Technique	Reduces	Section
Forwarding and bypassing	Potential data hazard stalls	C.2
Delayed branches and simple branch scheduling	Control hazard stalls	C.2
Basic compiler pipeline scheduling	Data hazard stalls	C.2, 3.2
Basic dynamic scheduling (scoreboarding)	Data hazard stalls from true dependences	C.7
Loop unrolling	Control hazard stalls	3.2
Branch prediction	Control stalls	3.3
Dynamic scheduling with renaming	Stalls from data hazards, output dependences, and antidependences	3.4
Hardware speculation	Data hazard and control hazard stalls	3.6
Dynamic memory disambiguation	Data hazard stalls with memory	3.6
Issuing multiple instructions per cycle	Ideal CPI	3.7, 3.8
Compiler dependence analysis, software pipelining, trace scheduling	Ideal CPI, data hazard stalls	H.2, H.3
Hardware support for compiler speculation	Ideal CPI, data hazard stalls, branch hazard stalls	H.4, H.5

Figure 3.1 The major techniques examined in Appendix C, Chapter 3, and Appendix H are shown together with the component of the CPI equation that the technique affects.

The *ideal pipeline CPI* is a measure of the maximum performance attainable by the implementation. By reducing each of the terms of the right-hand side, we decrease the overall pipeline CPI or, alternatively, increase the IPC (instructions per clock). The equation above allows us to characterize various techniques by what component of the overall CPI a technique reduces. Figure 3.1 shows the techniques we examine in this chapter and in Appendix H, as well as the topics covered in the introductory material in Appendix C. In this chapter, we will see that the techniques we introduce to decrease the ideal pipeline CPI can increase the importance of dealing with hazards.

### What Is Instruction-Level Parallelism?

All the techniques in this chapter exploit parallelism among instructions. The amount of parallelism available within a *basic block*—a straight-line code sequence with no branches in except to the entry and no branches out except at the exit—is quite small. For typical MIPS programs, the average dynamic branch frequency is often between 15% and 25%, meaning that between three and six instructions execute between a pair of branches. Since these instructions are likely to depend upon one another, the amount of overlap we can exploit within a basic block is likely to be less than the average basic block size. To obtain substantial performance enhancements, we must exploit ILP across multiple basic blocks.

The simplest and most common way to increase the ILP is to exploit parallelism among iterations of a loop. This type of parallelism is often called *loop-level parallelism*. Here is a simple example of a loop that adds two 1000-element arrays and is completely parallel:

```
for (i=0; i<=999; i=i+1)
x[i] = x[i] + y[i];
```

Every iteration of the loop can overlap with any other iteration, although within each loop iteration there is little or no opportunity for overlap.

We will examine a number of techniques for converting such loop-level parallelism into instruction-level parallelism. Basically, such techniques work by unrolling the loop either statically by the compiler (as in the next section) or dynamically by the hardware (as in Sections 3.5 and 3.6).

An important alternative method for exploiting loop-level parallelism is the use of SIMD in both vector processors and Graphics Processing Units (GPUs), both of which are covered in Chapter 4. A SIMD instruction exploits data-level parallelism by operating on a small to moderate number of data items in parallel (typically two to eight). A vector instruction exploits data-level parallelism by operating on many data items in parallel using both parallel execution units and a deep pipeline. For example, the above code sequence, which in simple form requires seven instructions per iteration (two loads, an add, a store, two address updates, and a branch) for a total of 7000 instructions, might execute in one-quarter as many instructions in some SIMD architecture where four data items are processed per instruction. On some vector processors, this sequence might take only four instructions: two instructions to load the vectors x and y from memory, one instruction to add the two vectors, and an instruction to store back the result vector. Of course, these instructions would be pipelined and have relatively long latencies, but these latencies may be overlapped.

# **Data Dependences and Hazards**

Determining how one instruction depends on another is critical to determining how much parallelism exists in a program and how that parallelism can be exploited. In particular, to exploit instruction-level parallelism we must determine which instructions can be executed in parallel. If two instructions are *parallel*, they can execute simultaneously in a pipeline of arbitrary depth without causing any stalls, assuming the pipeline has sufficient resources (and hence no structural hazards exist). If two instructions are dependent, they are not parallel and must be executed in order, although they may often be partially overlapped. The key in both cases is to determine whether an instruction is dependent on another instruction.

