# Data Representation – Floating Point

CSCI 2400 / ECE 3217: Computer Architecture

**Instructor:** 

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Slides adapted from Bryant & O'Hallaron's slides via Jason Fritts

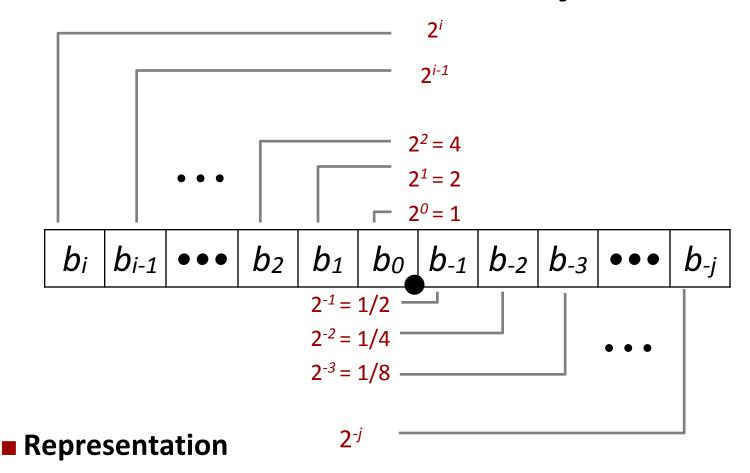
### **Today: Floating Point**

- **■** Background: Fractional binary numbers
- Example and properties
- **IEEE floating point standard: Definition**
- Floating point in C
- Summary

#### **Fractional binary numbers**

- What is 1011.101<sub>2</sub>?
- How can we express fractions like ¼ in binary?

### **Place-Value Fractional Binary Numbers**



- Bits to right of "binary point" represent fractional powers of 2
- Represents rational number:

$$\sum_{k=-i}^{i} b_k \times 2^k$$

### **Fractional Binary Numbers: Examples**

#### Value

#### Representation

$$5^{3}/_{4}$$
 $2^{7}/_{8}$ 
 $^{25}/_{64}$ 

$$= 4 + 1 + \frac{1}{2} + \frac{1}{4} = 5^{3}/_{4}$$

$$= 2 + \frac{1}{2} + \frac{1}{4} + \frac{1}{8} = 2^{7}/_{8}$$

$$= \frac{1}{4} + \frac{1}{8} + \frac{1}{64} = \frac{25}{64}$$

#### Observations

- Divide by 2 by shifting right
- Multiply by 2 by shifting left

#### Limitations

- Can only exactly represent numbers of the form  $x/2^k$
- Other rational numbers have repeating bit representations

<u>Value</u>	<u>Representation</u>
1/3	0.0101010101 <b>[01]</b> <sub>2</sub>
1/5	0.001100110011 <b>[0011]</b> <sub>2</sub>

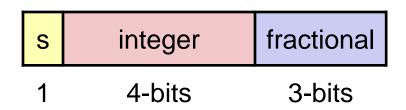
Limited range when used with "fixed point" representations

#### Convert:

 $255 \%_{16}$  to binary  $10101.10101_2$  to decimal

Suppose an 8-bit fixed-point representation with:

- One sign bit
- Four integer bits
- Three fractional bits



#### **Convert:**

 $12^{1}/_{8}$  to binary  $-6^{3}/_{8}$  to binary  $11010110_{2}$  to decimal

What bit pattern(s) have the largest positive value? What is it? What bit pattern(s) have the value closest to zero? What bit pattern(s) have the value of zero?

