

Complete and Practical Universal Instruction Selection

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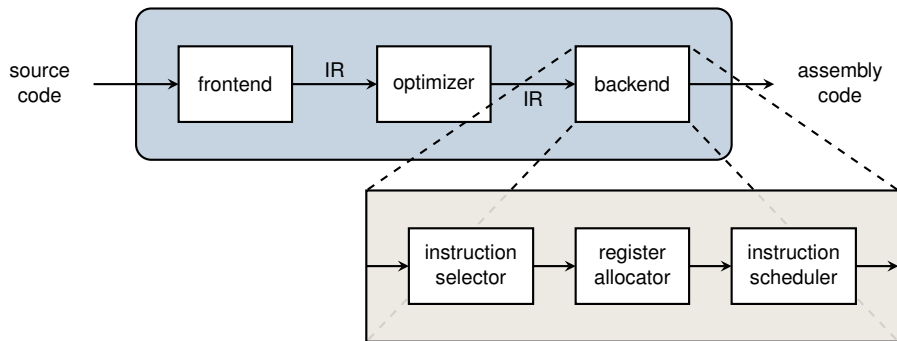
² RISE SICS, Sweden



CASES 2017, October 16

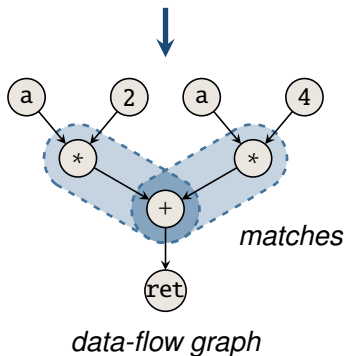
This research is supported by the
Swedish Research Council (VR 621-2011-6229) and LM Ericsson AB.

Inside a Typical Compiler

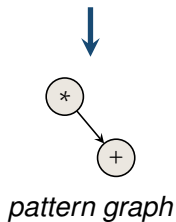


Graph-based Instruction Selection

```
int f(int a) {  
    int b = a * 2;  
    int c = a * 4;  
    return b + c;  
}
```



mulacc



Problem: Select matches such that data-flow graph is (optimally) covered (NP-complete in general)

State of the Art

- Selection is **greedy** and **local** to basic blocks
- Graphs capture **data flow only**
- Operations remain **fixed** to a given block
(lacks **global code motion**)

Observed:

- Global code motion **interacts** with selection of complex instructions
- Capturing interaction requires **non-greedy** approach

Consequence:

- **Failure to exploit** complex instructions of modern processors in embedded systems
- **Selection of complex** instruction with control flow using **handwritten, ad-hoc** routines

Talk Overview

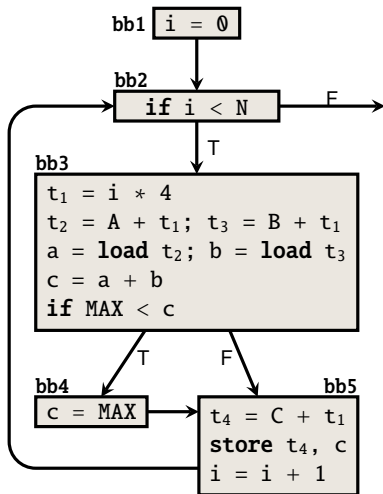
1. Introduction
2. A Motivating Example
3. Constraint Programming
4. Representations
5. Constraint Model
6. Experiments
7. Future Work and Conclusions

Outline

1. Introduction
- 2. A Motivating Example**
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Program Example

```
void satadd(int* A,  
            int* B,  
            int* C)  
{  
    int i = 0;  
    while (i < N) {  
        int c = A[i] + B[i];  
        if (MAX < c)  
            c = MAX;  
        C[i] = c;  
        i++;  
    }  
}
```

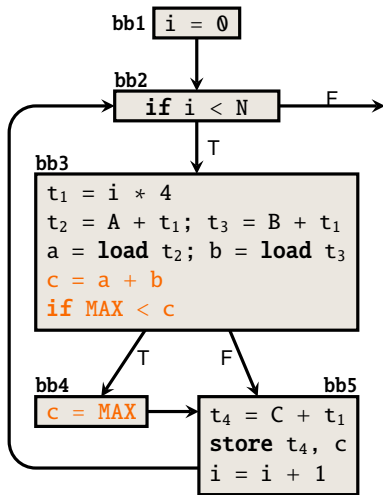


Instruction Examples

■ **satadd**

Problems:

- Incorporates control flow
- Extends across multiple blocks

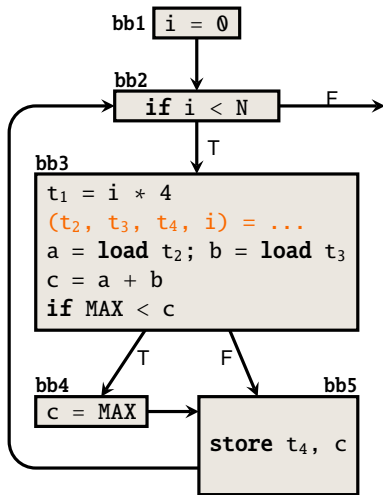


Instruction Examples

- **satadd**
- **add4**

Problems:

