
Topic 12: Acyclic Instruction Scheduling

COS 320

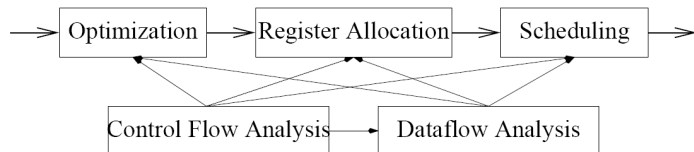
Compiling Techniques

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The Back End



The Back End:

1. Maps infinite number of virtual registers to finite number of real registers → *register allocation*
2. Removes inefficiencies introduced by front-end → *optimizer*
3. Removes inefficiencies introduced by programmer → *optimizer*
4. Adjusts pseudo-assembly composition and order to match target machine → *scheduler*

Scheduling

Multiply instruction takes 2 cycles...

```
1  r1 = r0 + 0
2  r2 = M[FP + A]
3  r3 = r0 + 4
4  r4 = M[FP + X]
```

LOOP:

```
1  r5 = r3 * r1
2
3  r5 = r2 + r5
4  M[r5] = r4
5  r1 = r1 + 1
6  BR r1 <= 10, LOOP
```

```
1  r1 = r0 + 0
2  r2 = M[FP + A]
3  r3 = r0 + 4
4  r4 = M[FP + X]
```

LOOP:

```
1  r5 = r3 * r1
2  r1 = r1 + 1
3  r5 = r2 + r5
4  M[r5] = r4
5  BR r1 <= 10, LOOP
```

Scheduling

Multiply instruction takes 2 cycles...

Machine executes 2 instructions per cycle...

```
1  r1 = r0 + 0
2  r2 = M[FP + A]
3  r3 = r0 + 4
4  r4 = M[FP + X]

1  r1 = r0 + 0    r2 = M[FP + A]
2  r3 = r0 + 4    r4 = M[FP + X]

LOOP:
1  r5 = r3 * r1
2  r1 = r1 + 1
3  r5 = r2 + r5
4  M[r5] = r4
5  BR r1 <= 10, LOOP

LOOP:
1  r5 = r3 * r1    r1 = r1 + 1
2
3  r5 = r2 + r5
4  M[r5] = r4      BR r1 <= 10, LOOP
```

Instruction Level Parallelism

- Instruction-Level Parallelism (ILP), the concurrent execution of independent assembly instructions.
- ILP is a cost effective way to extract performance from programs.
- Exploiting ILP requires global optimization and scheduling.
- Processors are becoming increasingly dependent on the ability of compilers to expose ILP.
 - Current state-of-the-art machines can execute 3 to 6 instructions per cycle if available. (i.e. Pentium III, DEC Alpha 21264)
 - Some processors rely on compiler for guidance. (i.e. Itanium)
- Current state-of-the-art compilers cannot expose this level of ILP in integer programs.

Data Dependence

- A *data dependence* is a constraint on scheduling arising from the flow of data between two instructions. Types:
 - RAW: An instruction u is *flow-dependent* on a preceding instruction d if u consumes a value computed by d .
 - WAR: An instruction d is *anti-dependent* on a preceding instruction u if d writes to a location read by u .
 - WAW: An instruction d_2 is *output-dependent* on a preceding instruction d_1 if d_1 writes to a location also written by d_2 .
- Types of data:
 - Register dependence
 - Memory dependence

Data Dependence

```
r1 = r2 + r3
```

```
Branch r1 <= 10, TRUE
```

```
r4 = r2 * r5
```

```
r5 = r4 + 1
```

TRUE:

```
r4 = r5 - 1
```

False Dependence

Eliminate WAW dependences

```
r1 =
```

```
branch
```

```
r1 =  
  = r1
```

Eliminate WAR dependences

```
  = r1  
r1 =  
  = r1
```

- Eliminate RAW dependences?
- Register allocation vs. splitting live ranges

Control Dependence

- A *control dependence* is a constraint on scheduling arising from the control flow of the program.

```
Branch r1 <= 10, TARGET1
```

```
Branch r2 <= 10, TARGET2
```

```
r4 = r3 + 5
```

TARGET1:

```
r5 = r4 - 1
```

TARGET2: (Assume: r4 not live here)

Control Dependences

Sources of Control Dependence

- Liveness
- Side-effects
 - Potentially Excepting Instructions (PEIs)
 - Memory Writes
 - Input/Output

Dependences

Latency

- Amount of time after the execution of an instruction that its result is ready.
- An instruction can have more than one latency!

Data Dependence Graph

- A *data dependence graph* consists of instructions and a set of directed data dependence edges among them in which each edge is labeled with its latency and type of dependence.
- Scheduling (code motion) must respect dependence graph.

