Chapter 3

Arithmetic for Computers

Arithmetic for Computers

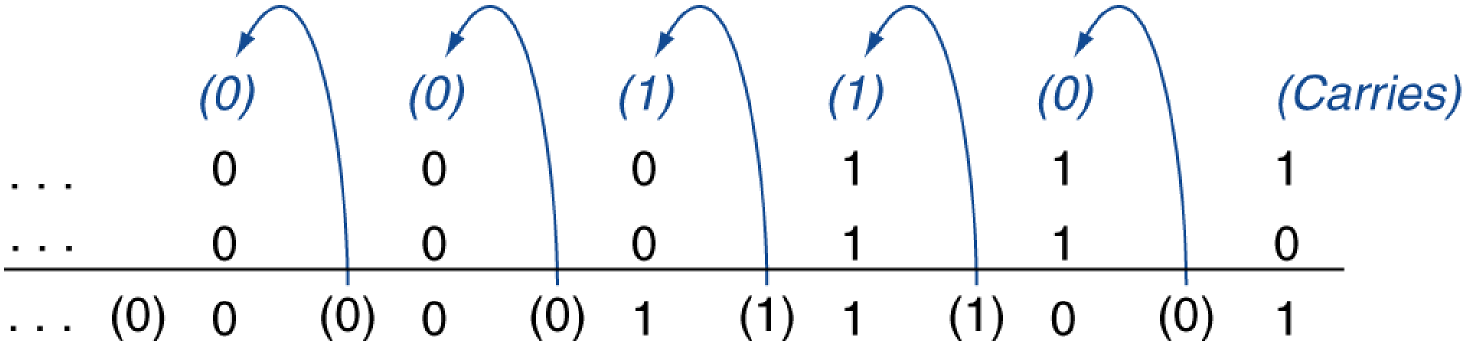
§3.1 Introduction

- Operations on integers
  - Addition and subtraction
  - Multiplication and division
  - Dealing with overflow
- Floating-point real numbers
  - Representation and operations

Integer Addition

§3.2 Addition and Subtraction

- Example: 7 + 6



```
      (0)     (0)     (1)     (1)     (0)     (Carries)
. . .    0       0       0       1       1       1
. . .    0       0       0       1       1       0
. . . (0) 0   (0) 0   (0) 1   (1) 1   (1) 0   (0) 1
```

- Overflow if result out of range
  - Adding +ve and –ve operands, no overflow
  - Adding two +ve operands
    - Overflow if result sign is 1
  - Adding two –ve operands
    - Overflow if result sign is 0

Integer Subtraction

- Add negation of second operand
- Example: 7 – 6 = 7 + (–6)

+7:	0000 0000 … 0000 0111
–6:	1111 1111 … 1111 1010
+1:	0000 0000 … 0000 0001

- Overflow if result out of range
  - Subtracting two +ve or two –ve operands, no overflow
  - Subtracting +ve from –ve operand
    - Overflow if result sign is 0
  - Subtracting –ve from +ve operand
    - Overflow if result sign is 1

Arithmetic for Multimedia

- Graphics and media processing operates on vectors of 8-bit and 16-bit data
  - Use 64-bit adder, with partitioned carry chain
    - Operate on 8×8-bit, 4×16-bit, or 2×32-bit vectors
  - SIMD (single-instruction, multiple-data)
- Saturating operations
  - On overflow, result is largest representable value
    - c.f. 2s-complement modulo arithmetic
  - E.g., clipping in audio, saturation in video