- Additions must be moved to same block (requires global code motion)
- Depending on hardware, may incur additional copy overhead



Universal Instruction Selection (UIS) [1]

- Handles **both control** and **data flow**
 - Enables complex instructions to be captured as pattern graphs
- Integrates **global instruction selection** (selects instructions for entire function) with **global code motion**
 - Facilitates selection of complex instructions
- Takes **data-copying overhead** into account
 - Prevents greedy selection of SIMD instructions
- Expressed as a **constraint model**
 - Potentially **optimal** w.r.t. the model
 - Allows time to be **traded** for quality

[1] G. Hjort Blindell, R. Castañeda Lozano, M. Carlsson, and C. Schulte. “Modeling Universal Instruction Selection”. In: *Proceedings of CP’15*. Springer, 2015, pp. 609–626.

Our Approach Is a Complement to Traditional Methods

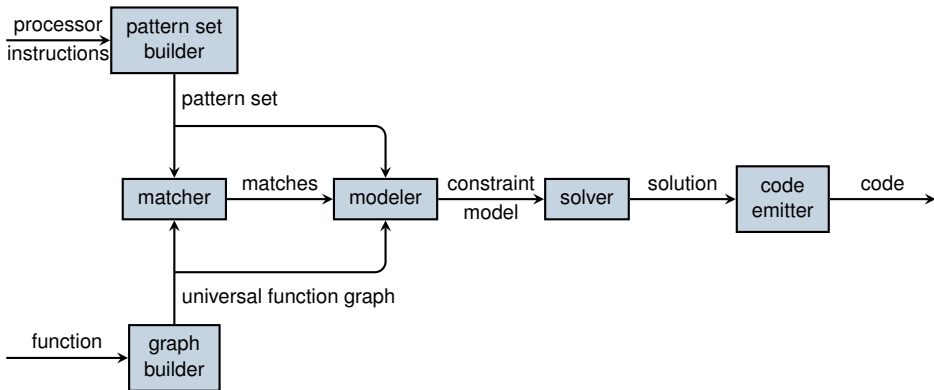
During development:

- Quick compilation times essential
- Code quality less important

Before deployment:

- Code quality essential
- Allow for long compilation times

Approach



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Constraint Programming

- Combinatorial optimization method
 - First **model** the problem, then **solve** the model
- Problems modeled as **constraint models**
 - **Variables** – decisions to be made?
 $x, y, z \in D$
 - **Constraints** – what constitute a solution?
 $x + y < z$
 - **Objective function** – what is the best solution?
maximize x
Orthogonal to the variables and constraints
 - Can be extended (**compositional**)
- Constraint models solved by interleaving
 - **Propagation** – remove values in conflict with constraint
 - **Search** – try and backtrack

Example: Sudoku

5	3			7				
6			1	9	5			
	9	8					6	
8				6				3
4			8		3			1
7				2				6
	6					2	8	x_{79} variable
			4	1	9			5
				8			7	9

Initially:

$$x_{79} \in \{1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

Row Constraint

5	3			7				
6			1	9	5			
	9	8					6	
8				6				3
4			8		3			1
7				2				6
x_{71}	6	x_{73}	x_{74}	x_{75}	x_{76}	2	8	x_{79}
			4	1	9			5
				8			7	9

Propagate $alldiff(x_{71}, 6, x_{73}, x_{74}, x_{75}, x_{76}, 2, 8, x_{79})$

$$x_{79} \in \{1, 3, 4, 5, 7, 9\}$$

Column Constraint

5	3			7				x_{19}
6			1	9	5			x_{29}
	9	8					6	x_{39}
8				6				3
4			8		3			1
7				2				6
	6					2	8	x_{79}
			4	1	9			5
				8			7	9

Propagate $alldiff(x_{19}, x_{29}, x_{39}, 3, 1, 6, x_{79}, 5, 9)$

$$x_{79} \in \{ \quad 4, \quad 7 \quad \}$$

Block Constraint

5	3			7				
6			1	9	5			
	9	8					6	
8				6				3
4			8		3			1
7				2				6
	6					2	8	x_{79}
			4	1	9	x_{87}	x_{88}	5
			8			x_{97}	7	9

Propagate $alldiff(2, 8, x_{79}, x_{87}, x_{88}, 5, x_{97}, 7, 9)$

$$x_{79} \in \left\{ \begin{array}{c} 4 \end{array} \right\}$$

After Propagation

5	3			7				
6			1	9	5			
	9	8					6	
8				6				3
4			8		3			1
7				2				6
	6					2	8	4
			4	1	9			5
				8			7	9

$$x_{79} = 4$$

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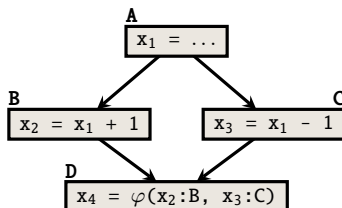
Representation