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# **Floating Point Representation**

Similar to scientific notation

**E.g.:** 
$$1.25 \times 10^3 = 1,250$$

**E.g.:** 
$$2.78 \times 10^{-2} = 0.0278$$

- FP is this concept but with an efficient binary format! But...
  - Uses base 2 instead of base 10
  - Places restrictions on how certain values are represented
  - Deals with finiteness of representation

### **Floating Point Representation**

#### Numerical Form:

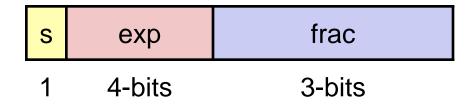
$$(-1)^{s} M 2^{E}$$

- Sign bit s determines whether number is negative or positive
- **Significand** (mantissa) *M* normally a fractional value in range [1.0, 2.0)
- **Exponent** *E* weights value by power of two

#### Encoding

- S is sign bit s
- exp field encodes E (but is not equal to E)
- frac field encodes M (but is not equal to M)

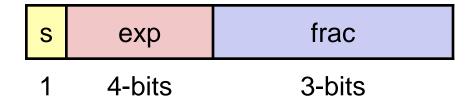
# **Tiny Floating Point Example**



#### 8-bit Floating Point Representation

- the sign bit is in the most significant bit
- the next four bits are the exponent (exp)
  - exp (not E) encoded as a 4-bit unsigned integer
  - Uses a bias to represent negative exponents
- the last three bits are the fraction (frac)
  - encodes fractional part of a fractional binary number

### **Tiny Floating Point Example**



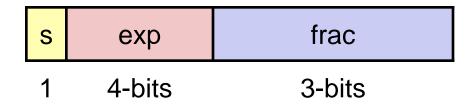
#### Fractional (Mantissa) Value

- Three bits- encoded left-to-right as b<sub>0</sub>b<sub>1</sub>b<sub>2</sub>
- Place value of  $b_0$  is 1/2, of  $b_1$  is 1/4, and  $b_2$  is 1/8
- Value usually includes an implied 1:

$$M = 1 + \frac{1}{2}b_0 + \frac{1}{4}b_1 + \frac{1}{8}b_2$$

■ If exp == 0000 then the implied 1 is not used

# **Tiny Floating Point Example**



#### Exponent bias

- enable exponent to represent both positive and negative powers of 2
- use half of range for positive and half for negative power
- given k exponent bits, bias is then  $2^{k-1}-1$

#### Exponent Value

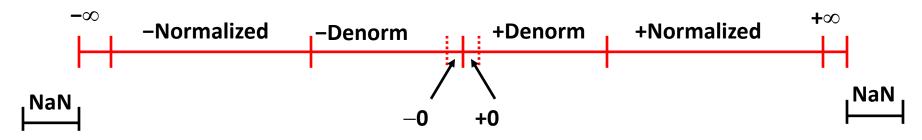
- is usually computed *E = exp bias*
- if exp == 0000 then *E* = 1 bias

### Floating Point Encodings and Visualization

#### ■ Five encodings:

- Two general forms: normalized, denormalized
- Three special values: zero, infinity, NaN (not a number)

<u>Name</u>	Exponent (exp)	Fraction (frac)
zero	exp == 0000	<b>frac</b> == 000
denormalized	exp == 0000	<b>frac</b> != 000
normalized	0000 < exp < 1111	<b>frac</b> != 000
infinity	<b>exp</b> == 1111	<b>frac</b> == 000
NaN	<b>exp</b> == 1111	<b>frac</b> != 000



