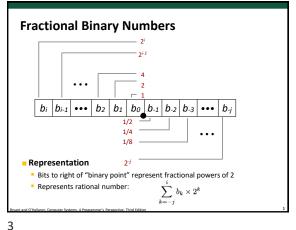
# **Floating Point**

(with contributions from Dr. Bin Ren, William & Mary Computer Science)

# **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>? **1/2 + 1/8 = 5/8** 11 5/8 or 11.625 What is 123.45 in binary? 123 = 1111011 • to get the .45, use repeated multiplication by 2 if product < 1, bit is 0</li> • if product >= 1, bit is 1 and subtract 1 from product .45 \* 2 = .9 \* 2 = (1.8 – 1) \* 2 = 1 (1.6 - 1) \* 2 = 1 (1.2 - 1) \* 2 =1 **1111011.01110** 

# **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)
  - compare 101.11<sub>2</sub> with 10.111<sub>2</sub>
- Multiply by 2 by shifting left
- compare 101.11, with 1011.1,
- Numbers of form 0.111111...2 are just below 1.0
  - $1/2 + 1/4 + 1/8 + ... + 1/2^{i} + ... \rightarrow 1.0$
- notation sometimes seen: 1.0 ε

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# **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

 1/3 0.0101010101[01]...2 · 1/5 0.001100110011[0011]...2

· 1/10 0.0001100110011[0011]...2

### Limitation #2

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

# Representable Numbers Numbers 0.111...11 base 2 represent numbers just below 1 0.111111 base 2 = 63/64 Only finite-length encodings 1/3 and 5/7 cannot be represented exactly Fractional binary notation can only represent numbers that can be written x \* 2y (i.e. 63/64 = 63\*2-6) Otherwise, approximated Increasing accuracy = lengthening the binary representation but still have finite space

Practice

Fractional value of the following binary values:

0.01 =
0.010 =
1.00110 =
11.001101 =

123.45 base 10
Binary value =
FYI also equals:
1.2345 x 10<sup>2</sup> in normalized form
12345 x 10<sup>2</sup> using significand/mantissa/coefficient and exponent

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

IEEE Standard 754

IEEE Standard 754

Established in 1985 as uniform standard for floating point arithmetic

Before that, many idiosyncratic formats

Supported by all major CPUS

Intel-based PCS

Apple

Unix/Linux

Driven by numerical concerns

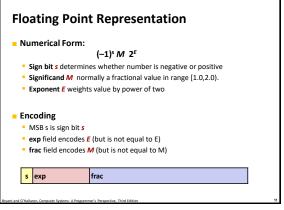
Nice standards for rounding, overflow, underflow

Hard to make fast in hardware

Numerical analysts predominated over hardware designers in defining standard

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**Precision options** Single precision: 32 bits frac s exp 8-bits 23-bits Double precision: 64 bits s exp frac 11-bits 52-bits Extended precision: 80 bits (Intel only) frac s exp 15-bits 63 or 64-bits

11 12

# $v = (-1)^s M 2^E$ **Normalized Values** When: exp ≠ 000...0 and exp ≠ 111...1 Exponent coded as a 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, E: -126...127) Double precision: 1023 (Exp: 1...2046, E: -1022...1023) Significand coded with implied leading 1: M = 1.xxx...x2 xxx...x: bits of frac field Minimum when frac = 000...0 (M = 1.0) Maximum when frac = 111...1 (M = 2.0 – ε) Get extra leading bit for "free"

**Bias Notes** 

- Biasing is done because exponents have to be signed values in order to be able to represent both tiny and huge values, but two's complement, the usual representation for signed values, would make comparison harder.
  - To solve this problem the exponent is biased to put it within an unsigned range suitable for comparison.
  - By arranging the fields so that the sign bit is in the most significant bit position, the biased exponent in the middle, then the mantissa in the least significant bits, the resulting value will be ordered properly, whether it's interpreted as a floating point or integer value. This allows high speed comparisons of floating point numbers using fixed point hardware.
- When interpreting the floating-point number, the bias is subtracted to retrieve the actual exponent.