Combination of two graphs:

- Extended SSA graph
- Extended control-flow graph

Static Single Assignment (SSA) Form

- Each variable must be defined exactly once
- Use φ -functions when definition depends on control flow
- Used in virtually all modern compilers



Example Function

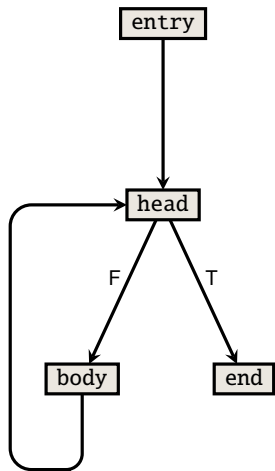
```
int fact(int n) {  
    int f = 1;  
    while (n > 1) {  
        f = f * n;  
        n--;  
    }  
    return f;  
}
```

In C

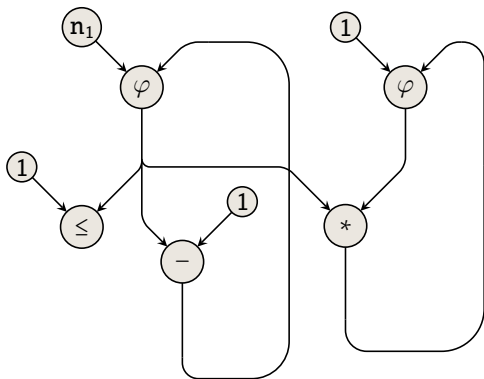
```
int fact(int n1) {  
    entry:  
        int f1 = 1;  
    head:  
        int f2 =  $\varphi$ (f1:entry, f3:body);  
        int n2 =  $\varphi$ (n1:entry, n3:body);  
        bool b = n2 <= 1;  
        if b goto end;  
    body:  
        int f3 = f2 * n2;  
        int n3 = n2 - 1;  
        goto head;  
    end:  
        return f2;  
}
```