### Data Dependences

There are three different types of dependences: *data dependences* (also called true data dependences), *name dependences*, and *control dependences*. An instruction *j* is *data dependent* on instruction *i* if either of the following holds:

- Instruction i produces a result that may be used by instruction j.
- Instruction j is data dependent on instruction k, and instruction k is data dependent on instruction i.

The second condition simply states that one instruction is dependent on another if there exists a chain of dependences of the first type between the two instructions. This dependence chain can be as long as the entire program. Note that a dependence within a single instruction (such as ADDD R1,R1,R1) is not considered a dependence.

For example, consider the following MIPS code sequence that increments a vector of values in memory (starting at 0 (R1) and with the last element at 8 (R2)) by a scalar in register F2. (For simplicity, throughout this chapter, our examples ignore the effects of delayed branches.)

```
Loop: L.D F0,0(R1) ;F0=array element
ADD.D F4,F0,F2 ;add scalar in F2
S.D F4,0(R1) ;store result
DADDUI R1,R1,#-8 ;decrement pointer 8 bytes
BNE R1,R2,L00P ;branch R1!=R2
```

The data dependences in this code sequence involve both floating-point data:

and integer data:

In both of the above dependent sequences, as shown by the arrows, each instruction depends on the previous one. The arrows here and in following examples show the order that must be preserved for correct execution. The arrow points from an instruction that must precede the instruction that the arrowhead points to.

If two instructions are data dependent, they must execute in order and cannot execute simultaneously or be completely overlapped. The dependence implies that there would be a chain of one or more data hazards between the two instructions. (See Appendix C for a brief description of data hazards, which we will define precisely in a few pages.) Executing the instructions simultaneously will cause a processor with pipeline interlocks (and a pipeline depth longer than the distance between the instructions in cycles) to detect a hazard and stall, thereby reducing or eliminating the overlap. In a processor without interlocks that relies on compiler scheduling, the compiler cannot schedule dependent instructions in such a way that they completely overlap, since the program will not execute correctly. The presence of a data dependence in an instruction sequence reflects a data dependence in the source code from which the instruction sequence was generated. The effect of the original data dependence must be preserved.

Dependences are a property of *programs*. Whether a given dependence results in an actual hazard being detected and whether that hazard actually causes a stall are properties of the *pipeline organization*. This difference is critical to understanding how instruction-level parallelism can be exploited.

A data dependence conveys three things: (1) the possibility of a hazard, (2) the order in which results must be calculated, and (3) an upper bound on how much parallelism can possibly be exploited. Such limits are explored in Section 3.10 and in Appendix H in more detail.

Since a data dependence can limit the amount of instruction-level parallelism we can exploit, a major focus of this chapter is overcoming these limitations. A dependence can be overcome in two different ways: (1) maintaining the dependence but avoiding a hazard, and (2) eliminating a dependence by transforming the code. Scheduling the code is the primary method used to avoid a hazard without altering a dependence, and such scheduling can be done both by the compiler and by the hardware.

A data value may flow between instructions either through registers or through memory locations. When the data flow occurs in a register, detecting the dependence is straightforward since the register names are fixed in the instructions, although it gets more complicated when branches intervene and correctness concerns force a compiler or hardware to be conservative.

Dependences that flow through memory locations are more difficult to detect, since two addresses may refer to the same location but look different: For example, 100(R4) and 20(R6) may be identical memory addresses. In addition, the effective address of a load or store may change from one execution of the instruction to another (so that 20(R4) and 20(R4) may be different), further complicating the detection of a dependence.

In this chapter, we examine hardware for detecting data dependences that involve memory locations, but we will see that these techniques also have limitations. The compiler techniques for detecting such dependences are critical in uncovering loop-level parallelism.