§3.5 Floating Point

Floating Point

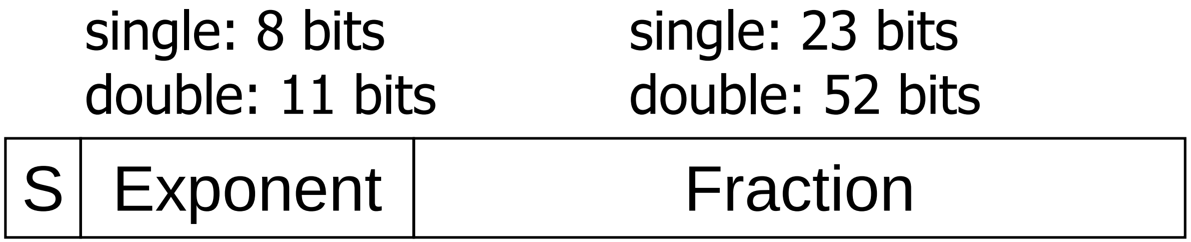
- Representation for non-integral numbers
  - Including very small and very large numbers
- Like scientific notation
  - -2.34×10^{56} normalized
  - $+0.002 \times 10^{-4}$
  - $+987.02 \times 10^{9}$

    not normalized
- In binary
  - $\pm 1.xxxxxxx_2 \times 2^{yyyy}$
- Types `float` and `double` in C

Floating Point Standard

- Defined by IEEE Std 754-1985
- Developed in response to divergence of representations
  - Portability issues for scientific code
- Now almost universally adopted
- Two representations
  - Single precision (32-bit)
  - Double precision (64-bit)

IEEE Floating-Point Format



$$x = (-1)^{S} \times (1 + \text{Fraction}) \times 2^{(\text{Exponent} - \text{Bias})}$$

- S: sign bit (0 ⇒ non-negative, 1 ⇒ negative)
- Normalize significand: 1.0 ≤ |significand| < 2.0
  - Always has a leading pre-binary-point 1 bit, so no need to represent it explicitly (hidden bit)
  - Significand is Fraction with the “1.” restored
- Exponent: excess representation: actual exponent + Bias
  - Ensures exponent is unsigned
  - Single: Bias = 127; Double: Bias = 1203

Single-Precision Range

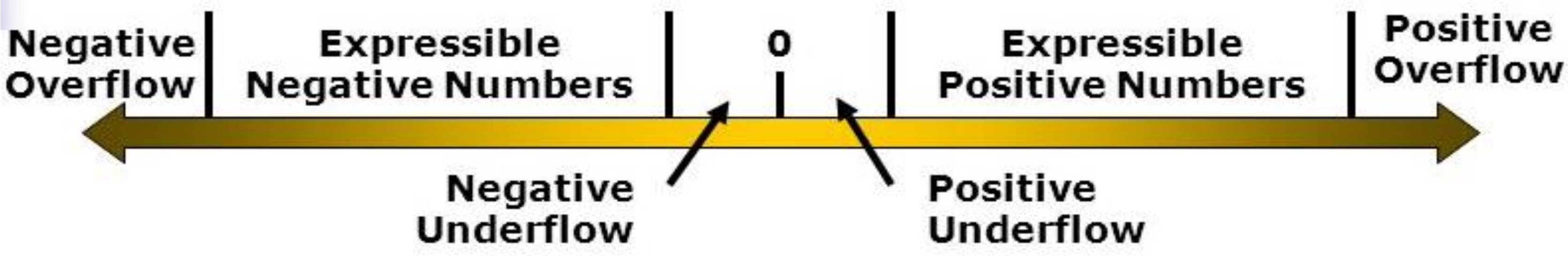
- Exponents 00000000 and 11111111 reserved
- Smallest value
  - Exponent: 00000001
    ⇒ actual exponent = 1 – 127 = –126
  - Fraction: 000…00 ⇒ significand = 1.0
  - $\pm 1.0 \times 2^{-126} \approx \pm 1.2 \times 10^{-38}$
- Largest value
  - Exponent: 11111110
    ⇒ actual exponent = 254 – 127 = +127
  - Fraction: 111…11 ⇒ significand ≈ 2.0
  - $\pm 2.0 \times 2^{+127} \approx \pm 3.4 \times 10^{+38}$

Double-Precision Range

- Exponents 0000…00 and 1111…11 reserved
- Smallest value
  - Exponent: 00000000001
    ⇒ actual exponent = 1 – 1023 = –1022
  - Fraction: 000…00 ⇒ significand = 1.0
  - $\pm 1.0 \times 2^{-1022} \approx \pm 2.2 \times 10^{-308}$
- Largest value
  - Exponent: 11111111110
    ⇒ actual exponent = 2046 – 1023 = +1023
  - Fraction: 111…11 ⇒ significand ≈ 2.0
  - $\pm 2.0 \times 2^{+1023} \approx \pm 1.8 \times 10^{+308}$

