Floating Point

Arquitectura de Computadores

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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
- Practice problem

Introduction

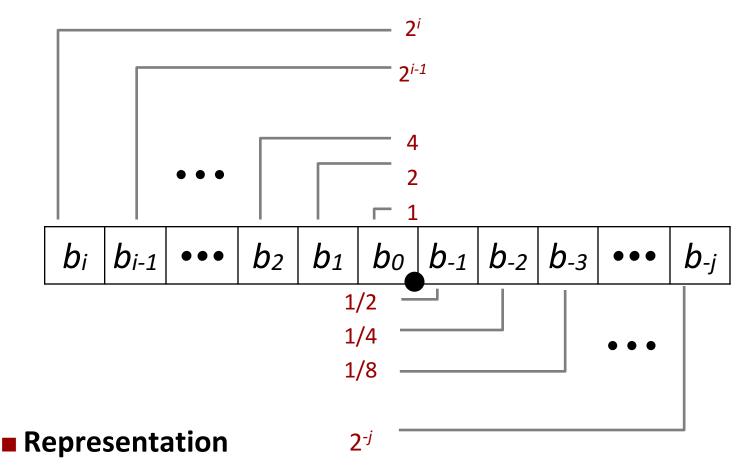
- A floating-point representation encodes rational numbers of the form $V = x \times 2^y$
- It is useful for performing computations involving
 - very large numbers (|V| ≫ 0)
 - numbers very close to 0 (|V| ≪ 1)
 - and more generally as an approximation to real arithmetic
- What issues come up when trying to devise a data representation for floating point numbers?
 - It turns out these issues are more complicated than representing integers

Fractional binary numbers

■ A first step in understanding floating-point numbers is to consider binary numbers having fractional values

■ What is 1011.101₂?

Fractional binary numbers



- Bits to right of "binary point" represent fractional powers of 2
- Represents rational number: $\sum_{k=-i}^{i} b_k imes 2^k$

Fractional binary numbers

- A first step in understanding floating-point numbers is to consider binary numbers having fractional values
- What is 1011.101₂?
- The binary representation of 11.625

$$1 \times 2^{3} + 1 \times 2^{1} + 1 \times 2^{0} + 1 \times 2^{-1} + 1 \times 2^{-3} =$$

$$8 + 2 + 1 + 1/2 + 1/8 =$$

$$11.625$$

Fractional binary numbers: Examples

Value

Representation

$$5.75 = 53/4$$
 101.11_2
 $2.875 = 27/8$ 10.111_2
 $1.4375 = 17/16$ 1.0111_2

Observations

- Divide by 2 by shifting binary point left (unsigned)
- Multiply by 2 by shifting binary point right
- Numbers of form 0.111111...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^k
 - Other rational numbers have repeating bit representations

Value	Representation
1/3	0.01010101[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
 - Not efficient for representing very large or very small numbers

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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 (s = 1) or positive (s = 0)
- Significand M normally a fractional value in range [1.0,2.0).
- Exponent E weights value by power of two

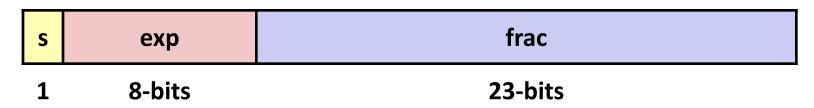
Encoding

- 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	ехр	frac
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Precision options

Single precision: 32 bits (float type in C)



Double precision: 64 bits (double type in C)



Extended precision: 80 bits (Intel only)

S	ехр	frac
1	15-bits	63 or 64-bits

Normalized values

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

The common case, 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: $M = 1.xxx...x_2$
 - xxx...x: bits of frac field
 - Minimum when frac = 000...0 (M = 1.0)
 - Maximum when frac = 111...1 (M = 2.0ε)
 - Get extra leading bit for "free"

Normalized encoding example

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

- **Value:** float f = 15213.0;
 - $15213_{10} = 11101101101101_2$ = $1.1101101101101_2 \times 2^{13}$

Significand

$$M = 1.101101101101_2$$

frac= $110110110110101000000000_2$ (23 bits)

Exponent

$$E = 13$$
 $Bias = 127$
 $Exp = 140 = 10001100_{2}$ (8 bits)

■ Result:

0 10001100 110110110110100000000

s exp frac

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
- Case: *exp* = 000...0, *frac* = 000...0
 - Represents zero value
 - Note distinct values: +0 and -0 (why?)
- Case: exp = 000...0, frac ≠ 000...0
 - Numbers closest to 0.0
 - Equispaced

Denormalized values

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

 $E = 1 - Bias$

- **■** 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), $\infty \infty$, $\infty \times 0$

Visualization: Floating point encodings

Normalized



Denormalized



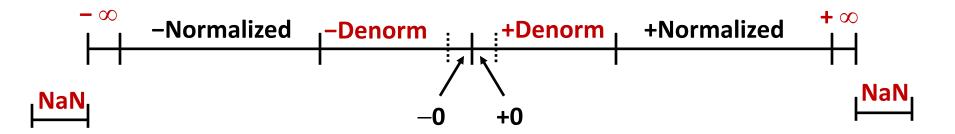
Infinity



NaN

S	11111111	≠ 0

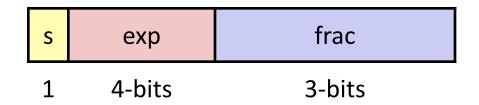
Visualization: Floating point encodings



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- **■** Practice problem

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 $2^4 1 = 7$
- The last three bits are the frac

