Automated Code Generation using Dynamic Programming Techniques

Igor Boehm

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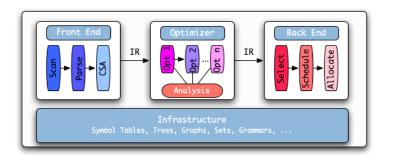


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Where and What is it good for?

Why do we need Code Generation Tools?

Writing a good compiler for a modern programming language is not easy...





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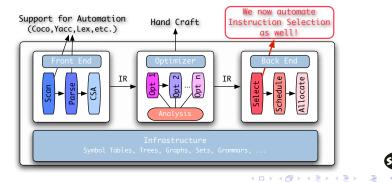
Image: A matrix

Motivation	Theoretical Preliminaries	Tree
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Where and What is it good for?

Tame Complexity through Automation!

 ...so we like to let tools figure out how to do the boringly mechanical parts, and only fill in the relevant bits and pieces.



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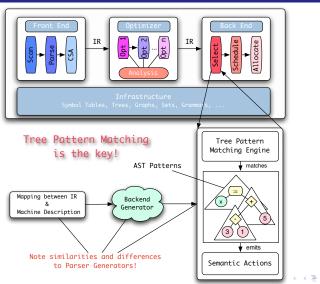
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Where and What is it good for?



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What do we actually want to do?

- Given an IR in the form of an AST:
 - We want the following:
 - obviously we want to generate Machine Code
 - but the generated code should be optimal!
 - concentrate on the essential part of the problem, namely the task of generating code for AST trees and subtrees.

We do not want:

- to clutter our code with boring tree traversals
- to worry about coding the details of optimal instruction selection



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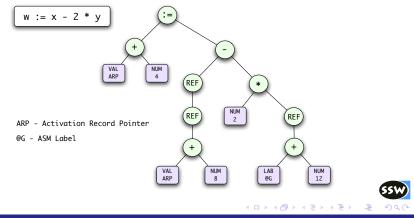
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Code Generation Example

Lets say we have an AST for the following operation and want to emit code for it:



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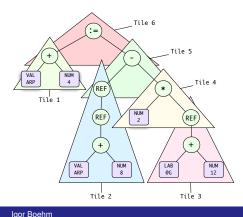
Theoretical Preliminaries

Free Pattern Matching Language

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Code Generation Example

If we were to code this manually, we would probably do something like the following:



- Each tile corresponds to a sequence of operations
- If those operations are emitted in the appropriate order they implement the tree



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So what's hard about this?

- Well, the hard part is to find the optimal set of tiles to tile the tree.
- In order to see why that is a problem we will connect tiles to AST subtrees by:
 - providing a set of rewrite rules
 - associate code templates with rewrite rules



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Code Generation Example

Rewrite Rules with Code Templates for Abstract Syntax Tree

	Rewrite Rule	Code Template
1	$Goal \rightarrow Assign$	
2	$Assign \rightarrow \leftarrow (Reg_1, Reg_2)$	store $r_2 \rightarrow r_1$
3	$Reg \rightarrow LAB_1$	$loadI \ l_1 \rightarrow r_{new}$
4	$Reg \rightarrow VAL_1$	
5	$Reg \rightarrow NUM_1$	load $n_1 \rightarrow r_{new}$
6	$Reg ightarrow REF(Reg_1)$	load $r_1 \rightarrow r_{new}$
7	$Reg \rightarrow REF(+(Reg_1, Reg_2))$	$loadAO \ r_1, r_2 \rightarrow r_{new}$
8	$Reg \rightarrow REF(+(Reg_1, NUM_2))$	$loadAI \ r_1, n_2 \rightarrow r_{new}$
9	$Reg \rightarrow REF(+(LAB_1, Reg_2))$	$loadAI \ r_2, l_1 \rightarrow r_{new}$
10	$Reg ightarrow + (Reg_1, Reg_2))$	add $r_1, r_2 \to r_{new}$
11	$Reg ightarrow + (Reg_1, NUM_2))$	$addI \ r_1, n_2 \rightarrow r_{new}$
12	$Reg \rightarrow +(LAB_1, Reg_2))$	$addI \ r_2, l_1 \rightarrow r_{new}$

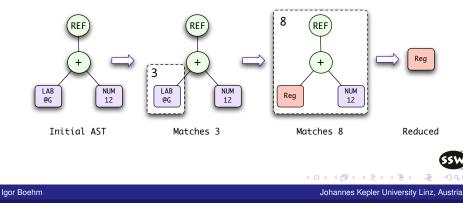
Table: Modified example from [Cooper & Torczon p.561]



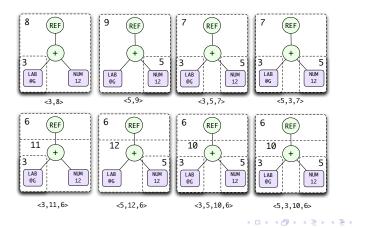
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 Given the previous rewrite rules for our AST, lets consider Tile 3 from our example and try to tile it:



But there is a plethora of rewrite sequences (or tilings) for this trivial subtree!!!





