

Floating Point

Lecture 3

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Today: Floating Point

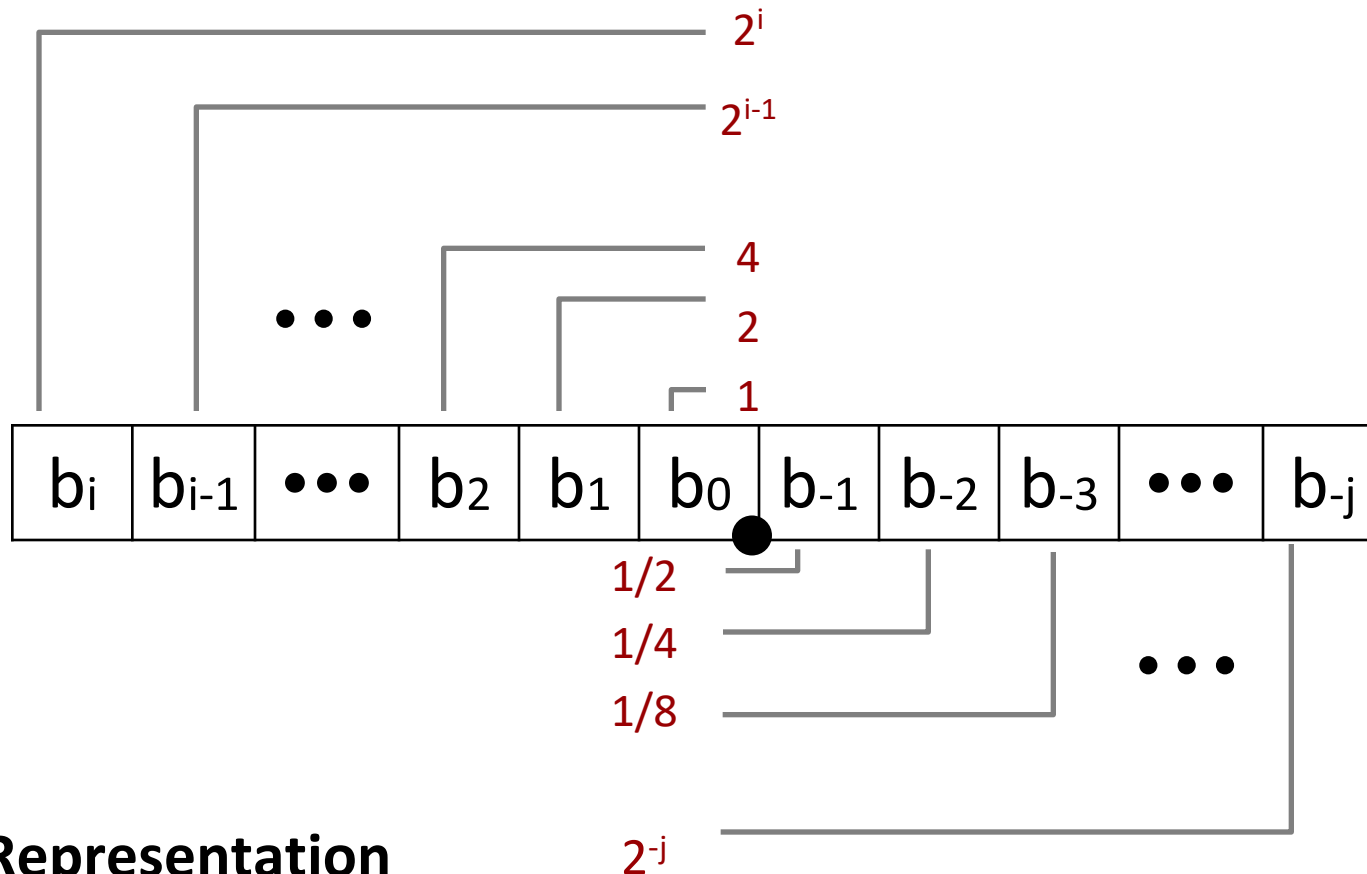
Textbook: [CS:APP3e] 2.4

- **Background: Fractional binary numbers**
- **IEEE floating point standard: Definition**
- **Example and properties**
- **Rounding, addition, multiplication**
- **Floating point in C**
- **Summary**

Fractional binary numbers

- What is 1011.101_2 ?

