

Compile-time Optimization of a Constraint-based Compiler Back-end



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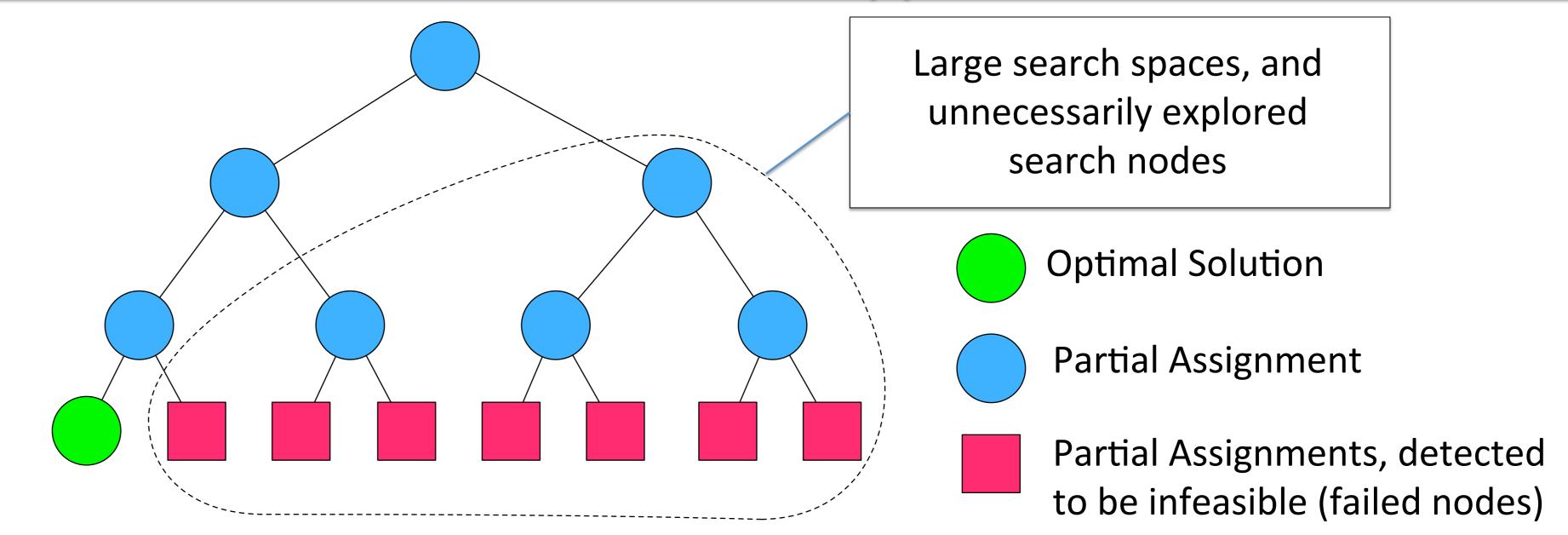
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Generating Optimal Code

compiler

- x combinatorial optimization
- = optimal code generation

The Problem with Combinatorial Approaches: The Search Effort



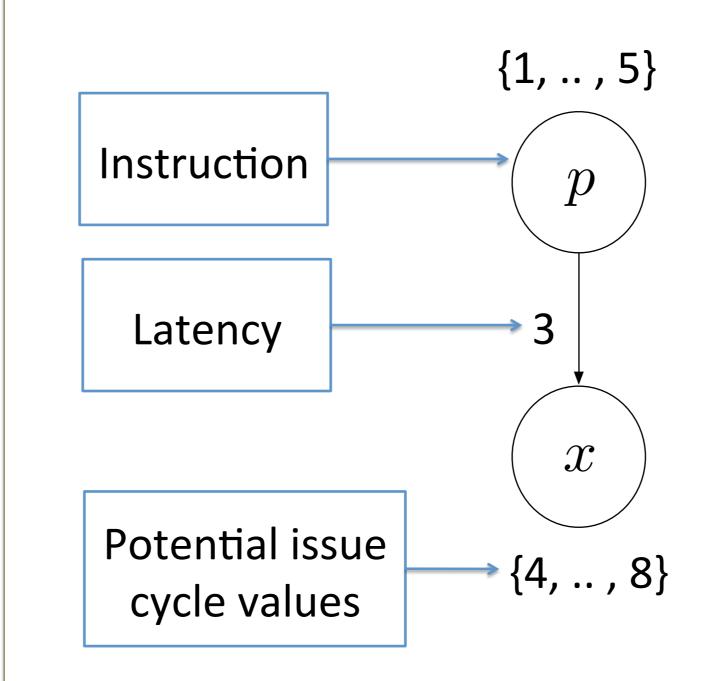
- Combinatorial approaches are often slower due to the large search effort
- For a constraint-based compiler, that would imply long compilation times

