

# Floating Point Arithmetic

CS 365

## Floating-Point

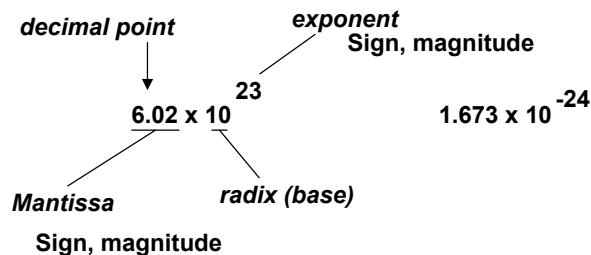
What can be represented in N bits?

- Unsigned                    0            to             $2^N$
- 2s Complement             $-2^{N-1}$     to             $2^{N-1} - 1$
- But, what about?
  - very large numbers?  
9,349,398,989,787,762,244,859,087,678
  - very small number?  
0.00000000000000000000000045691
  - rationals                     $2/3$
  - irrationals                  $\sqrt{2}$
  - transcendentals             $e, \pi$

## Floating Point

- We need a way to represent
  - numbers with fractions, e.g., 3.1416
  - very small numbers, e.g., .000000001
  - very large numbers, e.g.,  $3.15576 \times 10^9$
- Representation:
  - sign, exponent, significand:  $(-1)^{\text{sign}} \times \text{significand} \times 2^{\text{exponent}}$
  - more bits for significand gives more accuracy
  - more bits for exponent increases range
- IEEE 754 floating point standard:
  - single precision: 8 bit exponent, 23 bit significand
  - double precision: 11 bit exponent, 52 bit significand

## Recall Scientific Notation



IEEE F.P.  $\pm 1.M \times 2^e - 127$

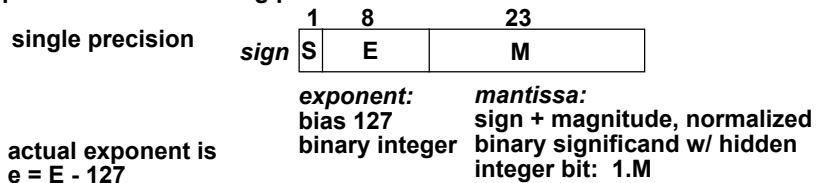
- Issues:
  - Arithmetic (+, -, \*, /)
  - Representation, Normal form
  - Range and Precision
  - Rounding
  - Exceptions (e.g., divide by zero, overflow, underflow)
  - Errors

## IEEE 754 floating-point standard

- Leading “1” bit of significand is implicit
- Exponent is “biased” to make sorting easier
  - all 0s is smallest exponent all 1s is largest
  - bias of 127 for single precision and 1023 for double precision
  - summary:  $(-1)^{\text{sign}} \times (1 + \text{significand}) \times 2^{\text{exponent} - \text{bias}}$
- Example:
  - decimal:  $-.75 = -3/4 = -3/2^2$
  - binary:  $-.11 = -1.1 \times 2^{-1}$
  - floating point: exponent = 126 = 01111110
  - IEEE single precision: 10111111010000000000000000000000

## IEEE 754 Standard

Representation of floating point numbers in IEEE 754 standard:



actual exponent is  
 $e = E - 127$

$$0 < E < 255$$

$$N = (-1)^S 2^{E-127} (1.M)$$

$$0 = 0 \text{ 00000000 } 0 \dots 0 \quad -1.5 = 1 \text{ 01111111 } 10 \dots 0$$

Magnitude of numbers that can be represented is in the range:

$$2^{-126} (1.0) \quad \text{to} \quad 2^{127} (2 - 2^{-23})$$

which is approximately:

$$1.8 \times 10^{-38} \quad \text{to} \quad 3.40 \times 10^{38}$$

## Floating Point Complexities

- Operations are somewhat more complicated
- In addition to overflow we can have “underflow”
- Accuracy can be a big problem
  - IEEE 754 keeps two extra bits, guard and round
  - four rounding modes
  - positive divided by zero yields “infinity”
  - zero divide by zero yields “not a number”
  - other complexities
- Implementing the standard can be tricky
- Not using the standard can be even worse
  - see text for description of 80x86 and Pentium bug!

## Floating Point Addition Example

Example: Add  $9.999 \times 10^1$  and  $1.610 \times 10^{-1}$  assuming 4 decimal digits