Fractional Binary Numbers



■ Representation

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

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

Fractional Binary Numbers: Examples

■ Value Representation

$5 \frac{3}{4}$	101.11_2
$2 \frac{7}{8}$	10.111_2
$1 \frac{7}{16}$	1.0111_2

■ Observations

- Divide by 2 by shifting right (unsigned)
- Multiply by 2 by shifting left
- Numbers of form $0.111111..._2$ are just below 1.0
 - $1/2 + 1/4 + 1/8 + \dots + 1/2^i + \dots \rightarrow 1.0$
 - Use notation $1.0 - \epsilon$

Representable Numbers

■ Limitation #1

- Can only exactly represent numbers of the form $x/2^k$
 - Other rational numbers have repeating bit representations
- Value Representation
 - $1/3$ $0.0101010101 [01] \dots_2$
 - $1/5$ $0.001100110011 [0011] \dots_2$
 - $1/10$ $0.0001100110011 [0011] \dots_2$

■ Limitation #2

- Just one setting of binary point within the w bits
 - Limited range of numbers (very small values? very large?)

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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 M normally a fractional value in range $[1.0, 2.0)$.
- Exponent E weights value by power of two

■ Encoding

- MSB 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)

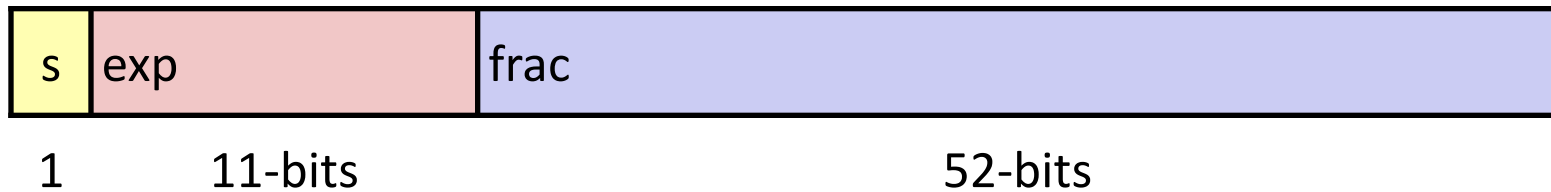


Precision options

- **Single precision: 32 bits**



- **Double precision: 64 bits**



- **Extended precision: 80 bits (Intel only)**



“Normalized” Values

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

- **When: $\text{exp} \neq 000\dots 0$ and $\text{exp} \neq 111\dots 1$**
- **Exponent coded as a biased value: $E = \text{Exp} - \text{Bias}$**
 - Exp: unsigned value of exp field
 - Bias = $2^{k-1} - 1$, where k is number of exponent bits
 - Single precision: 127 (Exp: 1...254, E: -126...127)
 - Double precision: 1023 (Exp: 1...2046, E: -1022...1023)
- **Significand coded with implied leading 1: $M = 1.\text{xxx}\dots\text{x}_2$**
 - xxx...x: bits of frac field
 - Minimum when frac=000...0 ($M = 1.0$)
 - Maximum when frac=111...1 ($M = 2.0 - \epsilon$)
 - Get extra leading bit for “free”