#### Name Dependences

The second type of dependence is a *name dependence*. A name dependence occurs when two instructions use the same register or memory location, called a *name*, but there is no flow of data between the instructions associated with that name. There are two types of name dependences between an instruction i that precedes instruction j in program order:

- 1. An *antidependence* between instruction *i* and instruction *j* occurs when instruction *j* writes a register or memory location that instruction *i* reads. The original ordering must be preserved to ensure that *i* reads the correct value. In the example on page 151, there is an antidependence between S.D and DADDIU on register R1.
- 2. An *output dependence* occurs when instruction i and instruction j write the same register or memory location. The ordering between the instructions

must be preserved to ensure that the value finally written corresponds to instruction *j*.

Both antidependences and output dependences are name dependences, as opposed to true data dependences, since there is no value being transmitted between the instructions. Because a name dependence is not a true dependence, instructions involved in a name dependence can execute simultaneously or be reordered, if the name (register number or memory location) used in the instructions is changed so the instructions do not conflict.

This renaming can be more easily done for register operands, where it is called *register renaming*. Register renaming can be done either statically by a compiler or dynamically by the hardware. Before describing dependences arising from branches, let's examine the relationship between dependences and pipeline data hazards.

#### Data Hazards

A hazard exists whenever there is a name or data dependence between instructions, and they are close enough that the overlap during execution would change the order of access to the operand involved in the dependence. Because of the dependence, we must preserve what is called *program order*—that is, the order that the instructions would execute in if executed sequentially one at a time as determined by the original source program. The goal of both our software and hardware techniques is to exploit parallelism by preserving program order *only where it affects the outcome of the program*. Detecting and avoiding hazards ensures that necessary program order is preserved.

Data hazards, which are informally described in Appendix C, may be classified as one of three types, depending on the order of read and write accesses in the instructions. By convention, the hazards are named by the ordering in the program that must be preserved by the pipeline. Consider two instructions i and j, with i preceding j in program order. The possible data hazards are

- RAW (*read after write*)—*j* tries to read a source before *i* writes it, so *j* incorrectly gets the *old* value. This hazard is the most common type and corresponds to a true data dependence. Program order must be preserved to ensure that *j* receives the value from *i*.
- WAW (write after write)—j tries to write an operand before it is written by i. The writes end up being performed in the wrong order, leaving the value written by i rather than the value written by j in the destination. This hazard corresponds to an output dependence. WAW hazards are present only in pipelines that write in more than one pipe stage or allow an instruction to proceed even when a previous instruction is stalled.
- WAR (write after read)—j tries to write a destination before it is read by i, so i incorrectly gets the new value. This hazard arises from an antidependence (or name dependence). WAR hazards cannot occur in most static issue pipelines—even deeper pipelines or floating-point pipelines—because all reads are early

(in ID in the pipeline in Appendix C) and all writes are late (in WB in the pipeline in Appendix C). A WAR hazard occurs either when there are some instructions that write results early in the instruction pipeline *and* other instructions that read a source late in the pipeline, or when instructions are reordered, as we will see in this chapter.

Note that the RAR (read after read) case is not a hazard.

## **Control Dependences**

The last type of dependence is a *control dependence*. A control dependence determines the ordering of an instruction, *i*, with respect to a branch instruction so that instruction *i* is executed in correct program order and only when it should be. Every instruction, except for those in the first basic block of the program, is control dependent on some set of branches, and, in general, these control dependences must be preserved to preserve program order. One of the simplest examples of a control dependence is the dependence of the statements in the "then" part of an if statement on the branch. For example, in the code segment

- S1 is control dependent on p1, and S2 is control dependent on p2 but not on p1. In general, two constraints are imposed by control dependences:
- 1. An instruction that is control dependent on a branch cannot be moved *before* the branch so that its execution *is no longer controlled* by the branch. For example, we cannot take an instruction from the then portion of an if statement and move it before the if statement.
- An instruction that is not control dependent on a branch cannot be moved after the branch so that its execution is controlled by the branch. For example, we cannot take a statement before the if statement and move it into the then portion.

When processors preserve strict program order, they ensure that control dependences are also preserved. We may be willing to execute instructions that should not have been executed, however, thereby violating the control dependences, *if* we can do so without affecting the correctness of the program. Thus, control dependence is not the critical property that must be preserved. Instead, the two properties critical to program correctness—and normally preserved by maintaining both data and control dependences—are the *exception behavior* and the *data flow*.

Preserving the exception behavior means that any changes in the ordering of instruction execution must not change how exceptions are raised in the program.