Floating-Point Precision

- Relative precision
  - all fraction bits are significant
  - Single: approx 2^{-23}
    - Equivalent to $23 \times \log_{10}2 \approx 23 \times 0.3 \approx 6$ decimal digits of precision
  - Double: approx 2^{-52}
    - Equivalent to $52 \times \log_{10}2 \approx 52 \times 0.3 \approx 16$ decimal digits of precision



Floating-Point Example（1/2）

- Represent –0.75
  - $-0.75 = (-1)^1 \times 1.1_2 \times 2^{-1}$
  - S = 1
  - Fraction = $1000\ldots00_2$
  - Exponent = –1 + Bias
    - Single: –1 + 127 = 126 = 01111110_2
    - Double: –1 + 1023 = 1022 = 01111111110_2
- Single: 1011111101000…00
- Double: 1011111111101000…00

Floating-Point Example（2/2）

- What number is represented by the single-precision float

  11000000101000…00

  - S = 1
  - Fraction = $01000\ldots00_2$
  - Fxponent = 10000001_2 = 129
- $x = (-1)^1 \times (1 + 01_2) \times 2^{(129 - 127)}$

  $= (-1) \times 1.25 \times 2^2$

  $= -5.0$

Denormal Numbers

- Exponent = 000...0 ⇒ hidden bit is 0

$$x = (-1)^S \times (0 + \text{Fraction}) \times 2^{(1\text{-Bias})}$$

- Smaller than normal numbers
  - allow for gradual underflow, with diminishing precision
- Denormal with fraction = 000...0

$$x = (-1)^S \times (0 + 0) \times 2^{(1\text{-Bias})} = \pm 0.0$$

Two representations of 0.0!

Infinities and NaNs

- Exponent = 111...1, Fraction = 000...0
  - ±Infinity
  - Can be used in subsequent calculations, avoiding need for overflow check
- Exponent = 111...1, Fraction ≠ 000...0
  - Not-a-Number (NaN)
  - Indicates illegal or undefined result
    - e.g., 0.0 / 0.0
  - Can be used in subsequent calculations