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Code Generation Example

How do we select the optimal sequence of rewrite rules?

- We need some metric which enables us to compare various selections of rewrite rules with each other.
 - Solution: Annotate rewrite rules with costs!
- We need some clever algorithm to sort out the optimal rewrite sequence.
 - Solution: Dynamic Programming will do the trick!



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Dynamic Programming

- Dynamic Programming is a method of solving problems exhibiting the following properties:
 - the approach for a given problem assumes a recursive solution, with a bottom-up evaluation of the solution.
 - sub-solutions can be recorded (e.g. in a table) for *reuse*.



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Dynamic Programming

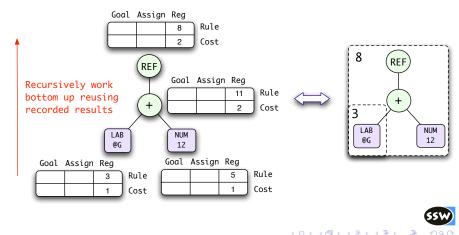
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Dynamic Programming

One optimal solution for our previous tiling problem looks as follows:



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Dynamic Programming

So we are almost there!

- We know how to optimally tile an AST using Dynamic Programming.
- We know how to specify rewrite rules (also called tree patterns or productions).
- We know how to specify code templates (semantic actions) for AST patterns.



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Tree Pattern Matching using Dynamic Programming works as follows:

- Two passes over the AST:
 - Pass 1: Finds the optimal tiling of an AST using Dynamic Programming.
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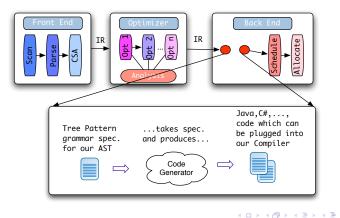


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Tree Pattern Matching using Dynamic Programming

Let's look at where our code generator is used in the compiler design process:





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Language Specification

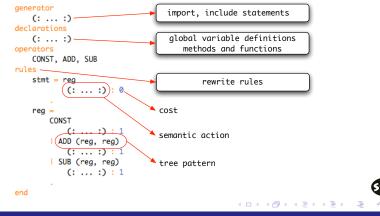
Theoretical Preliminaries

Tree Pattern Matching Language

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Tree Pattern Matching Language

General Structure of a Grammar Specification:

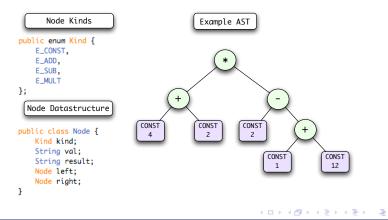


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Language Specification

We have the following Node definition together with an example Abstract Syntax Tree:



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Language Specification				
Example 1	instrlist.d :) : 1 ADD al (reg rl, reg r2) (: al.result = instrlist.d :) : 1 SUB sl (reg rl, reg r2) (: sl.result =	w LinkedList(); :)		
	chu		urr 3 e r 3 e r 1 e	E FURCE

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Language Specification

- While the previous example looks nice at first sight, it is still quite cumbersome:
 - Problem 1: The way in which result registers are returned is by abusing the node class!
 - **Problem 2:** The list holding instructions is global!
- We can do much better than this!!!



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Language Specification

We can use the Attribute Grammar formalism to correct our previous 'flaws':

- Solution to Problem 1: The register production produces a register as its output and takes a list of instructions as its input to append its instructions to.
- Solution to Problem 2: The instruction list is passed as a parameter and thus must not be declared as a 'global' class variable.
- Let's look at an example for clarification...



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Language Specification		
Example 2	<pre>operators Operators (ou CONST: ELCONST:, ANDO:ELADO SUB(:E_SUB:), MULT(:E_MULT:) rules Production rules for regi @out: produce a resu @in: list to add ins reg <: out String reg, List CONST c1 (: reg</pre>	<pre>ist; pritions (not needed nom) r Wode Kinds) sters: lt register twoctions to instr :> r egg.getNextReg(C); r.add('loadI " + c1.val + "," + reg); reg. getNextReg(C); .add('ad " + r2 *," + reg);</pre>

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- The simple example used up until now doesn't really convey the full power of the Tree Pattern Matching Language, namely:
 - Arbitrary nesting of patterns.
 - ▶ The ability to *sprinkle* semantic actions almost anywhere.
- ...those features will be revealed during my last presentation where I will demonstrate a complete working example.



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- An extensive prototyping phase helped to identify the core features of the tree pattern matching language by trying to solve real world problems with it.
- Before any implementation started a semantics for the language has been specified in terms of a denotational semantics and semantic algebras.
- A fully functional Lexer and Parser has already been implemented.



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- Implement good context sensitive analysis to ease development of Tree Pattern Matching grammar specifications.
- Finally, emit code for the given specifications!



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Thank you for your attention!



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