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# **Significand Notes**

- Represents the fraction, or precision bits of the number.
- It is composed of an implicit (i.e., hidden) leading bit and the fraction bits.
- In order to maximize the quantity of representable numbers, floating-point numbers are typically stored in normalized form.
  - This basically puts the radix point after the first non-zero digit
  - Nice optimization available in base two, since the only possible non-zero digit is 1.
- Thus, we can just assume a leading digit of 1, and don't need to represent it explicitly.
- As a result, the mantissa/significand has effectively 24 bits of resolution, by way of 23 fraction bits.

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Normalized Encoding Example
```

 $v = (-1)^s M 2^E$ E = Exp – Bias

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

1.1101101101101 frac= 

Exponent

13 Bias = 127 10001100. Exp = 140 =

- Result:
- 0 10001100 11011011011010000000000

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### **Normalized Encoding Example 2**

- Value: π, rounded to 24 bits of precision
- sign: 0

Significand

11.0010010000111111011011 (including hidden bit) м 1.10010010000111111011011, (x 21) 10010010000111111011011, frac =

Exponent

Bias = 127 Exp = 128 =

0 10000000 10010010000111111011011

100000002

**Normalized Encoding Practice** 

Value: -π

same as before; only sign bit changes

1 10000000 10010010000111111011011 frac

Value: -78 3/8 (-78.375)

 $1001110.011 = 1.001110011 \times 2^{6} + 127 = 133 \text{ or } 10000101$ 1 10000101 00111001100000000000000000

ехр

Value: 63 11/32 (127.34375)  $111111.01011 = 1.1111101011 \times 2^5 \quad 5 + 127 = 132 \text{ or } 10000100$ 0 10000100 111110101100000000000000

exp frac

Value: -1/64 (-0.015625)

 $0.000001 = 1.0 \times 2^{-6} -6 + 127 = 121 \text{ or } 01111001$ 

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### **Denormalized Values**

- Also called denormal or subnormal numbers
- Values that are very close to zero
  - Fill the "underflow" gap around zero
  - Gradual underflow = numeric values are spaced evenly near 0.0
- Any number with magnitude smaller than the smallest normal number
  - When the exponent field is all zeros
  - E = 1-bias
  - Significand M = f without implied leading 1
  - h = 0 (hidden bit)
- Representation of numeric value 0
  - -0.0 and +0.0 are considered different in some ways and the same in others

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### **Denormalized Values**

- In a normal floating point value, there are no leading zeros in the significand, instead leading zeros are moved to the exponent.
- e.g., 0.0123 would be written as 1.23 \* 10<sup>-2</sup>
- Denormalized numbers are numbers where this representation would result in an exponent that is too small (the exponent usually having a limited range). Such numbers are represented using leading zeros in the significand.

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### **Denormalized Values**

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

- **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 ≠ 000...0
    - Numbers closest to 0.0
    - Equispaced

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# **Denormalized Decoding Example**

- What is the decimal value of the following?
- 0 0000000 000001000011100000000
  - exp
  - sign: 0 (positive)
- we know it's a denormalized value because exp is all 0s
- Exponent

E = 1 - 127 = -126 (same for all denormalized numbers)

Significand

frac = 00000010000111000000000<sub>2</sub> M = 0.0000010000111000000000<sub>2</sub>

- Result:
- 0.00000010000111000000000 x  $2^{-126}$  = 1.0000111 x  $2^{-133}$
- 10000111 x  $2^{-140}$  = 135 x  $2^{-140}$  = 9.69 x  $10^{-41}$

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**Denormalized Encoding Example** 

- Value: -25 15/32 x 2<sup>-132</sup>
  - sign: 1
     power of 2 indicates denormalized number (< -126)</li>

Exponent

Bias = 127 $Exp = 00000000_2$  (same for all denormalized numbers)

E = 1 - 127 = -126

Significand

s = 11001.01111 x 2<sup>-132</sup> (move 6 places left to match -126) M = 0.01100101111000000000000002 (x 2<sup>-126</sup>) frac = 011001011111000000000000002

Result:

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1 00000000 01100101111000000000000

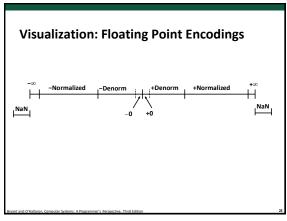
s exp
and O'Hallaron, Computer Systems: A Programmer's Perspective, Third Edition

frac