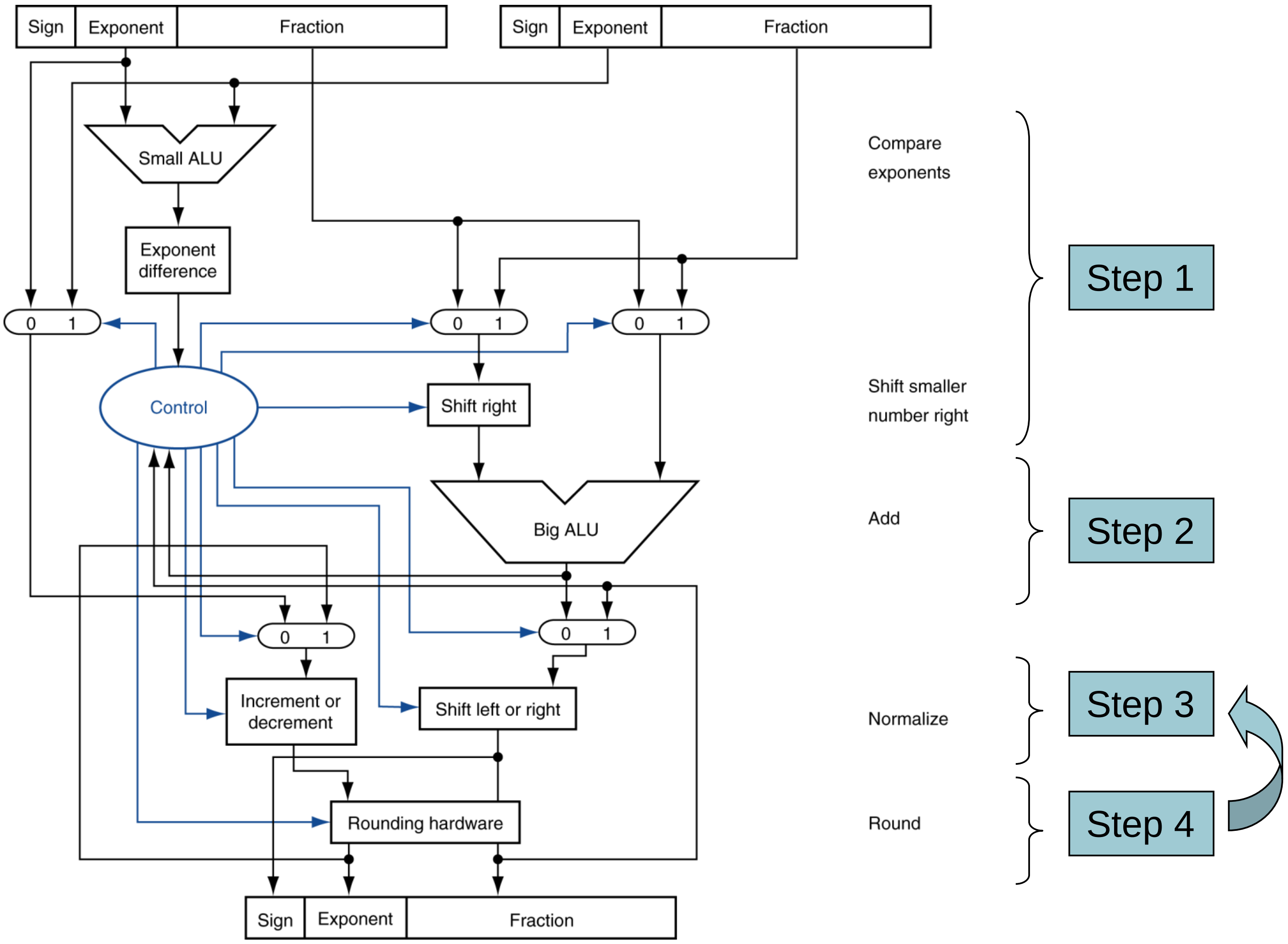
Floating-Point Addition（1/2）

- Consider a 4-digit decimal example
  - $9.999 \times 10^{1} + 1.610 \times 10^{-1}$
- 1. Align decimal points
  - Shift number with smaller exponent
  - $9.999 \times 10^{1} + 0.016 \times 10^{1}$
- 2. Add significands
  - $9.999 \times 10^{1} + 0.016 \times 10^{1} = 10.015 \times 10^{1}$
- 3. Normalize result & check for over/underflow
  - 1.0015×10^{2}
- 4. Round and renormalize if necessary
  - 1.002×10^{2}

Floating-Point Addition（2/2）

- Now consider a 4-digit binary example
  - $1.000_2 \times 2^{-1} + -1.110_2 \times 2^{-2}$ (0.5 + –0.4375)
- 1. Align binary points
  - Shift number with smaller exponent
  - $1.000_2 \times 2^{-1} + -0.111_2 \times 2^{-1}$
- 2. Add significands
  - $1.000_2 \times 2^{-1} + -0.111_2 \times 2^{-}1 = 0.001_2 \times 2^{-1}$
- 3. Normalize result & check for over/underflow
  - $1.000_2 \times 2^{-4}$, with no over/underflow
- 4. Round and renormalize if necessary
  - $1.000_2 \times 2^{-4}$ (no change) = 0.0625

FP Adder Hardware（1/2）



Compare exponents

Shift smaller number right

Step 1

Add

Step 2

Normalize

Step 3

Round

Step 4

FP Adder Hardware（2/2）

- Much more complex than integer adder
- Doing it in one clock cycle would take too long
  - Much longer than integer operations
  - Slower clock would penalize all instructions
- FP adder usually takes several cycles
  - Can be pipelined

