24 benchmark from MiBench[Guthaus01]

Accuracy

- avg modulo error = 15% E, <17% T
- coefficients of correlation = 0.978 E, 0.960 T

Speed

- simulation times 10,350 x shorter than ISS
- simulation only 2.2x slower than normal execution

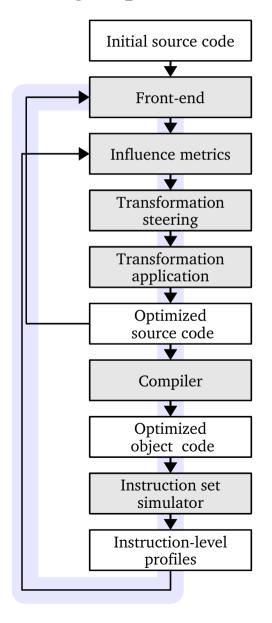
Robustness

 24/24 MiBench projects successfully processed

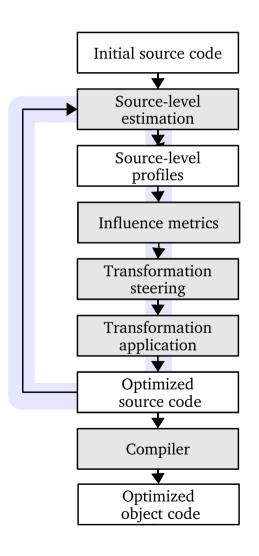
Optimization

A short-loop exploration methodology is needed

Long exploration loop



Short exploration loop



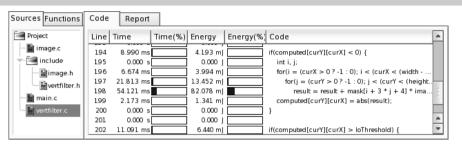
What we do and others can't

Import a project

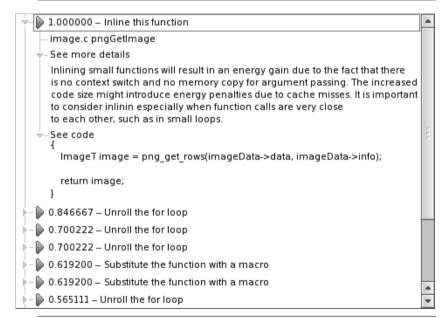
Analyze it

 Get source-level optimization directives, generated at the source level

 Apply them and see the result



File	Time	Energy
image.c	21.638 µs	16.561 μ
main.c	28.962 µs	21.158 μ
vertfilter.c	377.672 ms	421.048 m
(glibc)	305.800 µs	622.000 µ
TOTAL	378.029 ms	421.708 m



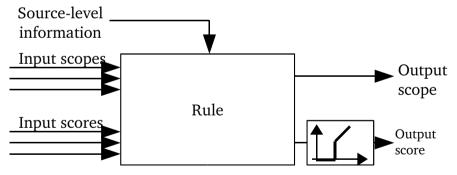
File	Time	Energy	
image.c	21.638 µs	16.561	μJ
main.c	28.962 μs	21.158	μJ
vertfilter.c	356.222 ms	396.261	mJ
(glibc)	305.800 μs	21.158	μJ
TOTAL	356.509 ms	396.921	mJ

What a short-loop methodology needs

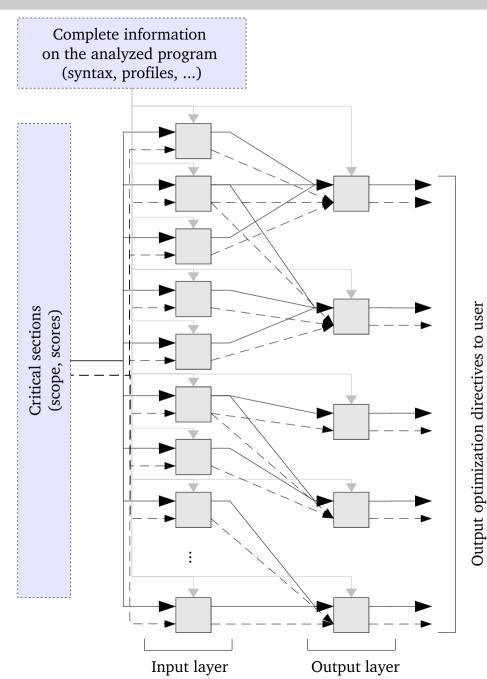
Problem	Task	Additional Requirements
source code analysis	analyze the code and determine which are the critical sections	analysis must be performed at source level; profile data must be available at source level
		SLE is the first approach
influence metrics	determine what is the gain in applying a trf over a section	
		Many exist, e.g. [Brandolese03]
transformation steering	decide which transformation to apply and where	steering engine must operate automatically on source-level data provided by above analysis and metrics
		None exist!
transformation application	apply transformation on the source code	
		e.g. [suif94]

How we perform transformation steering

- We employ a Network of Fuzzy Rules
- It is a modified version of a neural network; differences:
 - weights and connections model explicitly transformation influence metrics;
 - each rule (~neuron) accesses complete syntactic and profiling information;
- Base component: NFR rule



- Advantages:
 - scalable $O(n \cdot Q)$
 - modular (no IP disclosed)



Experimental results

Modelled transformations:

- 1) loop unrolling
- 2) function inlining
- 3) function replacement with macro
- 4) common subexpression elimination
- 5) strength reduction
- 6) type conversion elimination
- 7) standard library function factorization
- 8) memory allocation factorization
- 9) argument passing via pointer 10) function specialization
- Benchmarks:

4 applications (audio filter, hough transform, dijkstra, FFT);

- Energy gains: 5-22%
- Time gains: 8-20%