In SSA form

Goal: Connect The Graphs

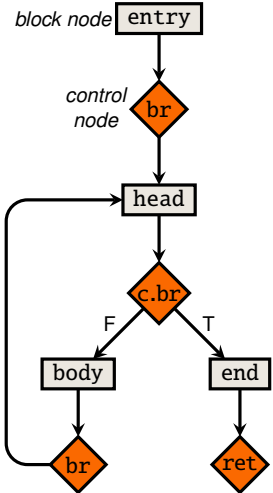


control-flow graph

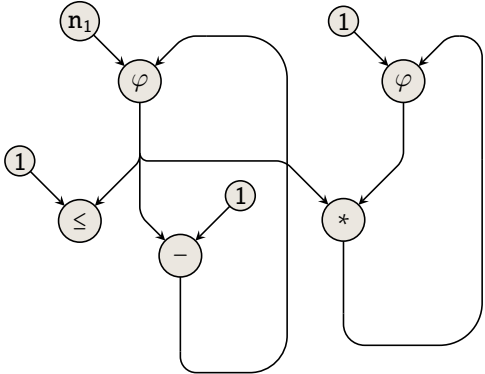


SSA graph

Extend the Control-Flow Graph

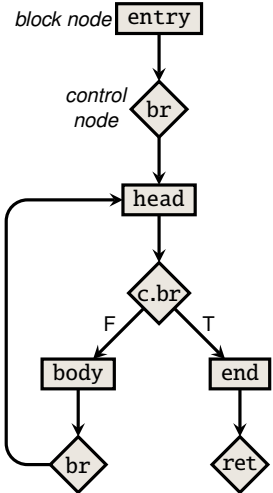


control-flow graph

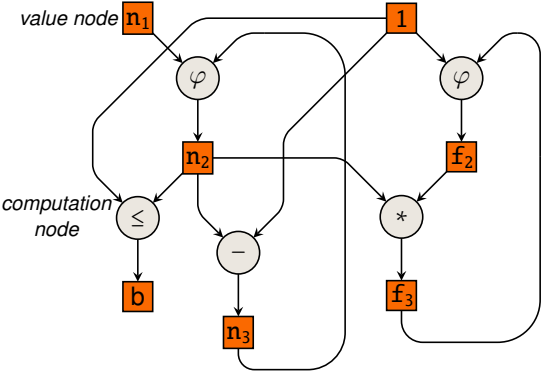


SSA graph

Extend the SSA Graph

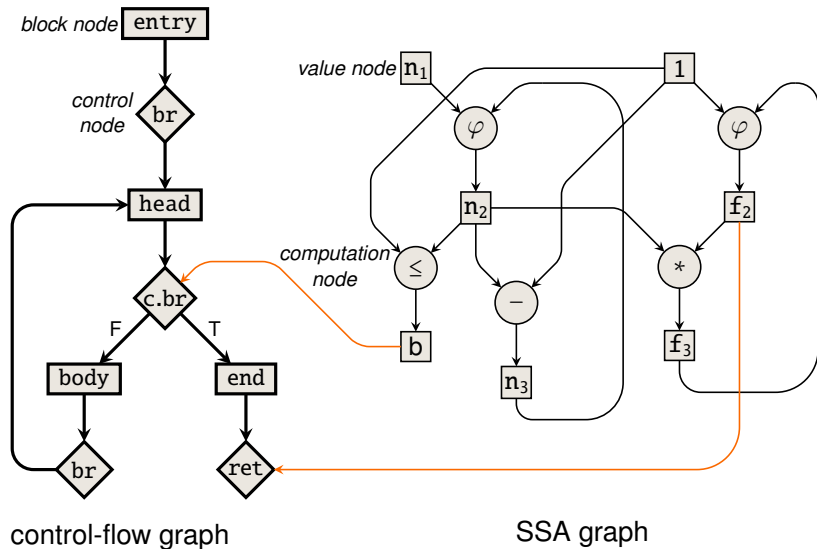


control-flow graph

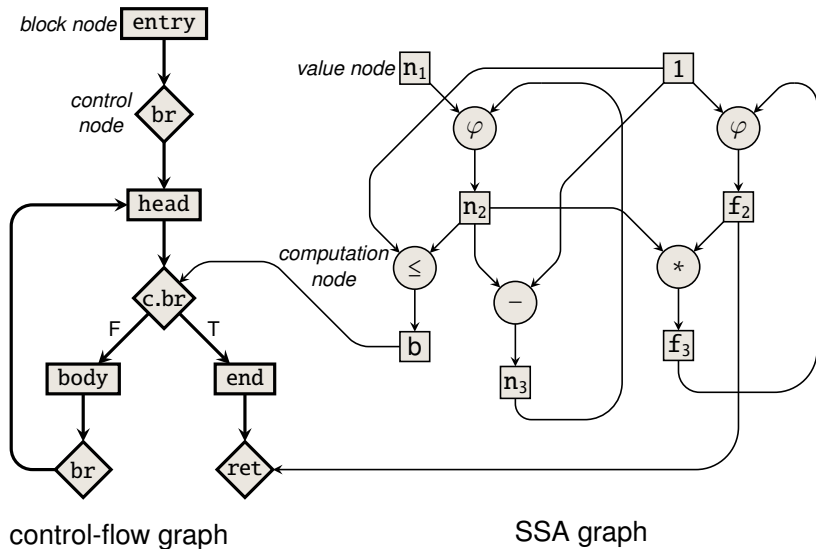


SSA graph

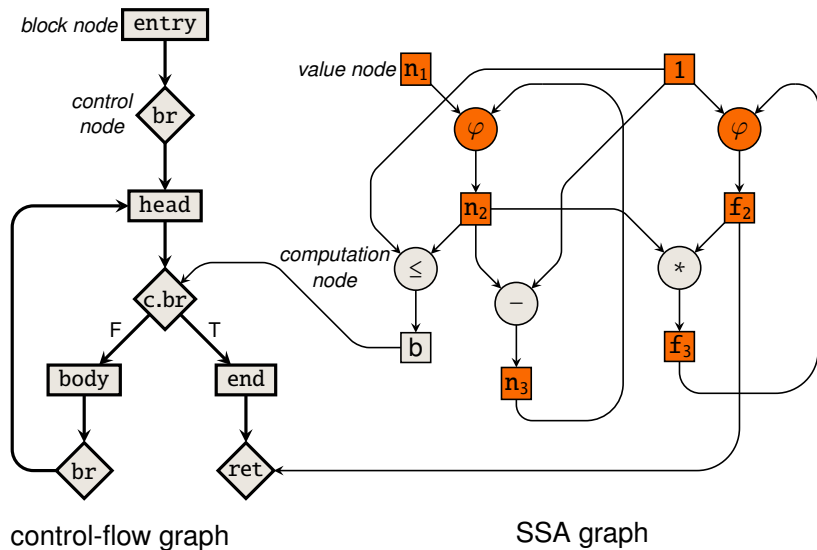
Add Missing Data-Flow Edges



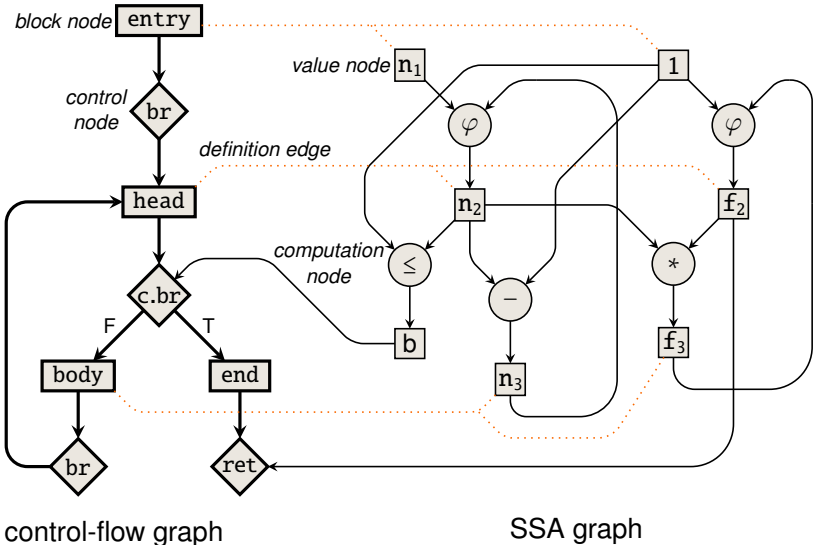
Goal: Prevent Moves That Break Semantics