# 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),  $\infty$   $\infty$ ,  $\infty \times 0$

Bryant and O'Hallaron, Computer Systems: A Programmer's Perspective, Third Editio



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**Interesting Numbers** {single,double} Description frac Numeric Value exp Zero 00...00 00...00 2-{23,52} x 2-{126,1022} 00...00 00...01 Smallest Pos. Denorm. Single ≈ 1.4 x 10<sup>-45</sup> Double ≈ 4.9 x 10<sup>-324</sup> (1.0 –  $\epsilon$ ) x 2<sup>-{126,1022}</sup> Largest Denormalized 00...00 11...11 Single ≈ 1.18 x 10<sup>-38</sup> Double ≈ 2.2 x 10<sup>-308</sup> Smallest Pos. Normalized 00...01 00...00 1.0 x  $2^{-\{126,1022\}}$  Just larger than largest denormalized 01...11 00...00  $(2.0 - \epsilon) \times 2^{\{127,1023\}}$  Largest Normalized 11...10 11...11 Single ≈ 3.4 x 10<sup>38</sup> Double ≈ 1.8 x 10<sup>308</sup>

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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 £rac |
| Same general form as IEEE Format |
| normalized, denormalized |
| representation of 0, NaN, infinity

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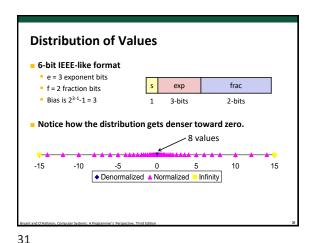
Normal	lize	s	exp	frac
		1	4-bits	3-bits
Requiren	nent			
<ul><li>Set bina</li></ul>	ry point so that nu	umbers of	form 1.xxxxx	:
<ul> <li>Adjust a</li> </ul>	II to have leading	one		
<ul> <li>Decre</li> </ul>	ement exponent a	s shift left		
<ul> <li>Incre</li> </ul>	ment exponent as	shift right	:	
Value	Binary	Fractio	n Ex	ponent
128	10000000	1.000	0000 7	
13	00001101	1.101	.0000 3	
17	00010001	1.000	1000 4	
19	00010011	1.001	1000 4	
138	10001010	1.000	1010 7	
63	00111111	1.111	1100 5	

**Dynamic Range (Positive Only)**  $v = (-1)^s M 2^E$ s exp frac n: E = Exp - Bias d: E = 1 - Bias0 0000 000 0 0000 001 1/8\*1/64 = 1/512 closest to zero Denormalized 0 0000 010 numbers 6/8\*1/64 = 6/512 7/8\*1/64 = 7/512 8/8\*1/64 = 8/512 9/8\*1/64 = 9/512 0 0000 110 0 0000 111 0 0001 000 largest denorm smallest norm 0 0001 001 0 0110 110 14/8\*1/2 = 14/16 15/8\*1/2 = 15/16 8/8\*1 = 1 9/8\*1 = 9/8 10/8\*1 = 10/8 closest to 1 below 0 0110 111 0 0111 000 -1 Normalized 0 0111 001 closest to 1 above 0 0111 010 14/8\*128 = 224 15/8\*128 = 240 0 1110 110 largest norm 0 1110 111 0 1111 000 n/a

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Distribution of Values (close-up view)

■ 6-bit IEEE-like format

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

■ 3-bits

2-bits

□ Denormalized ■ Normalized ■ Infinity

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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?
Cherwise OK
Denorm vs. normalized
Normalized vs. infinity

Floating Point

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Floating Point Operations: Basic Idea

| x + f y = Round (x + y)

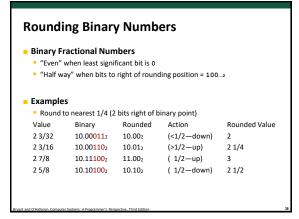
| x × f y = Round (x × y)

| 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 (-∞) \$1 \$1 \$2 -\$2 \$1 Round up (+∞) \$2 \$2 \$2 \$3 -\$1 Nearest Even (default) \$1 \$2 -\$2

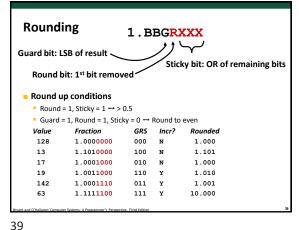
35 36

### 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 (Half way-round down) 7.88



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Issue					
- Roundir	ig may have caus	ed overflo	w		
- Handle	by shifting right o	nce & inc	rementing expo	onent	
Value	Rounded	Ехр	Adjusted	Result	
128	1.000	7		128	
13	1.101	3		13	
17	1.000	4		16	
19	1.010	4		20	
142	1.001	7		144	
63	10.000	5	1.000/6	64	

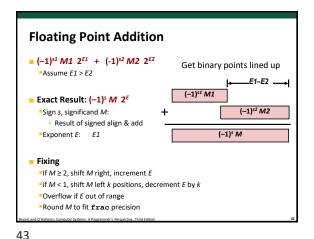
FP Multiplication					
■ (-1) <sup>s1</sup> M1 2 <sup>E1</sup> x	(-1) <sup>52</sup> M2 2 <sup>E2</sup>				
Exact Result: (-1	) <sup>s</sup> M 2 <sup>E</sup>				
Sign s:	s1 ^ s2				
Significand M:	M1 x M2				
Exponent E:	E1 + E2				
Fixing					
If M ≥ 2, shift M ri	ght, increment E				
If E out of range, or	overflow				
Round M to fit fr	ac precision				
<ul> <li>Implementation</li> </ul>					
<ul> <li>Biggest chore is m</li> </ul>	nultiplying significands				
Bryant and O'Hallaron, Computer Systems: A Program	nmer's Perspective, Third Edition	41			

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FP Multiplication Example
What is the product of the following?
Sign
 1^1 = 0
                               x 1.011
1 110
                               11 100
Exponent
                              + 1110 000
 E = 29 + 113 + 1 = 143 + 127 = 270 (1 0000 1110 - overflow)
                           = 1.001101 x 21
Significand
 frac =
         0 00001110 001101000000000000000000
```

