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# **Floating Point**

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# **Today: Floating Point**

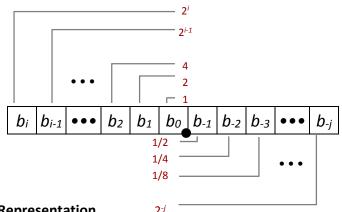
- Background: Fractional binary numbers
- IEEE floating point standard: Definition
- **■** Example and properties
- Rounding, addition, multiplication
- Floating point in C
- Summary

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# **Fractional binary numbers**

■ What is 1011.101<sub>2</sub>?

# **Fractional Binary Numbers**



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

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

## **Fractional Binary Numbers: Examples**

Value	Representation
5 3/4	101.112
2 7/8	10.1112
1 7/16	1.01112

#### Observations

- Divide by 2 by shifting right (unsigned)
- Multiply by 2 by shifting left
- Numbers of form 0.1111111...2 are just below 1.0
  - $1/2 + 1/4 + 1/8 + ... + 1/2^i + ... \rightarrow 1.0$
  - Use notation 1.0 ε

# **Representable Numbers**

#### ■ Limitation #1

- Can only exactly represent numbers of the form x/2<sup>k</sup>
  - Other rational numbers have repeating bit representations

Value	Representation
<b>1/</b> 3	0.0101010101[01]2
<b>1/</b> 5	0.001100110011[0011]2
<b>1/10</b>	0.0001100110011[0011]2

#### ■ Limitation #2

- Just one setting of decimal 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

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8

## **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)

s	exp	frac
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### **Precision options**

■ Single precision: 32 bits



■ Double precision: 64 bits



■ Extended precision: 80 bits (Intel only)

s	ехр	frac
1	15-bits	63 or 64-bits

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### 3 cases based on value of exp

- Normalized
  - When exp isn't all 0s or all 1s
  - Most common

#### Denomalized

- When exp is all 0s
- Different interpretation of E than normalized
- Used for +0 and -0
- (And other numbers close to 0)
- "Special"
  - When exp is all 1s
  - NaN, infinities

### "Normalized" Values

- When: exp ≠ 000...0 and exp ≠ 111...1
- Exponent coded as a biased value: E = Exp Bias
  - Exp: unsigned value exp
  - 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:  $M = 1.xxx...x_2$ 
  - xxx...x: bits of frac
  - Minimum when frac=000...0 (M = 1.0)
  - Maximum when frac=111...1 ( $M = 2.0 \varepsilon$ )
  - Get extra leading bit for "free"

# **Normalized Encoding Example**

- Value: Float F = 15213.0; ■ 15213<sub>10</sub> = 11101101101101<sub>2</sub> = 1.1101101101101<sub>2</sub> x 2<sup>13</sup>
- Significand

M = 1.11011011011012
frac = 1101101101101010000000000000002

Exponent

$$E = 13$$
 $Bias = 127$ 
 $Exp = 140 = 10001100_{2}$ 

■ Result:

### 0 10001100 11011011011010000000000

s exp

frac

13

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### **Denormalized Values**

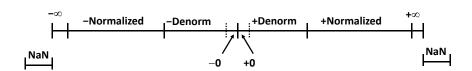
- **Condition:** exp = 000...0
- **Exponent value:** E = 1 Bias
  - (instead of E = 0 Bias)
- Significand coded with implied leading 0: M = 0.xxx...x2
  - \*xxx...x: bits of frac
- Cases
  - exp = 000...0, frac = 000...0
    - Represents zero value
    - Note distinct values: +0 and -0 (why?)
  - exp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

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### **Special Values**

- **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 ≠ 000...0
  - Not-a-Number (NaN)
  - Represents case when no numeric value can be determined
  - E.g., sqrt(-1),  $\infty \infty$ ,  $\infty \times 0$

# **Visualization: Floating Point Encodings**



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### **Tiny Floating Point Example**

s exp frac
1 4-bits 3-bits

#### ■ 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

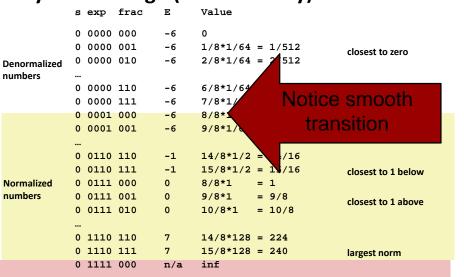
- normalized, denormalized
- representation of 0, NaN, infinity

17

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# **Dynamic Range (Positive Only)**



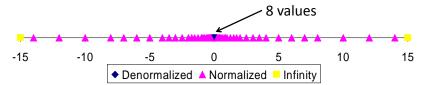
### **Distribution of Values**

#### ■ 6-bit IEEE-like format

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



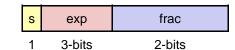
■ Notice how the distribution gets denser toward zero.



# **Distribution of Values (close-up view)**

#### ■ 6-bit IEEE-like format

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





21

### **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**

- $\mathbf{x} +_{\mathbf{f}} \mathbf{y} = \text{Round}(\mathbf{x} + \mathbf{y})$
- $\mathbf{x} \times_{\mathbf{f}} \mathbf{y} = \text{Round}(\mathbf{x} \times \mathbf{y})$
- Basic idea
  - First compute exact result
  - Make it fit into desired precision
    - Possibly overflow if exponent too large
    - Possibly round to fit into frac

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## 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 (+∞)	\$2	\$2	\$2	\$3	-\$1
<ul><li>Nearest Even (default)</li></ul>	\$1	\$2	\$2	\$2	<b>-</b> \$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

1.2349999	1.23	(Less than half way)
1.2350001	1.24	(Greater than half way)
1.2350000	1.24	(Half way—round up)
1.2450000	1.24	(Half way—round down)