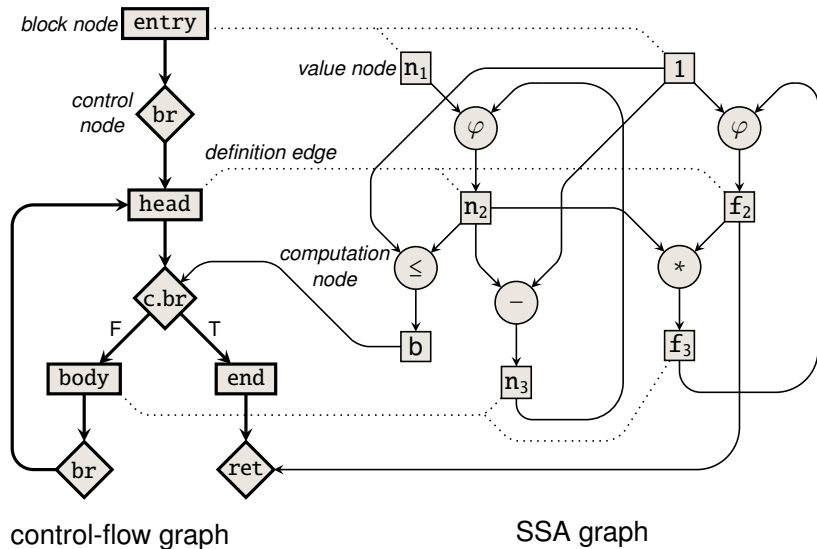
Such Moves Concern Data Used/Defined By φ 's



Definition Edges Prevent Moves



Universal Function (UF) Graph



Memory Operations and Function Calls

(Not in [1])

- May implicitly depend on each other (through external state)
- Moving to another block may break program semantics

Example

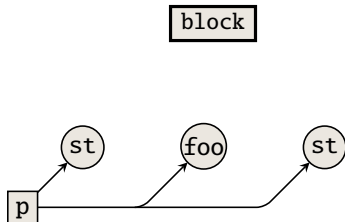
block:

...

store p, ...

call foo, p

store p, ...



Capture Implicit Deps Via State Nodes

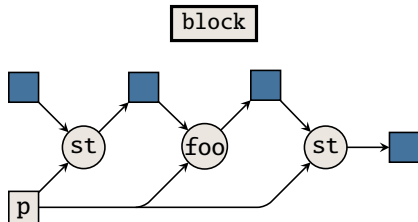
block:

...

store p, ...

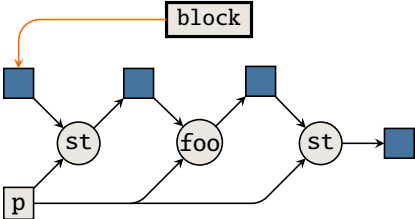
call foo, p

store p, ...



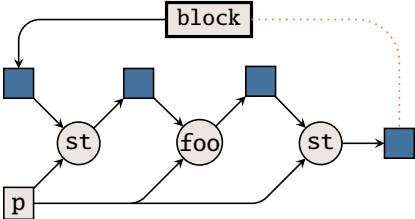
Data-Flow Edge Prevents “Upward” Moves

```
block:  
  ...  
  store p, ...  
  call foo, p  
  store p, ...
```



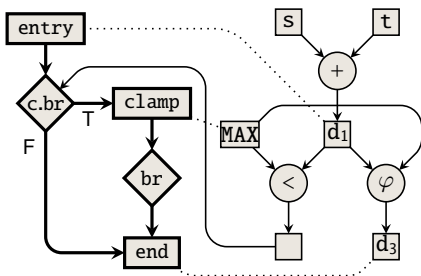
Definition Edge Prevents “Downward” Moves

```
block:  
  ...  
  store p, ...  
  call foo, p  
  store p, ...
```



Instruction Representation

- Apply same construction method as for UF graphs
 - Enables complex instructions to be captured as pattern graphs
 - Example: **satadd** (has both control and data flow)



Other Features (Not in [1])

- Insertion of additional jump instructions when necessary
 - Otherwise leads to model with no solutions
- Reuse of copied values
 - Leads to more efficient code
- Prevention of cyclic data dependencies
 - Otherwise leads to incorrect code

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Variables

- Which matches to select?
- In which blocks to place selected matches?
- In which locations to make values available?
- Which copied value to use?
- In what order to place blocks?

Constraints

Function:

- UF graph must be covered (graph partitioning)
- Values and states must be defined before use
- Placements restricted by definition edges
- ...