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**FP Addition Example** What is the sum of the following? 0 01111101 0000000000000000011000 .01 0000 0000 0000 0000 0011 000 Sign + 111 .00 0000 0000 0000 0000 0000 000 sign with larger exp = 0 111 .01 0000 0000 0000 0000 0011 000 = 1.1101 0000 0000 0000 0000 0011 x 2<sup>2</sup> Exponent = 1.1101 0000 0000 0000 0000 010 x 2<sup>2</sup>  $E = 1000\,0001$  (E2, or 2 + 127 = 129) Significand frac = 1101 0000 0000 0000 0000 0102 (after round to even) 

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Floating Point

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Example and properties

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Floating point in C

Summary

**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 • int → float · Cannot overflow; will round according to rounding mode int/float → double Exact conversion, as long as int has ≤ 53 bit word size float/double → int · Truncates fractional part; like rounding toward zero Not defined when out of range or NaN: Generally sets to TMin double → float · Can overflow (range smaller); may be rounded (precision smaller)

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**Floating Point Puzzles** For each of the following C expressions, either: Argue that it is true for all argument values Explain why not true F \* x == (int)(float) x T · x == (int)(double) x T • f == (float) (double) f int x = ...; F · d == (double)(float) d float f = ...;  $T \cdot f == -(-f);$ double d = ...; F • 2/3 == 2/3.0  $T \cdot d < 0.0 \Rightarrow ((d*2) < 0.0)$ Assume neither d nor f is NaN T · d > f ⇒ -f > -d  $T \cdot d \cdot d >= 0.0$  $F \cdot (d+f) - d == f$ 

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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