# CPT\_S 260 Intro to Computer Architecture Lecture 10

Floating Point Arithmetic February 2, 2022

**Ganapati Bhat** 

School of Electrical Engineering and Computer Science
Washington State University

## **Announcements**

- Homework 2 is online
  - Due next Friday

## Recap: Fractional Part in Binary Format

 Repeatedly multiply the fraction by 2 and save the resulting integer digits. The digits for the binary number are the 0,1 in order of their computation.

Convert 46.6875 to binary!

## **Recap: 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  $(0 \Rightarrow \text{non-negative}, 1 \Rightarrow \text{negative})$
- Normalize 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 = 1023

## **Recap: Bias Exponent Representation**

- 2's complement makes it difficult to compare exponents
- 1 is (111..111) where is 1 is (000....001). If we just look at 2's complement number, we cannot tell which has higher exponent
- IEEE 754 uses a bias of 127 for single precision, so an exponent of -1 is represented by the bit pattern of the value -1 + 127<sub>ten</sub>, or 126<sub>ten</sub> = 0111 1110<sub>two</sub>
- +1  $\rightarrow$  1 + 127, or 128<sub>ten</sub> = 1000 0000<sub>two</sub>

■ The exponent bias for *double* precision is 1023.

## **Single-Precision Range**

#### Exponents 00000000 and 11111111 reserved

#### Smallest value

- Exponent (also called biased exponent):  $00000001 \Rightarrow \text{actual exponent} = 1 127 = -126$
- Fraction: 000...00 ⇒ significand = 1.0
- $-\pm 1.0 \times 2^{-126} \approx \pm 1.2 \times 10^{-38}$

#### Largest value

- Exponent (also called biased exponent):  $111111110 \Rightarrow \text{actual exponent} = 254 127 = +127$
- 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 (next slides show for what these exponents are used)
  - 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

- Fraction: 111...11 ⇒ significand ≈ 2.0
- $-\pm 2.0 \times 2^{+1023} \approx \pm 1.8 \times 10^{+308}$

## **Special Cases**

#### Zero

Sign bit = 0; biased exponent = all 00 bits; and the fraction = all 00 bits;

#### Positive and Negative Infinity

 Sign bit = 00 for positive infinity, 11 for negative infinity; biased exponent = all 11 bits; and the fraction = all 00 bits;

#### NaN (Not-A-Number)

 Sign bit = 0 or 1; biased exponent = all 11 bits; and the fraction is anything but all 00 bits. NaN's occurs when one does an invalid operation on a floating point value, such as dividing by zero, or taking the square root of a negative number.

## Therefore, for Infinities and NaNs, we have

- Exponent = 111...1, Fraction = 000...0
  - ±Infinity -sign bit =0 for (+); sign bit =1 for (-)
  - 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 Precision**

#### Relative precision

- All fraction bits are significant
- Single: approx 2<sup>-23</sup>
  - Equivalent to 23 ×  $\log_{10}$ 2 ≈ 23 × 0.3 ≈ 6 decimal digits of precision
- Double: approx 2<sup>-52</sup>
  - Equivalent to  $52 \times \log_{10} 2 \approx 52 \times 0.3 \approx 16$  decimal digits of precision

# Floating-Point Example #1

- Represent –0.75
  - $-0.75 = (-1)^1 \times 1.1_2 \times 2^{-1}$
  - S = 1
  - Fraction = 1000...00
  - Exponent = -1 + Bias
    - Single: -1 + 127 = 126 = 01111110
    - Double: -1 + 1023 = 1022 = 01111111110
- Single: 1011111101000...00
- Double: 10111111111101000...00

# Floating-Point Example #2

- What number is represented by the single-precision float
- **11000000101000...00** 
  - S = 1
  - Fraction = 01000...00
  - Exponent = 10000001 = 129
- $\mathbf{x} = (-1) \times (1 + .01) \times 2^{(129-127)}$ 
  - $= (-1) \times 1.25 \times 2^2$
  - **■** = -5.0

### **Overflow and Underflow**

- Overflow occurs when a number is larger than the largest number that can be represented
- Underflow occurs when a number is smaller than the smallest number that can be represented

# **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
  - $-1.0015 \times 10^{2}$
- 4. Round and renormalize if necessary
  - $-1.002 \times 10^2$

# **Floating-Point Addition**

Now consider a 4-digit binary example

■ 
$$1.000 \times 2^{-1} + -1.110 \times 2^{-2} (0.5 + -0.4375)$$

- 1. Align binary points
  - Shift number with smaller exponent
  - $-1.000 \times 2^{-1} + -0.111 \times 2^{-1}$
- 2. Add significands