Normalized Encoding Example

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

$$E = \text{Exp} - \text{Bias}$$

■ Value: float $F = 15213.0;$

$$15213_{10} = 11101101101101_2$$

$$= 1.1101101101101_2 \times 2^{13}$$

■ Significand

$$M = 1.\underline{1101101101101}_2$$

$$\text{frac} = \underline{1101101101101}0000000000_2$$

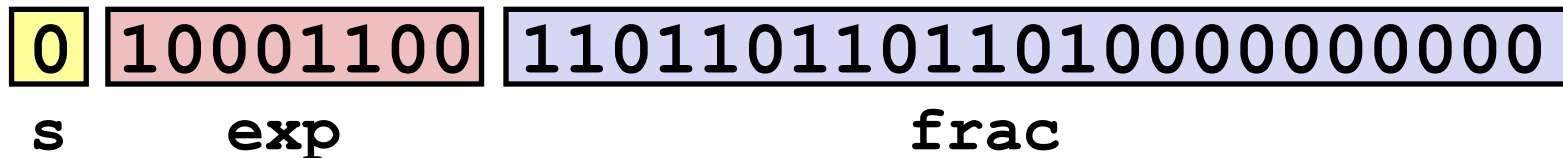
■ Exponent

$$E = 13$$

$$\text{Bias} = 127$$

$$\text{Exp} = 140 = 10001100_2$$

■ Result:



Denormalized Values

$$v = (-1)^s M 2^E$$
$$E = 1 - \text{Bias}$$

- Condition: $\text{exp} = 000\dots 0$
- Exponent value: $E = 1 - \text{Bias}$ (instead of $E = 0 - \text{Bias}$)
- Significand coded with implied leading 0: $M = 0.\text{xxx}\dots\text{x}_2$
 - $\text{xxx}\dots\text{x}$: bits of frac
- Cases
 - $\text{exp} = 000\dots 0, \text{frac} = 000\dots 0$
 - Represents zero value
 - Note distinct values: $+0$ and -0 (why?)
 - $\text{exp} = 000\dots 0, \text{frac} \neq 000\dots 0$
 - Numbers closest to 0.0
 - Equispaced

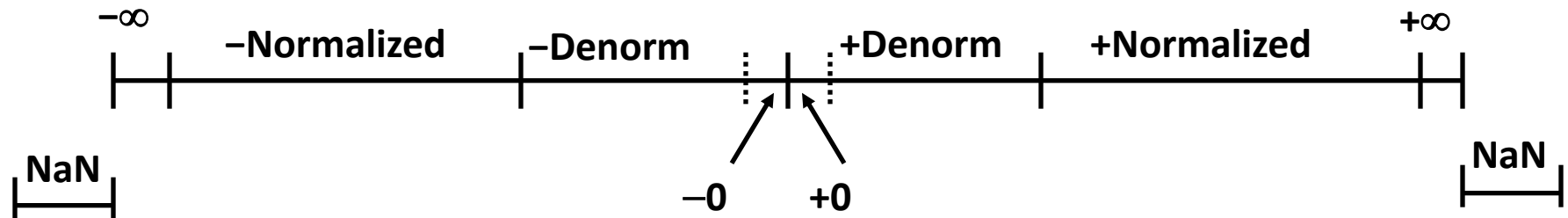
Special Values

- **Condition: $\text{exp} = 111\dots 1$**

- **Case: $\text{exp} = 111\dots 1, \text{frac} = 000\dots 0$**
 - 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: $\text{exp} = 111\dots 1, \text{frac} \neq 000\dots 0$**
 - Not-a-Number (NaN)
 - Represents case when no numeric value can be determined
 - E.g., $\text{sqrt}(-1)$, $\infty - \infty$, $\infty \times 0$