Floating-Point Multiplication（1/2）

- Consider a 4-digit decimal example
  - $1.110 \times 10^{10} \times 9.200 \times 10^{-5}$
- 1. Add exponents
  - For biased exponents, subtract bias from sum
  - New exponent = 10 + –5 = 5
- 2. Multiply significands
  - $1.110 \times 9.200 = 10.212 \;\Rightarrow\; 10.212 \times 10^{5}$
- 3. Normalize result & check for over/underflow
  - 1.0212×10^{6}
- 4. Round and renormalize if necessary
  - 1.021×10^{6}
- 5. Determine sign of result from signs of operands
  - $+1.021 \times 10^{6}$

Floating-Point Multiplication（2/2）

- Now consider a 4-digit binary example
  - $1.000_2 \times 2^{-1} \times -1.110_2 \times 2^{-2}$ (0.5 × –0.4375)
- 1. Add exponents
  - Unbiased: –1 + –2 = –3
  - Biased: (–1 + 127) + (–2 + 127) = –3 + 254 – 127 = –3 + 127
- 2. Multiply significands
  - $1.000_2 \times 1.110_2 = 1.1102 \Rightarrow 1.110_2 \times 2^{-3}$
- 3. Normalize result & check for over/underflow
  - $1.110_2 \times 2^{-3}$ (no change) with no over/underflow
- 4. Round and renormalize if necessary
  - $1.110_2 \times 2^{-3}$ (no change)
- 5. Determine sign: +ve × –ve ⇒ –ve
  - $-1.110_2 \times 2^{-3} = -0.21875$

FP Arithmetic Hardware

- FP multiplier is of similar complexity to FP adder
  - But uses a multiplier for significands instead of an adder
- FP arithmetic hardware usually does
  - Addition, subtraction, multiplication, division, reciprocal, square-root
  - FP ↔ integer conversion
- Operations usually takes several cycles
  - Can be pipelined

FP Instructions in RISC-V（1/2）

- Separate FP registers: f0, …, f31
  - double-precision
  - single-precision values stored in the lower 32 bits
- FP instructions operate only on FP registers
  - Programs generally don’t do integer ops on FP data, or vice versa
  - More registers with minimal code-size impact
- FP load and store instructions
  - `flw, fld`
  - `fsw, fsd`

FP Instructions in RISC-V（2/2）

- Single-precision arithmetic
  - `fadd.s, fsub.s, fmul.s, fdiv.s, fsqrt.s`
    - e.g., `fadd.s f2, f4, f6`
- Double-precision arithmetic
  - `fadd.d, fsub.d, fmul.d, fdiv.d, fsqrt.d`
    - e.g., `fadd.d f2, f4, f6`
- Single- and double-precision comparison
  - `feq.s, flt.s, fle.s`
  - `feq.d, flt.d, fle.d`
  - Result is 0 or 1 in integer destination register
    - Use beq, bne to branch on comparison result

FP Example: °F to °C

- C code:

```
float f2c (float fahr) {
  return ((5.0f/9.0f)*(fahr - 32.0f));
}
```

  - `fahr` in f10, result in f10, literals in global memory space

- Compiled RISC-V code:

```
f2c:
  flw    f0,const5(x3)   // f0 = 5.0f
  flw    f1,const9(x3)   // f1 = 9.0f
  fdiv.s f0, f0, f1      // f0 = 5.0f / 9.0f
  flw    f1,const32(x3)  // f1 = 32.0f
  fsub.s f10,f10,f1      // f10 = fahr – 32.0
  fmul.s f10,f0,f10      // f10 = (5.0f/9.0f) * (fahr–32.0f)
  jalr   x0,0(x1)        // return
```

FP Example: Array Multiplication（1/3）

- C = C + A × B
  - All 32 × 32 matrices, 64-bit double-precision elements
- C code:

```
void mm (double c[][],
         double a[][], double b[][]) {
  size_t i, j, k;
  for (i = 0; i < 32; i = i + 1)
    for (j = 0; j < 32; j = j + 1)
      for (k = 0; k < 32; k = k + 1)
        c[i][j] = c[i][j]
                  + a[i][k] * b[k][j];
}
```