1. Align decimal point of number with smaller exponent

$$1.610 \times 10^{-1} = 0.161 \times 10^0 = 0.0161 \times 10^1$$

Shift smaller number to right

2. Add significands

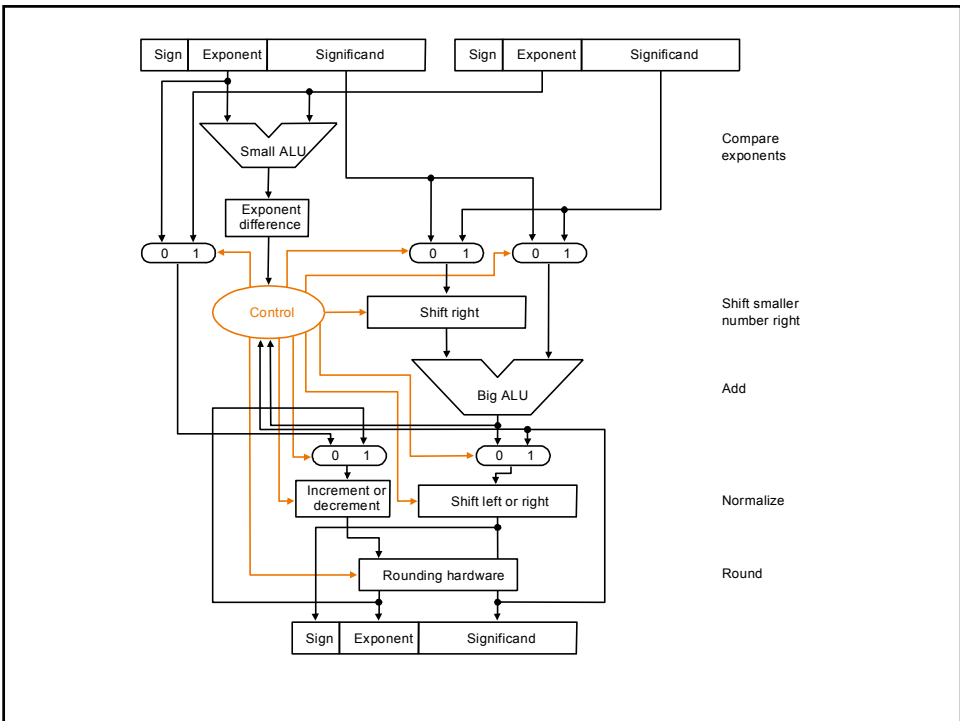
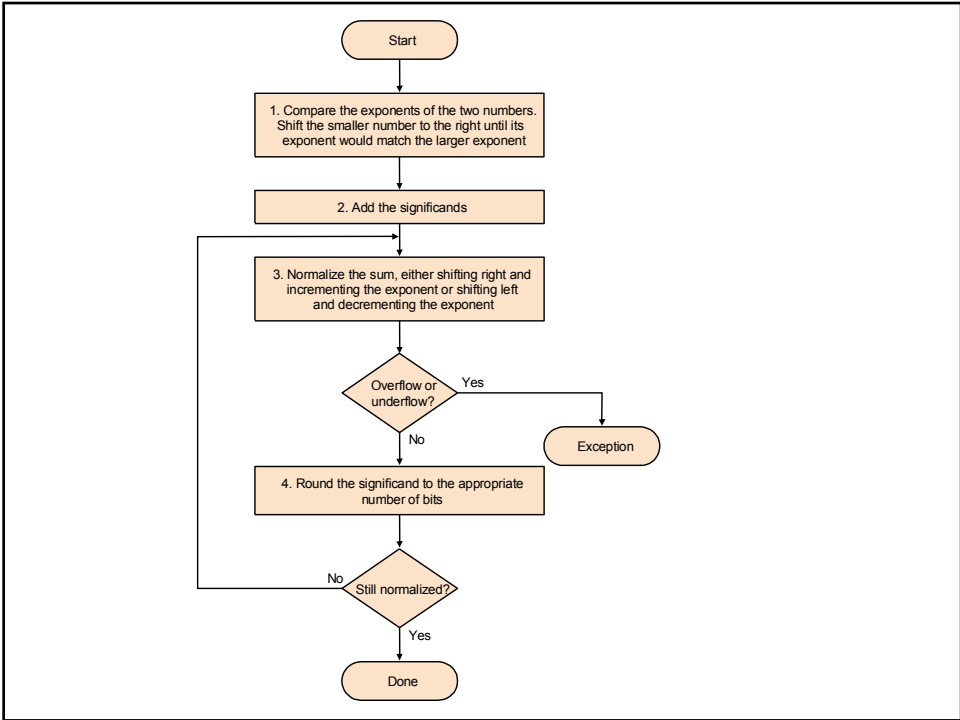
$$\begin{array}{r} 9.999 \\ 0.016 \\ \hline 10.015 \end{array} \quad \rightarrow \text{SUM} = 10.015 \times 10^1$$

NOTE: One digit of precision lost during shifting. Also sum is not normalized

3. Shift sum to put it in normalized form  $1.0015 \times 10^2$
4. Since significant only has 4 digits, we need to round the sum

$$\text{SUM} = 1.002 \times 10^2$$

NOTE: normalization maybe needed again after rounding,  
e.g, rounding 9.9999 you get 10.000



## Accurate Arithmetic – Guard & Round bits

- IEEE 754 standard specifies the use of 2 extra bits on the right during intermediate calculations – **Guard** bit and **Round** bit
- Example: Add  $2.56 \times 10^0$  and  $2.34 \times 10^2$  assuming 3 significant digits and **without guard and round bits**

$$2.56 \times 10^0 = 0.0256 \times 10^2$$

$$\begin{array}{r} 2.34 \\ \underline{0.02} \\ 2.36 \times 10^2 \end{array}$$

- **With guard and round bits**

$$\begin{array}{r} 2.34 \\ \underline{0.0256} \\ 2.3656 \times 10^2 \\ \text{ROUND} \rightarrow 2.37 \times 10^2 \end{array}$$

## Chapter Four Summary

- Computer arithmetic is constrained by limited precision
- Bit patterns have no inherent meaning but standards do exist
  - two's complement
  - IEEE 754 floating point
- Computer instructions determine “meaning” of the bit patterns
- Performance and accuracy are important so there are many complexities in real machines (i.e., algorithms and implementation).
- Next class: we are ready to move on (and implement the processor)