 $V = (-1)^s M 2^E$ 

# **Dynamic Range (Positives)**

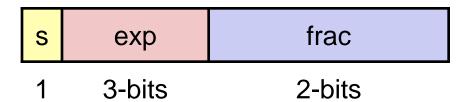
			0				·	norm: E = Exp — Bias
	s	exp	frac	E	Value			
	_							denorm: $E = 1 - Bias$
	0	0000	000	-6	0			
	0	0000	001	-6	1/8*1/64	= 1	1/51	closest to zero
Denormalized	0	0000	010	-6	2/8*1/64	= ;	2/51	
numbers								
	0	0000	110	-6	6/8*1/64	=	6/51	12
	0	0000	111	-6	7/8*1/64	=	7/51	largest denorm
	0	0001	000	-6	8/8*1/64	=	8/51	smallest norm
	0	0001	001	-6	9/8*1/64	=	9/51	L2
	•••							
	0	0110	110	-1	14/8*1/2	= 1	14/1	L6
	0	0110	111	-1	15/8*1/2	= ;	15/1	closest to 1 below
Normalized	0	0111	000	0	8/8*1	= 1	1	
numbers	0	0111	001	0	9/8*1	=	9/8	closest to 1 above
	0	0111	010	0	10/8*1	= 1	10/8	3
	•••							
	0	1110	110	7	14/8*128	= ;	224	
	0	1110	111	7	15/8*128	= :	240	largest norm
	0	1111	000	n/a	inf			infinity
	0	1111	ххх	n/a	NaN			NaN (not a number)

#### **Distribution of Values**

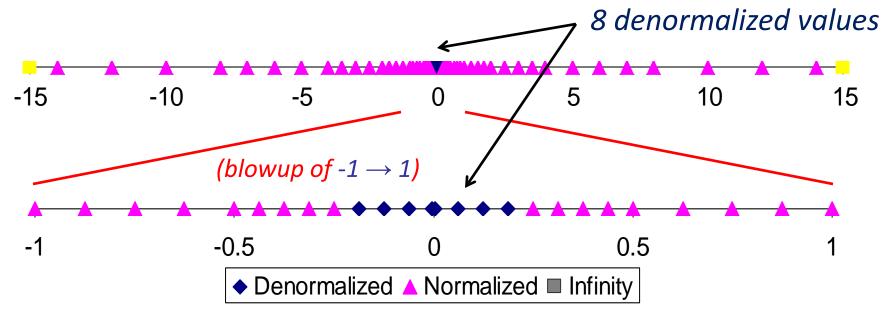
(reduced format from 8 bits to 6 bits for visualization)

#### 6-bit IEEE-like format

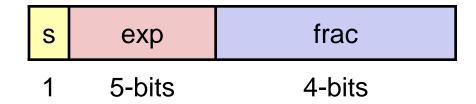
- e = 3 exponent bits
- f = 2 fraction bits
- Bias is  $2^{3-1}-1=3$



Notice how the distribution gets denser toward zero.

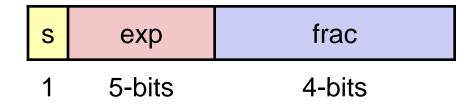


- 10-bit IEEE-like format
  - e = 5 exponent bits
  - f = 4 fraction bits



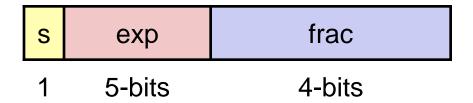
■ What is the exponent bias?

- 10-bit IEEE-like format
  - e = 5 exponent bits
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■ What is the exponent bias?  $2^{5-1} - 1 = 15$ 

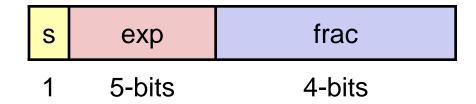
- 10-bit IEEE-like format
  - e = 5 exponent bits
  - f = 4 fraction bits



- What is the exponent bias?  $2^{5-1} 1 = 15$
- How many denormalized numbers are there?

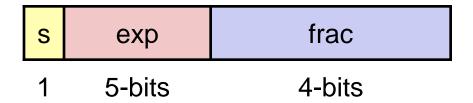
#### ■ 10-bit IEEE-like format

- e = 5 exponent bits
- f = 4 fraction bits



- What is the exponent bias?  $2^{5-1} 1 = 15$
- How many denormalized numbers are there?
  - Exponent = 00000, so  $2^4$  positive and  $2^4$  negative

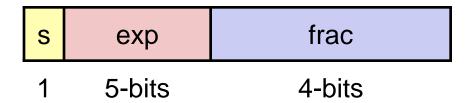
- 10-bit IEEE-like format
  - e = 5 exponent bits
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- What is the exponent bias?  $2^{5-1} 1 = 15$
- How many denormalized numbers are there?
  - Exponent = 00000, so  $2^4$  positive and  $2^4$  negative
- **■** What is the bit pattern of the maximum value number?
- What is the bit pattern of the number closest to zero?