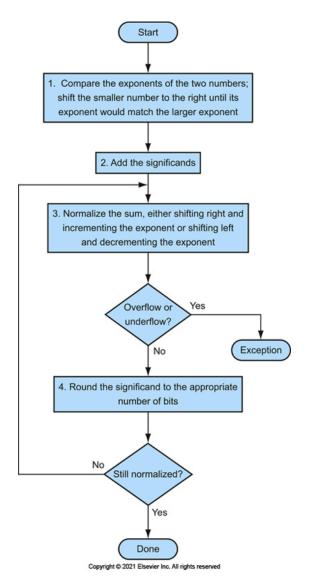
$$-1.000 \times 2^{-1} + -0.111 \times 2^{-1} = 0.001 \times 2^{-1}$$

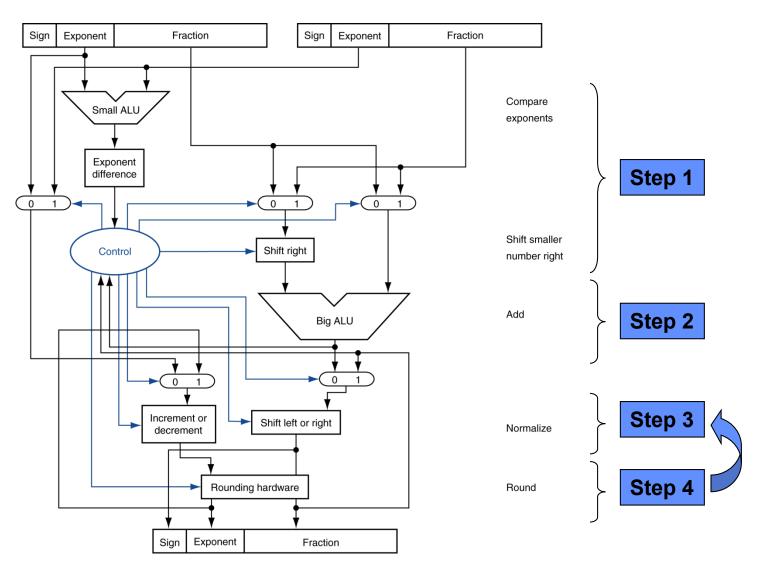
- 3. Normalize result & check for over/underflow
  - 1.000 ×  $2^{-4}$ , with no over/underflow
- 4. Round and renormalize if necessary
  - $1.000_2 \times 2^{-4}$  (no change) =  $0.0625_{10}$

### **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
  - $1.110 \times 9.200 = 10.212 \Rightarrow 10.212 \times 10^{5}$
- 3. Normalize result & check for over/underflow
  - 1.0212 × 10<sup>6</sup>
- 4. Round and renormalize if necessary
  - 1.021 × 10<sup>6</sup>
- 5. Determine sign of result from signs of operands
  - +1.021 × 10<sup>6</sup>

# 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

■ 
$$1.000_2 \times 1.110_2 = 1.110_2 \Rightarrow 1.110_2 \times 2^{-3}$$

#### 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

$$-1.110_2 \times 2^{-3} = -0.21875$$

## **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

### **Guard and Round Bits**

- Used to hold intermediate operands before truncating
- Guard is the first of two extra bits
- Round is the second extra bit

Number	Guard	Round
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- Example
- $-2.56 \times 10^{0} + 2.34 \times 10^{2}$

## **IEEE 754 Rounding Modes**

- Always round up (Towards infinity)
- Always round down (Towards negative infinity)
- Truncate
- Round to nearest even
- Tricky in the half way case (0.5)
  - If the LSB retained would be odd, add one
  - If the LSB would be even, truncate
  - Gives a zero in the LSB

## **Sticky Bit**

- A bit used to track whenever there are non-zero bits to the right of the round bit
- Helps differentiate between 0.5000000...0 and 0.500000...1 when rounding
- $\bullet$  5.01 × 10<sup>-1</sup> + 2.34 × 10<sup>2</sup>

## **Sticky Bit**

- A bit used to track whenever there are non-zero bits to the right of the round bit
- Helps differentiate between 0.5000000...0 and 0.500000...1 when rounding
- $\bullet$  5.01 × 10<sup>-1</sup> + 2.34 × 10<sup>2</sup>
- Without sticky bit
- Add 0.0050 and 2.34
- Sum of 2.3450
- Round off to 2.34

- With sticky bit
- Add 0.0050 and 2.34
- Sticky bit would be set
- Sum of 2.3450 with sticky bit 1
- Round off to 2.35