Resources

- What does “two instructions per cycle” mean?
- *Resource* - A function of the processor that can be used by only one instruction at a time.
- Examples:
 - Fetch units
 - Decode units
 - Execution units
 - Register ports

Pipelining

Resource Map

Scheduling

- The goal of *scheduling* is to construct a sort of the dependence graph that:
 - Produces the same result - respects dependences
 - Minimizes execution time - makes maximal use of machine resources
- Scheduling is NP-hard even with simple formulation of problem.
- Use Heuristics to approximate solution.
- In practice, is exhaustive search of all schedules practical in most cases?

Heuristic: List Scheduling

- List scheduling, the most common heuristic, is $O(n_2)$.
- Create *ready queue* to hold *ready* instructions.
- An instruction is *ready* when all incoming dependences are satisfied.
- A dependence is satisfied when source of dependence has been scheduled at least latency cycles earlier.

List Scheduling

```
build dependence graph
insert instructions with no incoming dependences into ready queue
WHILE (instruction are not scheduled) DO
    current_cycle_sched = FALSE
    FOREACH instruction i in ready queue DO
        IF (resources exist to schedule i in cycle) THEN
            schedule i, update ready queue
            current_cycle_sched = TRUE
    IF (NOT current_cycle_sched) THEN
        cycle++
        update ready queue
```

List Scheduling

```
    LOOP:
1   r5  =  r3  *  r1

2   r1  =  r1  +  1

3   r5  =  r2  +  r5

4   M[r5] = r4

5   BR r1  <=  10, LOOP
```

Hardware Scheduling

Machines can also do scheduling...

- hardware schedulers process code after it has been fetched
- hardware finds independent instructions
- works with legacy architectures (found in x86 & Pentium)
- program knowledge more precise at run-time - memory dependence

But compiler still important.

- Hardware schedulers have a small window.
- Hardware complexity increases.
- Hardware does not benefit directly from compiler optimization.

Expression Reformulation

Loop Unrolling

```
sum = 0;
for i = 1 to 30:
    sum = sum + A[i];
```

```
0    r1 = 0                r2 = 0
```

```
Loop:
```

```
0    r3 = M[r1 + A]        r1 = r1 + 1
```

```
1
```

```
2    r2 = r2 + r3          BR r1 < 30, Loop
```

Renaming

```
0    r1 = 0                r2 = 0
```

```
Loop:
```

```
0    r3 = M[r1 + A]        r1 = r1 + 1
```

```
1
```

```
2    r2 = r2 + r3
```

```
3    r3 = M[r1 + A]        r1 = r1 + 1
```

```
4
```

```
5    r2 = r2 + r3          BR r1 < 30, Loop
```

Accumulator Expansion

```
0    r1 = 0                r2 = 0
```

```
Loop:
```

```
0    r3 = M[r1 + A]        r1 = r1 + 1
```

```
1    r4 = M[r1 + A]        r1 = r1 + 1
```

```
2    r2 = r2 + r3
```

```
3    r2 = r2 + r4          BR r1 < 30, Loop
```


Accumulator Expansion

```
0   r1 = 0           r2 = 0

Loop:

0   r3 = M[r1 + A]    r1 = r1 + 1
1   r4 = M[r1 + A]    r1 = r1 + 1
2   r5 = M[r1 + A]    r1 = r1 + 1    r2 = r2 + r3
3   r2 = r2 + r4
4   r2 = r2 + r5      BR r1 < 30, Loop
```

Induction Variable Elimination

```
0   r1 = 0           r23 = 0
1   r24 = 0          r25 = 0

Loop:

0   r3 = M[r1 + A]    r1 = r1 + 1
1   r4 = M[r1 + A]    r1 = r1 + 1
2   r5 = M[r1 + A]    r1 = r1 + 1    r23 = r23 + r3
3   r24 = r24 + r4
5   r25 = r25 + r5    BR r1 < 30, Loop

0   r2 = r23 + r24
1   r2 = r2 + r25
```

Loop Unrolling and Optimization

```
0   r13 = 0           r14 = 1
1   r15 = 2           r23 = 0
2   r24 = 0           r25 = 0

Loop:

0   r3 = M[r13 + A]    r13 = r13 + 3    r4 = M[r14 + A]
    r14 = r14 + 3      r5 = M[r15 + A]    r15 = r15 + 3

1

2   r23 = r23 + r3     r24 = r24 + r4    r25 = r25 + r5
    BR r13 < 30, Loop

0   r2 = r23 + r24
1   r2 = r2 + r25
```