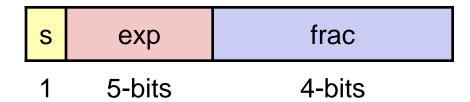
#### 10-bit IEEE-like format

- e = 5 exponent bits
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- What is the exponent bias?  $2^{5-1} 1 = 15$
- How many denormalized numbers are there?
  - Exponent = 00000, so  $2^4$  positive and  $2^4$  negative
- What is the bit pattern of the maximum value number?
  - Sign = 0, Exponent = 11110, frac=1111, so 0111101111
- What is the bit pattern of the number closest to zero?
  - Sign = ?, Exponent = 00000, frac=0001, so ?00000001

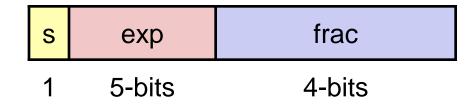
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What is the bit pattern of the smallest positive normal number?

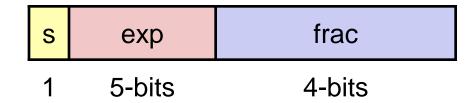
#### 10-bit IEEE-like format

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- What is the bit pattern of the smallest positive normal number?
  - Sign = 0, exp = 00001, frac = 0000; so 0000010000

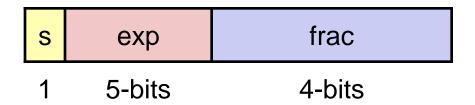
- 10-bit IEEE-like format
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- What is the bit pattern of the smallest positive normal number?
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- **■** What is the value of the smallest positive normal number?

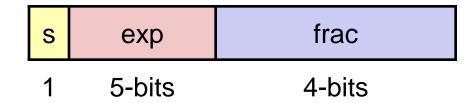
#### ■ 10-bit IEEE-like format

- e = 5 exponent bits
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- What is the bit pattern of the smallest positive normal number?
  - Sign = 0, exp = 00001, frac = 0000; so 0000010000
- **■** What is the value of the smallest positive normal number?
  - Value =  $(-1)^s \times M \times 2^E$
  - S = 0
  - Exponent bias = 15, so E = 1 15 = -14
  - M = 1 + 0 ×  $\frac{1}{2}$  + 0 ×  $\frac{1}{4}$  + 0 ×  $\frac{1}{8}$  + 0 ×  $\frac{1}{16}$  = 1
  - Value =  $(-1)^0 \times 1 \times 2^{-14} = \frac{1}{2^{14}} = 0.00006103515$

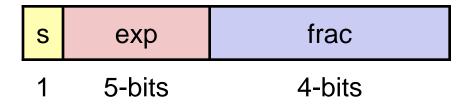
- 10-bit IEEE-like format
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■ Given a 32-bit floating point number, and a 32-bit integer, which can represent more discrete values?

#### ■ 10-bit IEEE-like format

- e = 5 exponent bits
- f = 4 fraction bits



- Given a 32-bit floating point number, and a 32-bit integer, which can represent more discrete values?
  - Both can represent  $2^{32}$  values, but some bit patterns duplicate values, e.g. +0/-0,  $+\infty/-\infty$ , and many NaNs (exponent = 11...1, frac != 00...0)

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### **IEEE Floating Point**

#### IEEE Standard 754

- Established in 1985 as uniform standard for floating point arithmetic
  - Before that, many idiosyncratic formats
- Supported by all major CPUs