Visualization: Floating Point Encodings



Interesting Numbers

{single, double}

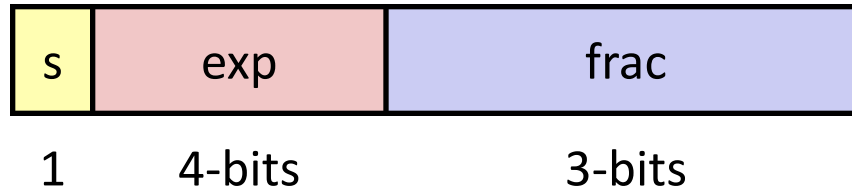
<i>Description</i>	<i>exp</i>	<i>frac</i>	<i>Numeric Value</i>
■ Zero	00...00	00...00	0.0
■ Smallest Pos. Denorm.	00...00	00...01	$2^{-\{23,52\}} \times 2^{-\{126,1022\}}$
<ul style="list-style-type: none"> ■ Single $\approx 1.4 \times 10^{-45}$ ■ Double $\approx 4.9 \times 10^{-324}$ 			
■ Largest Denormalized	00...00	11...11	$(1.0 - \epsilon) \times 2^{-\{126,1022\}}$
<ul style="list-style-type: none"> ■ Single $\approx 1.18 \times 10^{-38}$ ■ Double $\approx 2.2 \times 10^{-308}$ 			
■ Smallest Pos. Normalized	00...01	00...00	$1.0 \times 2^{-\{126,1022\}}$
<ul style="list-style-type: none"> ■ Just larger than largest denormalized 			
■ One	01...11	00...00	1.0
■ Largest Normalized	11...10	11...11	$(2.0 - \epsilon) \times 2^{\{127,1023\}}$
<ul style="list-style-type: none"> ■ Single $\approx 3.4 \times 10^{38}$ ■ Double $\approx 1.8 \times 10^{308}$ 			

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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, with a bias of 7
- the last three bits are the **frac**

■ Same general form as IEEE Format

- normalized, denormalized
- representation of 0, NaN, infinity

Dynamic Range (Positive Only)

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

n: $E = \text{Exp} - \text{Bias}$
d: $E = 1 - \text{Bias}$

closest to zero

largest denorm

smallest norm

closest to 1 below

closest to 1 above

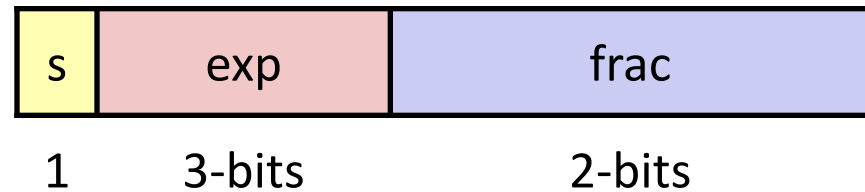
largest norm

s	exp	frac	E	Value
0	0000	000	-6	0
0	0000	001	-6	$1/8 * 1/64 = 1/512$
0	0000	010	-6	$2/8 * 1/64 = 2/512$
...				
0	0000	110	-6	$6/8 * 1/64 = 6/512$
0	0000	111	-6	$7/8 * 1/64 = 7/512$
0	0001	000	-6	$8/8 * 1/64 = 8/512$
0	0001	001	-6	$9/8 * 1/64 = 9/512$
...				
0	0110	110	-1	$14/8 * 1/2 = 14/16$
0	0110	111	-1	$15/8 * 1/2 = 15/16$
0	0111	000	0	$8/8 * 1 = 1$
0	0111	001	0	$9/8 * 1 = 9/8$
0	0111	010	0	$10/8 * 1 = 10/8$
...				
0	1110	110	7	$14/8 * 128 = 224$
0	1110	111	7	$15/8 * 128 = 240$
0	1111	000	n/a	inf