Same general form as IEEE Format

- Normalized, denormalized
- Representation of 0, NaN, infinity

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

Dynamic range (positive only)

				\ .		n: E = Exp - Bias
	s	exp	frac	E	Value	d: E = 1 - Bias
	0	0000	000	-6	0	
	0	0000	001	-6	1/8*1/64 = 1/512	closest to zero
Denormalized	0	0000	010	-6	2/8*1/64 = 2/512	
numbers						
	0	0000	110	-6	6/8*1/64 = 6/512	
	0	0000	111	-6	7/8*1/64 = 7/512	largest denormalized
	0	0001	000	-6	8/8*1/64 = 8/512	smallest normalized
	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	closest to 1 below
Normalized	0	0111	000	0	8/8*1 = 1	
numbers	0	0111	001	0	9/8*1 = 9/8	closest to 1 above
	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	largest normalized
	0	1111	000	n/a	inf	

Interesting numbers

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

{single,double}

Description	ехр	frac	Numeric Value
Zero	0000	0000	0.0
Smallest positive denorm.	0000	0001	$2^{-\{23,52\}} \times 2^{-\{126,1022\}}$
■ Single $\approx 1.4 \times 10^{-45}$			
■ Double $\approx 4.9 \times 10^{-324}$			
Largest denormalized	0000	1111	$(1.0 - \varepsilon) \times 2^{-\{126,1022\}}$
Single ≈ 1.18 x 10 ⁻³⁸			
■ Double $\approx 2.2 \times 10^{-308}$			
Smallest positive normalized	0001	0000	1.0 x $2^{-\{126,1022\}}$
Just larger than largest denorm	nalized		
One	0111	0000	1.0
Largest Normalized	1110	1111	$(2.0 - \varepsilon) \times 2^{\{127,1023\}}$
Single ≈ 3.4 x 10 ³⁸			

Special properties of the IEEE encoding

- **■** Floating-point zero same as integer zero
 - All bits are equal to 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, including infinity
 - What should comparison yield?
 - Otherwise OK
 - Denormalized 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

Rounding

- For a value x, find the "closest" matching value x' that can be represented in the desired floating-point format
 - One key problem is to define the direction to round a value that is halfway between two possibilities

IEEE floating-point rounding modes

	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 (+∞)	2	2	2	3	-1
Round to nearest, ties to 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 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

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

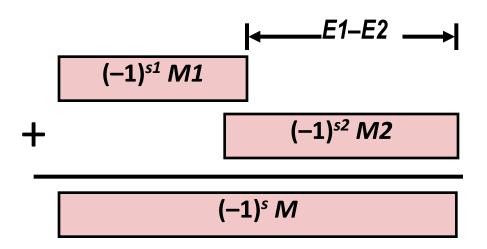
Implementation

Biggest inconvenience is multiplying significands

Floating point addition

- - **A**ssume *E1* > *E2*
- Exact Result: $(-1)^s M 2^E$
 - ■Sign *s*, significand *M*:
 - Result of signed align & add
 - Exponent *E*: *E1*

Get binary points lined up



Fixing

- If $M \ge 2$, normalize shift M right, increment E
- •if M < 1, normalize 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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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

Floating Point in C: Exercise

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

```
\bigcup x == (int)(float) x
\square x == (int) (double) x
\Box f == (float)(double) f
d == (double) (float) d
\Box f == -(-f);
\square 2/3 == 2/3.0
\Box d > f \Rightarrow -f > -d
\Box d * d >= 0.0
\Box (d+f)-d == f
```

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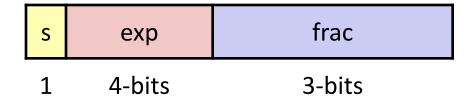
Summary

- IEEE Floating Point has clear mathematical properties
- Represents numbers of form (-1)^s M 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

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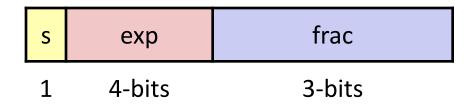
Practice problem



■ Convert 8-bit unsigned numbers to tiny floating point format

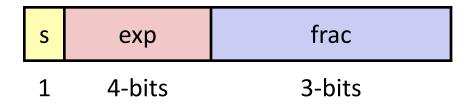
Value	Binary
128	10000000
15	00001101
33	00010001
35	00010011
138	10001010
63	00111111

Practice problem



- Steps:
- Normalize to have leading 1
- 2. Round to fit within fraction
- 3. Postnormalize to deal with effects of rounding

Step 1: 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	1000000	1.0000000	7
15	00001101	1.1010000	3
17	00010001	1.0001000	4
19	00010011	1.0011000	4
138	10001010	1.0001010	7
63	00111111	1.1111100	5

Step 2: Rounding

1.BBGRXXX

Guard bit: LSB of result

Round bit: 1st bit removed

Sticky bit: OR of remaining bits

Round up conditions

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

Value	Fraction	GRS	Incr?	Rounded
128	1.0000000	000	N	1.000
15	1.1010000	100	N	1.101
17	1.0001000	010	N	1.000
19	1.0011000	110	Y	1.010
138	1.0001010	011	Y	1.001
63	1.1111100	111	Y	10.000

Step 3: Postnormalize

Issue

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

Value	Rounded	Exp	Adjusted	Result
128	1.000	7		128
15	1.101	3		15
17	1.000	4		16
19	1.010	4		20
138	1.001	7		134
63	10.000	5	1.000/6	64