#### Driven by numerical concerns

- Nice standards for rounding, overflow, underflow
- Hard to make fast in hardware
  - Numerical analysts predominated over hardware designers in defining standard

# **Floating Point Representation**

#### Numerical Form:

$$(-1)^{s} M 2^{E}$$

- Sign bit s determines whether number is negative or positive
- Significand (mantissa) M normally a fractional value in range [1.0, 2.0)
- **Exponent** *E* weights value by power of two

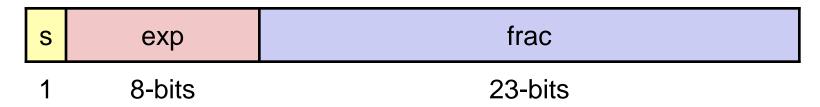
#### Encoding

- S is sign bit s
- exp field encodes E (but is not equal to E)
- frac field encodes M (but is not equal to M)

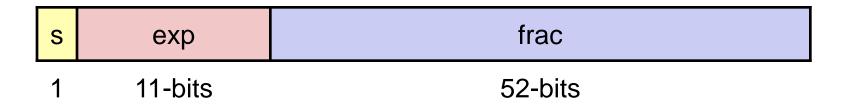
S
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#### **Precisions**

■ Single precision: 32 bits (c type: float)



Double precision: 64 bits (c type: double)



Extended precision: 80 bits (Intel only)

S	exp	frac
1	15-bits	63 or 64-bits

#### **Normalized Values**

- Condition:  $exp \neq 000...0$  and  $exp \neq 111...1$
- **Exponent coded as biased value:** E = Exp Bias
  - Exp: unsigned value of exp field
  - $Bias = 2^{k-1} 1$ , where k is number of exponent bits
    - Single precision: 127 (*exp*: 1...254  $\Rightarrow$  *E*: -126...127)
    - Double precision: 1023 (*exp*: 1...2046  $\Rightarrow$  *E*: -1022...1023)
- Significand coded with implied leading 1:  $M = 1.xxx...x_2$ 
  - xxx...x: bits of frac
- Decimal value of normalized FP representations:
  - Single-precision:  $Value_{10} = (-1)^s \times 1. frac \times 2^{exp-127}$
  - Double-precision:  $Value_{10} = (-1)^s \times 1. frac \times 2^{exp-1023}$

### **Normalized Encoding Example**

■ Value: float F = 15213.0; shift binary point by K bits so that ■ 15213<sub>10</sub> = 11101101101101<sub>2</sub> only one leading 1 bit remains on the left side of the binary point  $= 1.1101101101101_2 \times 2^{13}$ (here, shifted right by 13 bits, so K = 13), then multiply by  $2^{K}$  (here,  $2^{13}$ ) Significand 1.1101101101101, M 1101101101101000000000<sub>2</sub> frac= ■ Exponent (E = Exp - Bias) F 13 Bias =  $Exp = E + Bias = 140 = 10001100_{2}$ 10001100 11011011011010000000000 frac S

#### **Denormalized Values**

- **Condition:** exp = 000...0
- Exponent value: E = -Bias + 1 (instead of E = 0 Bias)
- Significand coded with implied leading 0: M = 0.xxx...x2
  - \*xxx...x: bits of frac
- Cases
  - exp = 000...0, frac = 000...0
    - Represents zero value
    - Note distinct values: +0 and -0 (why?)
  - exp = 000...0,  $frac \neq 000...0$ 
    - Numbers very close to 0.0
    - Lose precision as get smaller
    - Equispaced

### **Special Values**

- Special condition: exp = 111...1
- Case: exp = 111...1, frac = 000...0
  - Represents value ∞ (infinity)
  - Operation that overflows
  - Both positive and negative
  - E.g.,  $1.0/0.0 = -1.0/-0.0 = +\infty$ ,  $1.0/-0.0 = -\infty$
- Case: exp = 111...1,  $frac \neq 000...0$ 
  - Not-a-Number (NaN)
  - Represents case when no numeric value can be determined
  - E.g., sqrt(-1),  $\infty \infty$ ,  $\infty \times 0$