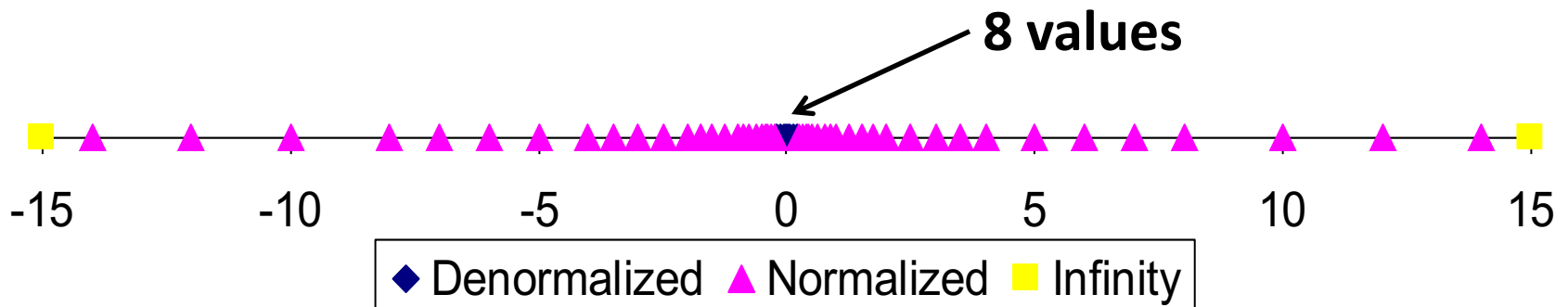
Distribution of Values

■ 6-bit IEEE-like format

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



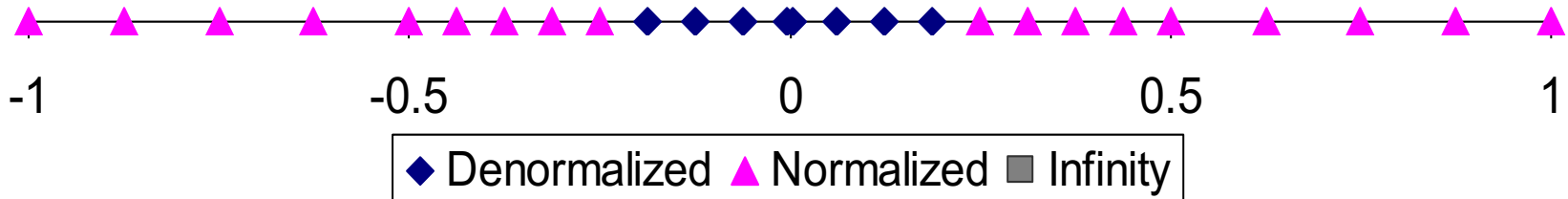
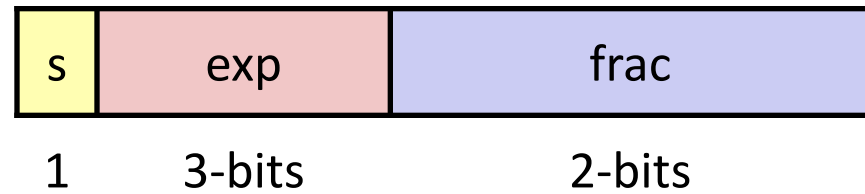
■ Notice how the distribution gets denser toward zero.



Distribution of Values (close-up view)

■ 6-bit IEEE-like format

- $e = 3$ exponent bits
- $f = 2$ fraction bits
- Bias is 3



Special Properties of the IEEE Encoding

- **FP Zero Same as Integer Zero**
 - All bits = 0

- **Can (Almost) Use Unsigned Integer Comparison**
 - Must first compare sign bits
 - Must consider $-0 = 0$
 - NaNs problematic
 - Will be greater than any other values
 - What should comparison yield?
 - Otherwise OK
 - Denorm vs. normalized
 - Normalized vs. infinity

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Floating Point Operations: Basic Idea

- $x +_f y = \text{Round}(x + y)$
- $x \times_f y = \text{Round}(x \times y)$
- **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	-\$1
■ Round down ($-\infty$)	\$1	\$1	\$1	\$2	-\$2
■ Round up ($+\infty$)	\$2	\$2	\$2	\$3	-\$1
■ Nearest Even (default)	\$1	\$2	\$2	\$2	-\$2

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
$2 \frac{3}{32}$	$10.00\textcolor{red}{11}_2$	10.00_2	(< $1/2$ —down)	2
$2 \frac{3}{16}$	$10.00\textcolor{red}{110}_2$	10.01_2	(> $1/2$ —up)	$2 \frac{1}{4}$
$2 \frac{7}{8}$	$10.11\textcolor{red}{100}_2$	11.00_2	($1/2$ —up)	3
$2 \frac{5}{8}$	$10.10\textcolor{red}{100}_2$	10.10_2	($1/2$ —down)	$2 \frac{1}{2}$