Processor:

- Values must be in compatible locations
- Fall-through conditions must be fulfilled

Objective Function

- Minimize execution time

- Typical implementation:

$$\sum_{m \in M} \mathbf{sel}[m] \times \mathit{cost}(m) \times \mathit{freq}(\mathit{blockOf}(m))$$

- Execution frequencies computed statically (by LLVM)
- Apply refined implementation to increase propagation

- [minimize code size, ...]

Techniques to Improve Solving

- Implied and dominance breaking constraints
- Cost bounding
- Presolving

Outline

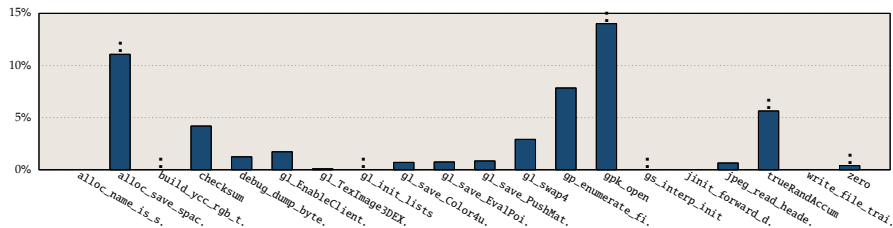
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Setup

- Randomly selected 20 functions from MEDIABENCH using k -means clustering
 - Medium-size functions (50–200 LLVM operations)
 - No scalar or floating-point operations
- Chose HEXAGON 5 as target
 - Rich instruction set
 - Used in many embedded systems
- Found matches using VF2 [3]
 - Pattern graphs can be arbitrarily complex
- Modeled using MINIZINC
- Solved using CHUFFED
- Timed out after 10 minutes
 - No improvements observed after ~5 minutes

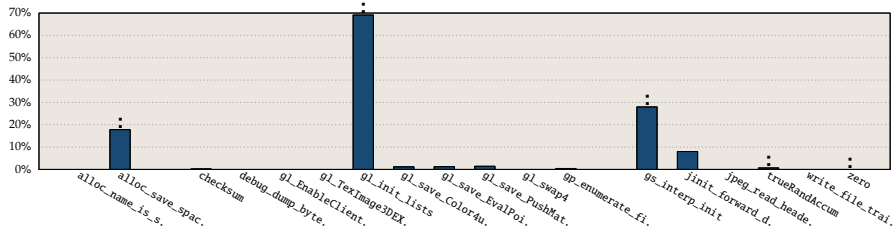
[3] Cordella et al. “An Improved Algorithm for Matching Large Graphs”. In: *Proceedings of GbRPR'01*, pp. 149–159. Springer, 2001.

Our Approach vs LLVM 3.8



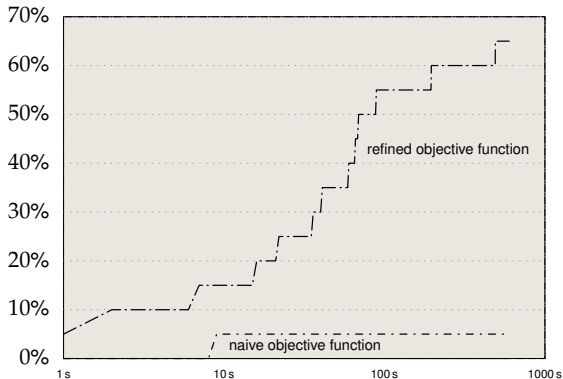
- Compared: estimated speedup
- Baseline: solutions produced by LLVM
- Dots on bars indicate timeouts
- Geometric mean improvement: 2.5%
- Speedups due to global code motion
 - move loading of constants to blocks with lower exec. freq.
 - selection of auto-increment memory instruction and block ordering
 - better sequence led to fewer jump instructions

Value Reuse vs Without



- Compared: estimated speedup
- Baseline: solutions produced without value reuse
- Dots on bars indicate timeouts
- Geometric mean improvement: 5.4%
- Better due to less constant reloading
 - crucial in initialization routines

Refined vs Naive Objective Function



- Compared: cumulative number of optimality proofs
- Refined objective function is essential for scalability

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Future Work

- **Address** model limitations
 - Lacks recomputation – relax *exact* coverage
- **Extend** toolchain to produce executable code
 - Engineering task
- **Integrate** instruction scheduling and register allocation [4]
 - Code generation tasks interact with one another – feasible because constraint models are compositional
- **Make available** on Github as part of Unison
 - <https://github.com/unison-code/unison>

[4] Castañeda Lozano et al. “Combinatorial Spill Code Optimization and Ultimate Coalescing”. In: *Proceedings of LCTES’14*, pp. 23–32. ACM, 2014.