## **Interesting Numbers**

■ Double  $\approx 1.8 \times 10^{308}$ 

{single,double}

Description	exp	frac	Numeric Value		
Zero	0000	0000	0.0		
Smallest Pos. Denorm.	0000	0001	$2^{-\{23,52\}} \times 2^{-\{126,1022\}}$		
■ Single $\approx 1.4 \times 10^{-45}$					
■ Double $\approx 4.9 \times 10^{-324}$					
<ul><li>Largest Denormalized</li></ul>	0000	1111	$(1.0 - \varepsilon) \times 2^{-\{126,1022\}}$		
■ Single $\approx 1.18 \times 10^{-38}$					
■ Double $\approx 2.2 \times 10^{-308}$					
Smallest Pos. Normalized	0001	0000	1.0 x $2^{-\{126,1022\}}$		
<ul><li>Just larger than largest denormalized</li></ul>					
One	0111	0000	1.0		
<ul><li>Largest Normalized</li></ul>	1110	1111	$(2.0 - \varepsilon) \times 2^{\{127,1023\}}$		
Single ≈ 3.4 x 10 <sup>38</sup>					

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### Floating Point in C

#### C Guarantees Two Levels

- •float single precision
- **double** double precision

### Conversions/Casting

- Casting between int, float, and double changes bit representation
- double/float → int
  - Truncates fractional part
  - Like rounding toward zero
  - Not defined when out of range or NaN: Generally sets to TMin
- int → double
  - Exact conversion, as long as int has ≤ 53 bit word size
- int → float
  - Will round according to rounding mode

### **Today: Floating Point**

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## Floating Point Operations: Basic Idea

$$\mathbf{x} +_{\mathbf{f}} \mathbf{y} = \text{Round}(\mathbf{x} + \mathbf{y})$$

$$\mathbf{x} \times_{\mathbf{f}} \mathbf{y} = \text{Round}(\mathbf{x} \times \mathbf{y})$$

#### Basic idea

- First compute exact result
- Make it fit into desired precision
  - Possibly overflow if exponent too large
  - Possibly round to fit into frac

## Rounding

Rounding Modes (illustrate with \$ rounding)

	\$1.40	\$1.60	\$1.50	\$2.50	-\$1.50
Towards zero	\$1	\$1	\$1	\$2	<b>-</b> \$1
Round down (-∞)	\$1	\$1	\$1	\$2	<b>-</b> \$2
■ Round up $(+\infty)$	\$2	\$2	\$2	\$3	<b>-</b> \$1
<ul><li>Nearest Even (default)</li></ul>	\$1	\$2	\$2	\$2	<b>-\$2</b>

### Closer Look at Round-To-Even

#### Default Rounding Mode

- Hard to get any other kind without dropping into assembly
- All others are statistically biased
  - Sum of set of positive numbers will consistently be over- or underestimated

### Applying to Other Decimal Places / Bit Positions

- When exactly halfway between two possible values
  - Round so that least significant digit is even
- E.g., round to nearest hundredth

7.8949999	7.89	(Less than half way)
7.8950001	7.90	(Greater than half way)
7.8950000	7.90	(Half way—round up)
7.8850000	7.88	(Half way—round down)

### **Rounding Binary Numbers**

#### Binary Fractional Numbers

- "Even" when least significant bit is 0
- "Half way" when bits to right of rounding position = 100...2

#### Examples

Round to nearest 1/4 (2 bits right of binary point)

Value	Binary	Rounded	Action	Rounded Value
2 3/32	10.000112	10.002	(<1/2—down)	2
2 3/16	10.00 <mark>110</mark> 2	10.012	(>1/2—up)	2 1/4
2 7/8	10.11 <mark>100</mark> 2	11.002	( 1/2—up)	3
2 5/8	10.10 <mark>100</mark> 2	10.102	( 1/2—down)	2 1/2