The Key to Speeding Up Solution Search

- The Unison project [1] uses Constraint Programming (CP), a combinatorial optimization approach, to implement a constraint-based compiler
- In CP, a problem is modeled with variables and relations among these variables (that is, constraints)
- A constraint may be a base constraint (modeling core problem), or an implied constraint (modeling logically redundant relations)

Implied constraints may provide a different point-ofview on the problem, that may help to skip exploring infeasible assignments at an early stage of search

Base Constraints for Instruction Scheduling



Precedence Constraint

An instruction \boldsymbol{x} may not start execution before its predecessor \boldsymbol{p} was issued and its latency has passed:

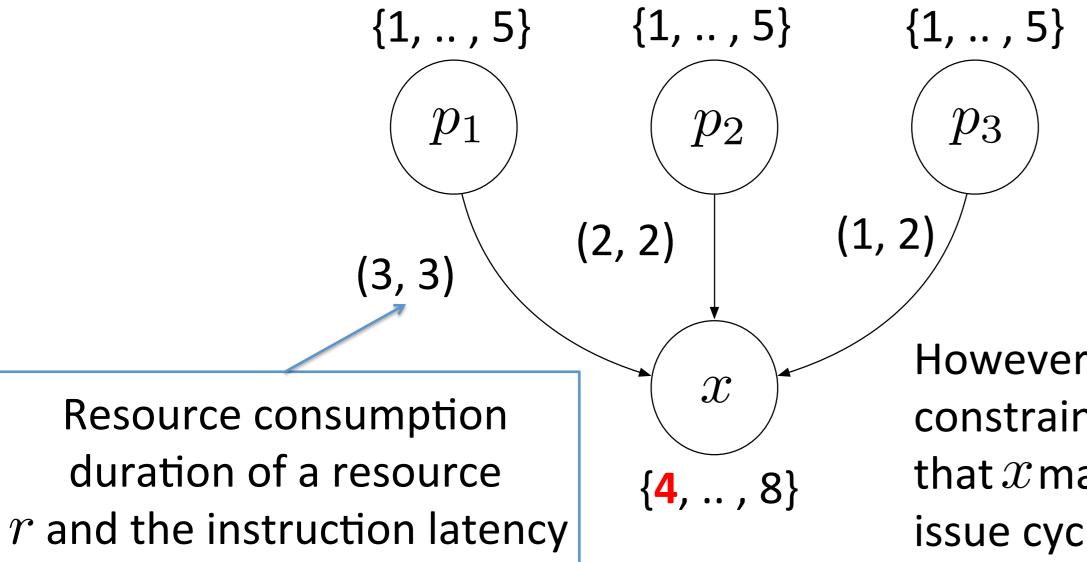
$$c_x \ge c_p + \operatorname{lat}(p)$$

with c_* denoting the issue cycle lat(*) denoting the latency

The precedence constraint holds for the example on the left.

Looking at the bigger picture

Insight: If we take several predecessors into consideration, we find out that x can never start at issue cycle 4: all of its predecessors consume the same unique resource.



However, the precedence constraints do not detect that x may not start at issue cycle 4!