Conclusions

- Made UIS **complete** by:
 - extending it to handle memory operations and function calls
 - introducing methods to insert jump instructions where necessary and forbid cyclic data dependencies
- Made UIS **practical** by:
 - extending constraint model with value reuse to improve code quality
 - introducing solving techniques that increase scalability and robustness
 - demonstrating approach to be competitive with LLVM for up to medium-sized functions
- Showed that **combinatorial optimization** for instruction selection is **well-suited** to exploit modern processors in embedded systems

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- 8. Extra Material**

Constraints: Global Instruction Selection

- Every operation must be covered by exactly one selected match:

$$\mathbf{omatch}[o] = m \Leftrightarrow \mathbf{sel}[m], \forall o \in O, \forall m \in M_o \quad (1)$$

- Every datum must be defined by exactly one selected match:

$$\mathbf{dmatch}[d] = m \Leftrightarrow \mathbf{sel}[m], \forall d \in D, \forall m \in M_d \quad (2)$$

Constraints: Global Code Motion

- Operations covered by the same match must be placed in the same block:

$$\begin{aligned} \mathbf{sel}[m] \Rightarrow \mathbf{oplace}[o_1] = \mathbf{oplace}[o_2], \\ \forall m \in M, \forall o_1, o_2 \in \mathit{covers}(m) \end{aligned} \quad (3)$$

- Matches with an entry block must be placed in the entry block:

$$\begin{aligned} \mathbf{sel}[m] \Rightarrow \mathbf{oplace}[o] = b, \\ \forall m \in M, \forall o \in \mathit{covers}(m), \forall b \in \mathit{entry}(m) \end{aligned} \quad (4)$$

- Data must be defined before use:

$$\begin{aligned} \mathbf{dplace}[d] \in \mathit{dom}(\mathbf{oplace}[o]), \\ \forall m \in M_{\bar{\varphi}}, \forall d \in \mathit{uses}(m), \forall o \in \mathit{covers}(m) \end{aligned} \quad (5)$$

Constraints: Global Code Motion

- Restrictions by the definition edges must be enforced:

$$\mathbf{dplace}[d] = b, \forall \{d, b\} \in DE \quad (6)$$

- Data must be defined in either block wherein the match is placed or in a spanned block:

$$\mathbf{sel}[m] \Rightarrow \mathbf{dplace}[\mathbf{alt}[p]] \in \{\mathbf{oplace}[o]\} \cup \mathit{spans}(m), \quad (7)$$
$$\forall m \in M, \forall p \in \mathit{defines}(m), \forall o \in \mathit{covers}(m)$$

- No data must be placed in a consumed block:

$$\mathbf{sel}[m] \Rightarrow \mathbf{oplace}[o] \neq b, \quad (8)$$
$$\forall o \in O, \forall m \in M, \forall b \in \mathit{consumes}(m)$$

Constraints: Inactive Data

- Data defined by a kill match must be inactive:

$$\begin{aligned} \mathbf{sel}[m] &\Leftrightarrow \mathbf{inactive}[\mathbf{alt}[p]], \\ &\forall m \in M_x, \forall p \in \mathit{defines}(m) \end{aligned} \tag{9}$$

- Data used by non-kill match must be active:

$$\begin{aligned} \mathbf{sel}[m] &\Rightarrow \neg \mathbf{inactive}[\mathbf{alt}[p]], \\ &\forall m \in M_{\bar{x}}, \forall p \in \mathit{uses}(m) \end{aligned} \tag{10}$$

Constraints: Data Copying

- Data locations used and defined by matches must be compatible:

$$\begin{aligned} \mathbf{sel}[m] \Rightarrow \mathbf{loc}[\mathbf{alt}[p]] &\in \mathit{stores}(m, p), \\ \forall m \in M, \forall p \in P \text{ s.t. } \mathit{stores}(m, p) &\neq \emptyset \end{aligned} \quad (11)$$

- Intermediate values must not be reused by other matches:

$$\begin{aligned} \mathbf{sel}[m] \Rightarrow \mathbf{loc}[\mathbf{alt}[p]] &= l_{\text{null}}, \\ \forall m \in M, \forall p \in \mathit{intvalues}(m) \end{aligned} \quad (12)$$

Constraints: Block Ordering

- Blocks must be placed in a sequence:

$$\text{circuit}(\cup_{b \in B} \{\mathbf{succ}[b]\}) \quad (13)$$

-

$$\begin{aligned} \mathbf{succ}[\text{entry}(m)] &= b \vee \\ (\mathbf{succ}[\mathbf{succ}[\text{entry}(m)]] &= b \wedge \text{empty}(\mathbf{succ}[\text{entry}(m)])), \quad (14) \\ \forall \langle m \rangle b \in J, \end{aligned}$$

where

$$\text{empty}(b) \equiv \mathbf{oplace}[o] \neq b \vee \mathbf{omatch}[o] \in M_{\perp}, \forall o \in O$$

Constraints: Cyclic Data Dependencies

- Combinations leading to cyclic data dependencies must be forbidden:

$$\sum_{m \in f} \mathbf{sel}[m] < |f|, \forall f \in F \quad (15)$$

Refined Objective Function

- Construct cost matrix:

$$C = \left[\left\langle o, m, b, \text{freq}(b) \times \text{cost}(m, o) \right\rangle \left| \begin{array}{l} m \in M, \\ o \in \text{covers}(m), \\ b \in B \end{array} \right. \right] \quad (16)$$

