

Computer Architecture and Operating Systems Lecture 7: Floating-Point Format

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Floating-Point Format

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

single: 8 bits single: 23 bits double: 11 bits double: 52 bits

S Exponent Fraction

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

- **S**: sign bit $(0 \Rightarrow \text{non-negative}, 1 \Rightarrow \text{negative})$
- Normalized significand: 1.0 ≤ |significand| < 2.0</p>
 - 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: $00000001 \Rightarrow \text{actual exponent} = 1 127 = -126$
 - Fraction: $000...00 \Rightarrow \text{significand} = 1.0$
 - $\pm 1.0 \times 2^{-126} \approx \pm 1.2 \times 10^{-38}$

Largest value

- exponent: $111111110 \Rightarrow$ 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: $0000000001 \Rightarrow \text{actual exponent} = 1 1023 = -1022$
 - Fraction: $000...00 \Rightarrow \text{significand} = 1.0$
 - $\pm 1.0 \times 2^{-1022} \approx \pm 2.2 \times 10^{-308}$

Largest value

- 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

- Represent –0.75
 - $-0.75 = (-1)^1 \times 1.1_2 \times 2^{-1}$
 - -S = 1
 - Fraction = 1000...00₂
 - Exponent = -1 + Bias
 - Single: $-1 + 127 = 126 = 011111110_2$
 - Double: $-1 + 1023 = 1022 = 011111111110_2$
- ■Single: 1011111101000...00
- Double: 10111111111101000...00

Floating-Point Example

 What number is represented by the single-precision float 11000000101000...00

- -S = 1
- Fraction = $01000...00_2$
- **-** 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

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

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

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

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

Special Values

Single precision		Double precision		Object represented
Exponent	Fraction	Exponent	Fraction	
0	0	0	0	0
0	Nonzero	0	Nonzero	± denormalized number
1-254	Anything	1-2046	Anything	± floating-point number
255	0	2047	0	± infinity
255	Nonzero	2047	Nonzero	NaN (Not a Number)

Floating-Point Addition

- 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

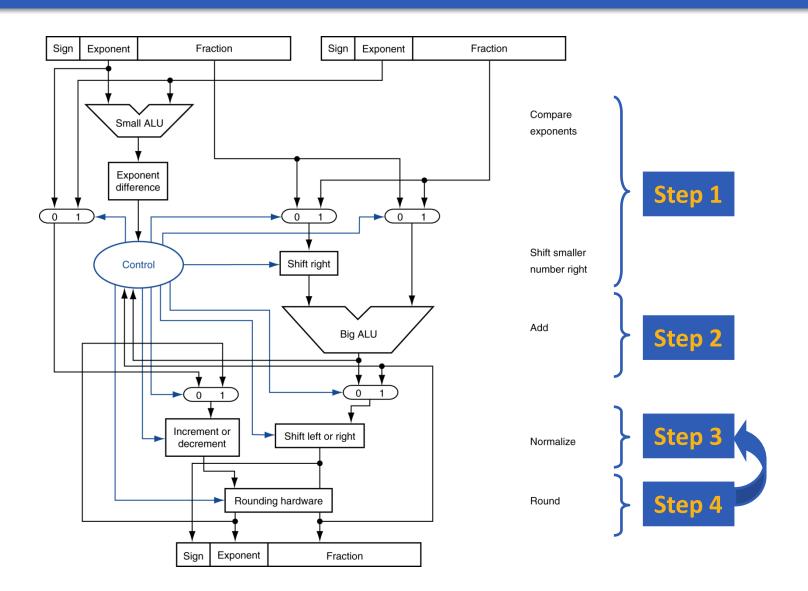
Floating-Point Addition

- 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

- 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

FP Adder Hardware



Floating-Point Multiplication

- 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
 - \blacksquare 1.110 × 9.200 = 10.212 \Rightarrow 10.212 × 10⁵
- 3. Normalize result & check for over/underflow
 - -1.0212×10^6
- 4. Round and renormalize if necessary
 - 1.021 × 10⁶
- 5. Determine sign of result from signs of operands
 - $+1.021 \times 10^6$

Floating-Point Multiplication

- Now consider a 4-digit binary example
 - $-1.000_2 \times 2^{-1} \times -1.110_2 \times 2^{-2} (0.5 \times -0.4375)$
- 1. Add exponents
 - Unbiased: -1 + -2 = -3
 - Biased: (-1 + 127) + (-2 + 127) = -3 + 254 127 = -3 + 127
- 2. Multiply significands
 - $1.000_2 \times 1.110_2 = 1.1102 \implies 1.110_2 \times 2^{-3}$
- 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 \times -ve \Rightarrow -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

- ■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

- Single-precision arithmetic
 - fadd.s, fsub.s, fmul.s, fdiv.s, fsqrt.se.g., fadds.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
- Branch on FP condition code true or false
 - b.cond

FP Example: °F to °C

C code:

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

- 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

- $\blacksquare C = C + A \times B$
 - 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];
}</pre>
```

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

FP Example: Array Multiplication

RISC-V code:

```
mm: . . .
      lί
            x28,32
                        // x28 = 32 (row size/loop end)
      lί
            x5,0
                        // i = 0; initialize 1st for loop
      lί
            x6,0 // j = 0; initialize 2nd for loop
L1:
L2: li
                  // k = 0; initialize 3rd for loop
            x7,0
      slli x30,x5,5 // x30 = i * 2**5 (size of row of c)
            x30,x30,x6 // x30 = i * size(row) + j
      add
      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]
      slli
            x29,x7,5 // x29 = k * 2**5 (size of row of b)
L3:
            x29,x29,x6 // x29 = k * size(row) + j
      add
      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

...

```
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]
bltu x7, x28, L3 // if (k < 32) go to L3
fsd f0,0(x30) // c[i][j] = f0
bltu x6,x28,L2 // if (j < 32) go to L2
addi x5, x5, 1 // i = i + 1
     x5,x28,L1 // if (i < 32) go to L1
bltu
```

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

Associativity

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

		(x+y)+z	x+(y+z)
X	-1.50E+38		-1.50E+38
У	1.50E+38	0.00E+00	
Z	1.0	1.0	1.50E+38
		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

Any Questions?

```
__start: addi t1, zero, 0x18
    addi t2, zero, 0x21

cycle: beg t1, t2, done
    slt t0, t1, t2
    bne t0, zero, if_less
    nop
    sub t1, t1, t2
    j cycle
    nop

if_less: sub t2, t2, t1
    j cycle

done: add t3, t1, zero
```