  - Addresses of c, a, b in x10, x11, x12, and i, j, k in x5, x6, x7

FP Example: Array Multiplication（2/3）

- RISC-V code:

```
mm:...
        addi    x28,x0,32        // x28 = 32 (row size/loop end)
        addi    x5,x0,0          // i = 0; initialize 1st for loop
L1:     addi    x6,x0,0          // j = 0; initialize 2nd for loop
L2:     addi    x7,x0,0          // k = 0; initialize 3rd for loop
        slli  x30,x5,5      // x30 = i * 2**5 (size of row of c)
        add   x30,x30,x6    // x30 = i * size(row) + j
        slli  x30,x30,3     // x30 = byte offset of [i][j]
        add   x30,x10,x30   // x30 = byte address of c[i][j]
        fld   f0,0(x30)     // f0 = c[i][j]
L3:     slli  x29,x7,5      // x29 = k * 2**5 (size of row of b)
        add   x29,x29,x6    // x29 = k * size(row) + j
        slli  x29,x29,3     // x29 = byte offset of [k][j]
        add   x29,x12,x29   // x29 = byte address of b[k][j]
        fld   f1,0(x29)     // f1 = b[k][j]
```

FP Example: Array Multiplication（3/3）

…

```
slli   x29,x5,5     // x29 = i * 2**5 (size of row of a)
add    x29,x29,x7   // x29 = i * size(row) + k
slli   x29,x29,3    // x29 = byte offset of [i][k]
add    x29,x11,x29  // x29 = byte address of a[i][k]
fld    f2,0(x29)    // f2 = a[i][k]
fmul.d f1, f2, f1   // f1 = a[i][k] * b[k][j]
fadd.d f0, f0, f1   // f0 = c[i][j] + a[i][k] * b[k][j]
addi   x7,x7,1      // k = k + 1
bltu   x7,x28,L3    // if (k < 32) go to L3
fsd    f0,0(x30)    // c[i][j] = f0
addi   x6,x6,1      // j = j + 1
bltu   x6,x28,L2    // if (j < 32) go to L2
addi   x5,x5,1      // i = i + 1
bltu   x5,x28,L1    // if (i < 32) go to L1
```

Accurate Arithmetic

- IEEE Std 754 specifies additional rounding control
  - Extra bits of precision (guard, round, sticky)
  - Choice of rounding modes
  - Allows programmer to fine-tune numerical behavior of a computation
- Not all FP units implement all options
  - Most programming languages and FP libraries just use defaults
- Trade-off between hardware complexity, performance, and market requirements

Right Shift and Division

§3.9 Fallacies and Pitfalls

- Left shift by *i* places multiplies an integer by 2^i
- Right shift divides by 2^i?
  - Only for unsigned integers
- For signed integers
  - Arithmetic right shift: replicate the sign bit
  - e.g., –5 / 4
    - $11111011_2 >> 2 = 11111110_2 = -2$
    - Rounds toward –∞
  - c.f. $11111011_2 >>> 2 = 00111110_2 = +62$

Associativity

- Parallel programs may interleave operations in unexpected orders
  - Assumptions of associativity may fail

		(x+y)+z	x+(y+z)
x	-1.50E+38	0.00E+00	-1.50E+38
y	1.50E+38		1.50E+38
z	1.0	1.0	
		1.00E+00	0.00E+00

- Need to validate parallel programs under varying degrees of parallelism

Who Cares About FP Accuracy?

- Important for scientific code
  - But for everyday consumer use?
    - “My bank balance is out by 0.0002¢!” ☹

- The Intel Pentium FDIV bug
  - The market expects accuracy
  - See Colwell, *The Pentium Chronicles*

Concluding Remarks

- Bits have no inherent meaning
  - Interpretation depends on the instructions applied
- Computer representations of numbers
  - Finite range and precision
  - Need to account for this in programs

Concluding Remarks

- ISAs support arithmetic
  - Signed and unsigned integers
  - Floating-point approximation to reals
- Bounded range and precision
  - Operations can overflow and underflow