25

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## **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	$10.00110_2$	10.012	(>1/2—up)	2 1/4
2 7/8	$10.11100_2$	11.002	( 1/2—up)	3
2 5/8	$10.10100_2$	10.102	( 1/2—down)	2 1/2

int->fp

## **FP Multiplication**

- $\blacksquare$  (-1)<sup>s1</sup> M1 2<sup>E1</sup> x (-1)<sup>s2</sup> M2 2<sup>E2</sup>
- Exact Result: (-1)<sup>s</sup> M 2<sup>E</sup>

Sign s: s1 ^ s2
 Significand M: M1 x M2
 Exponent E: E1 + E2

#### Fixing

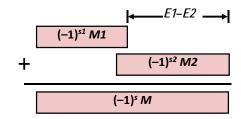
- If  $M \ge 2$ , shift M right, increment E
- If E out of range, overflow
- Round M to fit frac precision

### **■** Implementation

Biggest chore is multiplying significands

### **Floating Point Addition**

- $\blacksquare$  (-1)<sup>51</sup> M1 2<sup>E1</sup> + (-1)<sup>52</sup> M2 2<sup>E2</sup>
  - **A**ssume *E1* > *E2*
- Exact Result: (-1)<sup>s</sup> M 2<sup>E</sup>
  - ■Sign *s*, significand *M*:
    - Result of signed align & add
  - ■Exponent *E*: *E1*



#### 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

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29

### **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

## Some implications

- Order of operations is important
  - **3.14+(1e20-1e20)** versus (3.14+1e20)-1e20
  - 1e20\*(1e20-1e20) versus (1e20\*1e20)-(1e20\*1e20)
- **■** Compiler optimizations impeded
  - E.g., Common sub-expression elimination

```
double x=a+b+c;
double y=b+c+d;
```

May not equal

```
double temp=b+c;
double x=a+temp;
double y=temp+d;
```

### **Floating Point Puzzles**

- For each of the following C expressions, either:
  - Argue that it is true for all argument values
  - Explain why not true

int x = ...:

float f = ...;

double d = ...;

Assume neither

d nor f is NaN

- x == (int)(float) x
- x == (int)(double) x
- f == (float)(double) f
- d == (float) d
- f == -(-f);
- 2/3 == 2/3.0
- 2.0/3==2/3.0
- d < 0.0
- $\Rightarrow$  ((d\*2) < 0.0)
- $d > f \Rightarrow -f > -d$
- d \* d >= 0.0
- (d+f)-d == f

### **Summary**

- IEEE Floating Point has clear mathematical properties
- 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

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33

### **More Slides**

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## **Interesting Numbers**

### {single,double}

			(2211920,000020)
Description	ехр	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 $\approx 2.2 \times 10^{-308}$			
<ul><li>Smallest Pos. Normalized</li></ul>	0001	0000	1.0 x 2 <sup>-{126,1022}</sup>
<ul><li>Just larger than largest deno</li></ul>	rmalized		
<ul><li>One</li></ul>	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>			

## **Mathematical Properties of FP Add**

#### **■** Compare to those of Abelian Group

- Closed under addition?
  - But may generate infinity or NaN
- Commutative?
- Associative?
  - Overflow and inexactness of rounding
- 0 is additive identity?
- Every element has additive inverse
  - Except for infinities & NaNs

#### Monotonicity

- $a \ge b \Rightarrow a+c \ge b+c$ ?
  - Except for infinities & NaNs

37

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## **Mathematical Properties of FP Mult**

### **■** Compare to Commutative Ring

Closed under multiplication?

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

- But may generate infinity or NaN
- Multiplication Commutative?
- Multiplication is Associative?
  - Possibility of overflow, inexactness of rounding
- 1 is multiplicative identity?
- Multiplication distributes over addition?
  - Possibility of overflow, inexactness of rounding

#### Monotonicity

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

## **Creating Floating Point Number**

#### Steps

Normalize to have leading 1

s exp frac
1 4-bits 3-bits

Round to fit within fraction

Postnormalize to deal with effects of rounding

#### ■ Case Study

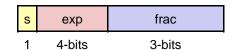
Convert 8-bit unsigned numbers to tiny floating point format

#### **Example Numbers**

128	10000000
14	00001101
33	00010001
35	00010011
138	10001010
63	00111111

39

# **Normalize**



#### ■ Requirement

- Set binary point so that numbers of form 1.xxxxx
- Adjust all to have leading one
  - Decrement exponent as shift left

Value	Binary	Fraction	Exponent
128	10000000	1.0000000	7
14	00001101	1.1010000	3
17	00010001	1.0001000	4
19	00010011	1.0011000	4
138	10001010	1.0001010	7
63	00111111	1.1111100	5

Rounding

1.BBGRXXX

Guard bit: LSB of result -

Sticky bit: OR of remaining bits

Round bit: 1st bit removed

### ■ Round up conditions

- Round = 1, Sticky = 1 → > 0.5
- Guard = 1, Round = 1, Sticky = 0 → Round to even

Value	Fraction	GRS	Incr?	Rounded
128	1.0000000	000	N	1.000
14	1.1010000	100	N	1.101
17	1.0001000	010	N	1.000
19	1.0011000	110	Υ	1.010
138	1.0001010	011	Υ	1.001
63	1.1111100	111	Υ	10.000

41

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### **Postnormalize**

#### Issue

- Rounding may have caused overflow
- Handle by shifting right once & incrementing exponent

Value	Rounded	Ехр	Adjusted	Result
128	1.000	7		128
14	1.101	3		14
17	1.000	4		16
19	1.010	4		20
138	1.001	7		134
63	10.000	5	1.000/6	64

back

42