- Restrict the cost for each operation:

$$\text{table}(\langle o, \mathbf{omatch}[o], \mathbf{oplace}[o], \mathbf{ocost}[o] \rangle, C), \quad (17)$$

$\forall o \in O$

- Compute total cost:

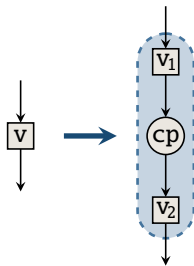
$$\mathbf{cost} = \sum_{o \in O} \mathbf{ocost}[o] \quad (18)$$

Cost Bounding

- Bound total cost:

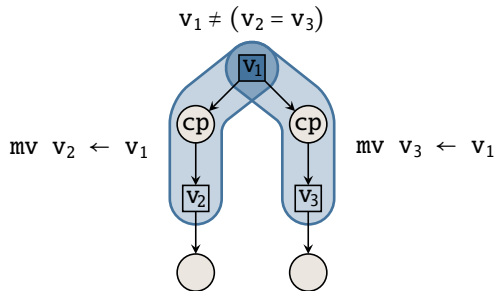
$$C_{\text{relaxed}} \leq \mathbf{cost} < C_{\text{llvm}} \quad (19)$$

Copy Extension

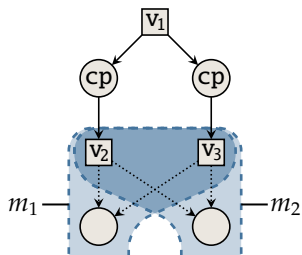


- When locations for v_1 and v_2 can be the same, select special *null-copy pattern* with zero cost
- Otherwise select appropriate copy instruction

May Lead to Redundant Copies

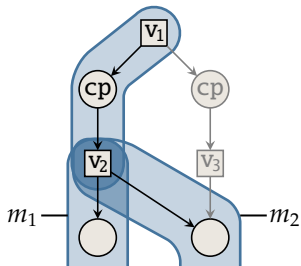


Alternative Values ...



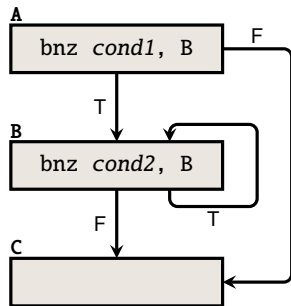
- v_2 and v_3 are *copy-related*
- m_1 and m_2 may use either value

... Enable Value Reuse



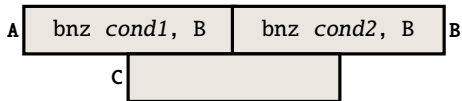
- v_2 and v_3 are *copy-related*
- m_1 and m_2 may use either value

Case Requiring Additional Jump Insertion



- `bnz` falls to next instruction if `cond = F`

As Is: No Valid Order



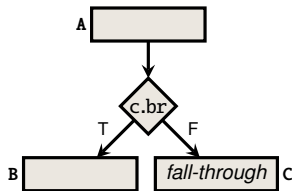
Requires Additional Jump Instruction

A	<code>bnz cond1, B</code> <code>br C</code>
B	<code>bnz cond2, B</code>
C	

B	<code>bnz cond2, B</code> <code>br C</code>
A	<code>bnz cond1, B</code>
C	

Extend Pattern Set With Dual-Target Branch Patterns

For each pattern with fall-through condition:

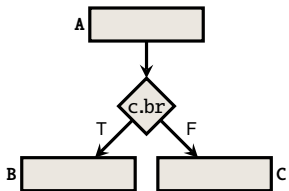


Emit:

bnz cond, B

Cost:

1



Emit:

bnz cond, B

br C

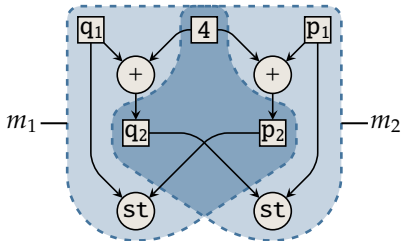
Cost:

$1 + \text{cost}(\text{br})$

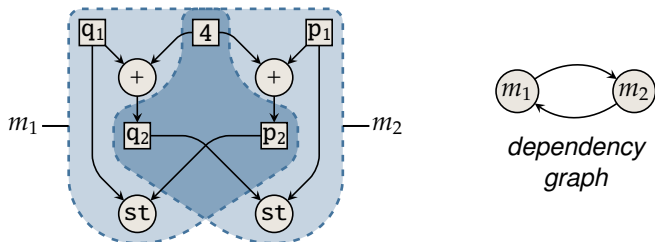
Example at Risk of Cyclic Data Dependency

...

```
p2 = p1 + 4  
store q1, p2  
q2 = q1 + 4  
store p1, q2
```



Forbidding Cyclic Data Dependencies



- For each cycle in dependency graph, not all matches may be selected
- Similar to method used by Ebner et al. [2]

[2] Ebner et al. "Generalized Instruction Selection Using SSA-Graphs." In: *Proceedings of LCTES'08*, pp. 31–40. ACM, 2008.