An implied constraint: The Predecessor Constraint

For each instruction i, its predecessor set P and a resource r, add a **predecessor constraint** (extended from [2]):

$$lower(c_i) \ge \min\{lower(c_p) \mid p \in P\}$$

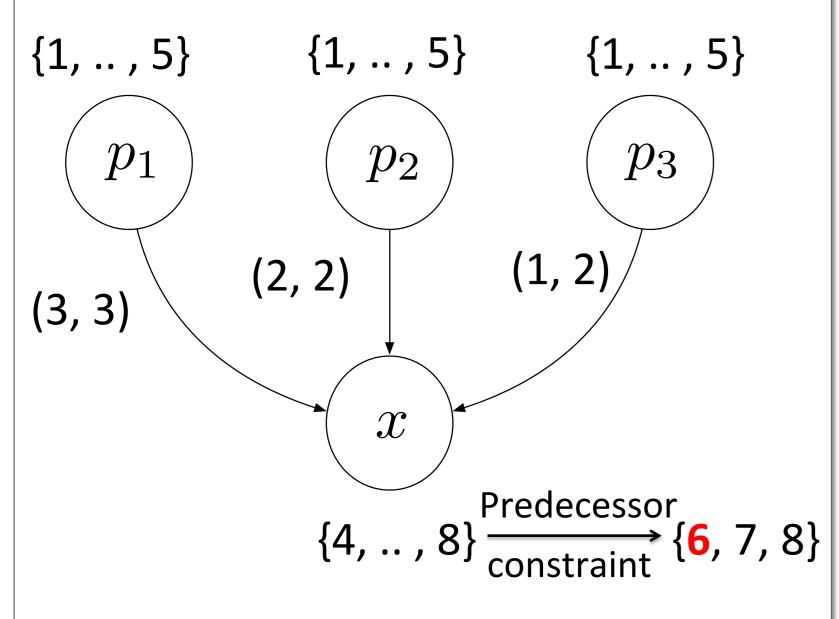
$$+ \left\lceil \frac{\sum_{p \in P} \operatorname{dur}(p,r) * \operatorname{con}(p,r)}{\operatorname{cap}(r)} \right\rceil$$

$$- \max\{\operatorname{dur}(p,r) \mid j \in P\}$$

$$+ \min\{\operatorname{lat}(p) \mid p \in P\}$$

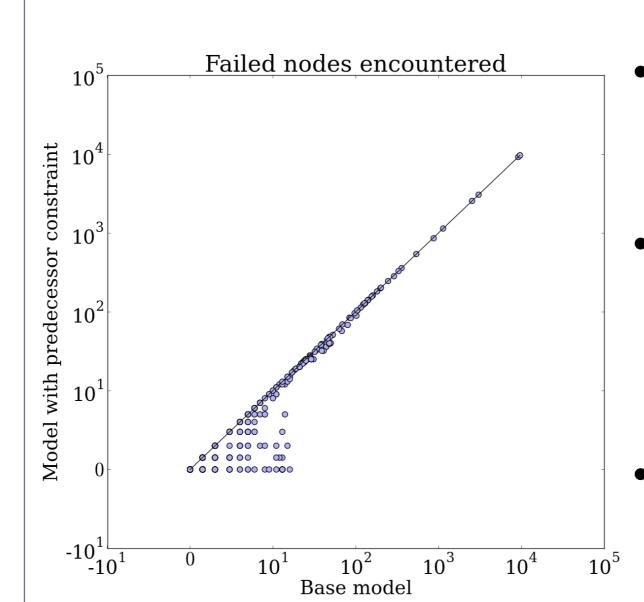
with cap(r) being the capacity of resource r con(p,r) denoting how many units of r are consumed by p dur(p,r) denoting the number of cycles in which p consumes r lower(*) being the lower bound of a domain

Applying the Constraint



Assuming $cap(r) = con(p_*, r) = 1$

Impact of Predecessor Constraints on bzip2



- Figure shows encountered failed nodes for base and extended model during solution search for a basic block
- Extended model with predecessor constraints significantly cuts down the number of failed nodes (in total for 291 basic blocks)
- Achieves to find optimal solutions for four more basic blocks within a time limit of 30 seconds

References

[1] R. Castañeda Lozano, M. Carlsson, G. Hjort Blindell, and C. Schulte. Combinatorial spill code optimization and ultimate coalescing. In Proc. of LCTES'14
[2] A. M. Malik, J. McInnes, and P. van Beek. Optimal basic block instruction scheduling for multipleissue processors using constraint programming. In International Journal on Artificial Intelligence Tools 2008