### **Scientific Notation Multiplication**

- $\blacksquare$   $(2.5 \times 10^3) \times (3.0 \times 10^2) = ?$
- Compute result by pieces:
  - Sign:  $sign_{left} * sign_{right} = 1*1 = 1$
  - Significand :  $M_{left} * M_{right} = 2.5 * 3.0 = 7.5$
  - Exponent:  $E_{left} + E_{right} = 3 + 2 = 5$

Result:  $7.5 \times 10^5$ 

### **FP Multiplication**

- $\blacksquare$   $(-1)^{s1} M1 2^{E1} \times (-1)^{s2} M2 2^{E2}$
- **Exact Result:**  $(-1)^s M 2^E$ 
  - Sign s: s1 ^ s2
  - Significand *M*: *M1* x *M2*
  - Exponent E: E1 + E2

#### Fixing

- If  $M \ge 2$ , shift M right, increment E
- If *E* out of range, overflow
- Round M to fit frac precision

#### Implementation

Biggest chore is multiplying significands

### **Scientific Notation Addition**

- $\blacksquare$   $(2.5 \times 10^3) + (3.0 \times 10^2) = ?$
- Assume E<sub>left</sub> is larger that E<sub>right</sub>
- Align by decimal point:
  - Significand :

$$\frac{2.5}{+.3}$$

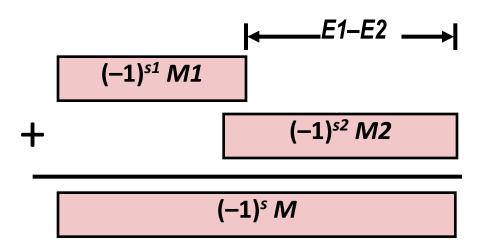
• Exponent :  $E = E_{left} = 3$ 

Result:  $2.8 \times 10^3$ 

# **Floating Point Addition**

- $\blacksquare (-1)^{s1} M1 2^{E1} + (-1)^{s2} M2 2^{E2}$ 
  - **A**ssume *E1* > *E2*
- Exact Result:  $(-1)^s M 2^E$ 
  - ■Sign *s*, significand *M*:
    - Result of signed align & add
  - ■Exponent *E*: *E*1

Get binary points lined up



#### Fixing

- ■If  $M \ge 2$ , shift M right, increment E
- •if M < 1, shift M left k positions, decrement E by k
- ■Overflow if *E* out of range
- ■Round *M* to fit **frac** precision

# **Mathematical Properties of FP Add**

- Compare to those of Abelian Group
  - Commutative?

Yes

- (a + b) = (b + a)
- Associative?

No

- Overflow and inexactness of rounding
- $\bullet$  (3.14+1e10)-1e10 = 0, 3.14+(1e10-1e10) = 3.14

## **Mathematical Properties of FP Mult**

#### Compare to Commutative Ring

• Multiplication Commutative?

Yes

- **Ex:** (1e20\*1e-20) = (1e-20\*1e20)
- Multiplication is Associative?

No

- Possibility of overflow, inexactness of rounding
- Ex: (1e20\*1e20) \*1e-20= inf, 1e20\* (1e20\*1e-20) = 1e20
- Multiplication distributes over addition?

No

- Possibility of overflow, inexactness of rounding
- $\blacksquare$  1e20\*(1e20-1e20) = 0.0, 1e20\*1e20 1e20\*1e20 = NaN

### Summary

- Represents numbers of form M x 2<sup>E</sup>
- One can reason about operations independent of implementation
  - As if computed with perfect precision and then rounded
- Not the same as real arithmetic
  - Violates associativity/distributivity
  - Makes life difficult for compilers & serious numerical applications programmers