FP Multiplication

■ $(-1)^{s1} M1 2^{E1} \times (-1)^{s2} M2 2^{E2}$

■ **Exact Result:** $(-1)^s M 2^E$

- Sign s: $s1 \wedge s2$
- Significand M: $M1 \times M2$
- Exponent E: $E1 + E2$

■ Fixing

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

■ Implementation

- Biggest chore is multiplying significands

Floating Point Addition

$$\blacksquare (-1)^{s1} M1 2^{E1} + (-1)^{s2} M2 2^{E2}$$

- Assume $E1 > E2$

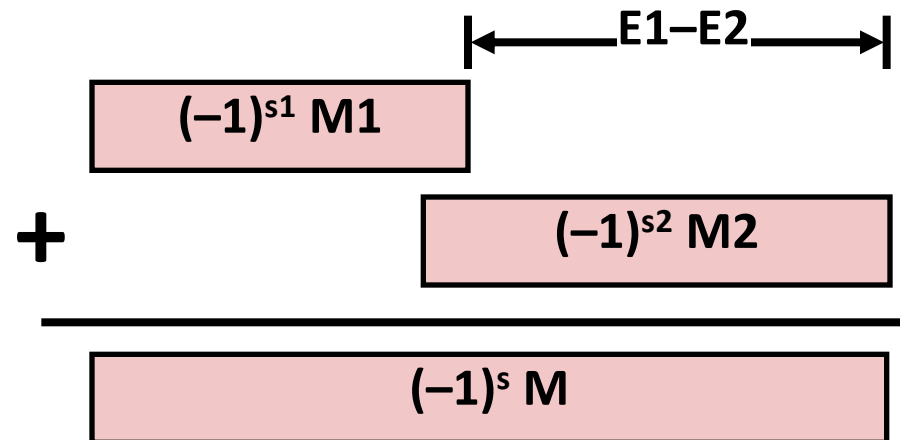
$$\blacksquare \text{Exact Result: } (-1)^s M 2^E$$

- Sign s , significand M :
 - Result of signed align & add
- Exponent E : $E1$

Fixing

- If $M \geq 2$, 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

Get binary points lined up



Mathematical Properties of FP Add

■ Compare to those of Abelian Group

- Closed under addition? **Yes**
 - But may generate infinity or NaN
- Commutative? **Yes**
- Associative? **No**
 - Overflow and inexactness of rounding
 - Ex: $(3.14 + 1e10) - 1e10 = 0$, $3.14 + (1e10 - 1e10) = 3.14$
- 0 is additive identity? **Yes**
- Every element has additive inverse? **Almost**
 - Yes, except for infinities & NaNs

■ Monotonicity

- $a \geq b \Rightarrow a + c \geq b + c$ **Almost**
 - Except for infinities & NaNs

Mathematical Properties of FP Mult

■ Compare to Commutative Ring

- Closed under multiplication? **Yes**
 - But may generate infinity or NaN
- Multiplication Commutative? **Yes**
- Multiplication is Associative? **No**
 - Possibility of overflow, inexactness of rounding
 - Ex: $(1e20 * 1e20) * 1e-20 = \text{inf}$, $1e20 * (1e20 * 1e-20) = 1e20$
- 1 is multiplicative identity? **Yes**
- Multiplication distributes over addition? **No**
 - Possibility of overflow, inexactness of rounding
 - $1e20 * (1e20 - 1e20) = 0.0$, $1e20 * 1e20 - 1e20 * 1e20 = \text{NaN}$

■ Monotonicity

- $a \geq b \ \& \ c \geq 0 \Rightarrow a * c \geq b * c$? **Almost**
 - Except for infinities & NaNs

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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

Summary

- IEEE Floating Point has clear mathematical properties
- Represents numbers of form $M \times 2^E$
- 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