

Floating Point

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Giving credit where credit is due

- Most of slides for this lecture are based on slides created by Drs. Bryant and O'Hallaron, Carnegie Mellon University.
- I have modified them and added new slides.

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Topics

- IEEE Floating Point Standard
- Rounding
- Floating Point Operations
- Mathematical properties

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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

```
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`
- `d < 0.0` \Rightarrow `((d*2) < 0.0)`
- `d > f` \Rightarrow `-f < -d`
- `d * d >= 0.0`
- `(d+f) - d == f`

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IEEE Floating Point

IEEE Standard 754

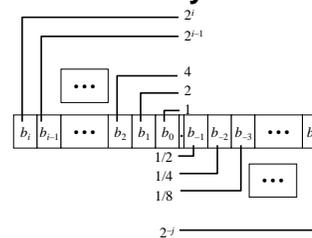
- Established in 1985 as uniform standard for floating point arithmetic
 - Before that, many idiosyncratic formats
- Supported by all major CPUs

Driven by Numerical Concerns

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

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Fractional Binary Numbers



Representation

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

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

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Frac. Binary Number Examples

Value	Representation
5 3/4	101.11 ₂
2 7/8	10.111 ₂
63/64	0.111111 ₂

Observations

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

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Representable Numbers

Limitation

- Can only exactly represent numbers of the form $x/2^k$
- Other 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

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Floating Point Representation

Numerical Form

- $-1^s M 2^E$
 - Sign bit s determines whether number is negative or positive
 - Significant M normally a fractional value in range [1.0, 2.0).
 - Exponent E weights value by power of two

Encoding



- MSB is sign bit
- exp field encodes E
- frac field encodes M

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Floating Point Precisions

Encoding



- MSB is sign bit
- exp field encodes E
- frac field encodes M

Sizes

- Single precision: 8 exp bits, 23 frac bits
 - 32 bits total
- Double precision: 11 exp bits, 52 frac bits
 - 64 bits total
- Extended precision: 15 exp bits, 63 frac bits
 - Only found in Intel-compatible machines
 - Stored in 80 bits
 - > 1 bit wasted

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“Normalized” Numeric Values

Condition

- exp \neq 000...0 and exp \neq 111...1

Exponent coded as *biased* value

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

- Exp : unsigned value denoted by exp
 - Bias: Bias value
 - Single precision: 127 (Exp: 1...254, E: -126...127)
 - Double precision: 1023 (Exp: 1...2046, E: -1022...1023)
 - in general: $\text{Bias} = 2^{e-1} - 1$, where e is number of exponent bits

Significant coded with implied leading 1

$$M = 1.\text{xxxx}\dots\text{x}_2$$

- xxxx...x: bits of frac
- Minimum when 000...0 ($M = 1.0$)
- Maximum when 111...1 ($M = 2.0 - \epsilon$)
- Get extra leading bit for “free”

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Normalized Encoding Example

Value

- Float $F = 15213.0$;
 - $15213_{10} = 11101101101101_2 = 1.1101101101101_2 \times 2^{13}$

Significant

$$M = 1.1101101101101_2$$

$$\text{frac} = \underline{1101101101101}000000000_2$$

Exponent

$$E = 13$$

$$\text{Bias} = 127$$

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

Floating Point Representation (from Lecture 2):

Hex:	4	6	6	D	B	4	0	0
Binary:	0100	0110	0110	1101	1011	0100	0000	0000
140:	100	0110	0					
15213:	1110		1101	1011	01			

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Denormalized Values

Condition

- $\text{exp} = 000\dots 0$

Value

- Exponent value $E = -\text{Bias} + 1$
- Significand value $M = 0.\text{xxx}\dots\text{x}_2$
 - $\text{xxx}\dots\text{x}$: bits of frac

Cases

- $\text{exp} = 000\dots 0, \text{frac} = 000\dots 0$
 - Represents value 0
 - Note that have distinct values +0 and -0
- $\text{exp} = 000\dots 0, \text{frac} \neq 000\dots 0$
 - Numbers very close to 0.0
 - Lose precision as get smaller
 - "Gradual underflow"

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Special Values

Condition

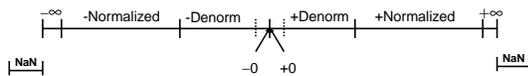
- $\text{exp} = 111\dots 1$

Cases

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

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Summary of Floating Point Real Number Encodings



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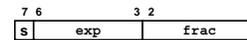
Tiny Floating Point Example

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



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Values Related to the Exponent

Exp	exp	E	2^E	
0	0000	-6	1/64	(denorms)
1	0001	-6	1/64	
2	0010	-5	1/32	
3	0011	-4	1/16	
4	0100	-3	1/8	
5	0101	-2	1/4	
6	0110	-1	1/2	
7	0111	0	1	
8	1000	+1	2	
9	1001	+2	4	
10	1010	+3	8	
11	1011	+4	16	
12	1100	+5	32	
13	1101	+6	64	
14	1110	+7	128	
15	1111	n/a		(inf, NaN)

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Dynamic Range

	s	exp	frac	E	Value
	0	0000	000	-6	0
	0	0000	001	-6	$1/8 * 1/64 = 1/512$ ← closest to zero
	0	0000	010	-6	$2/8 * 1/64 = 2/512$
Denormalized numbers
	0	0000	110	-6	$6/8 * 1/64 = 6/512$
	0	0000	111	-6	$7/8 * 1/64 = 7/512$ ← largest denorm
	0	0001	000	-6	$8/8 * 1/64 = 8/512$ ← smallest norm
	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 numbers	0	0111	000	0	$8/8 * 1 = 1$
	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 norm
	0	1111	000	n/a	inf

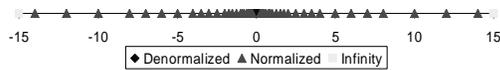
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Distribution of Values

6-bit IEEE-like format

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

Notice how the distribution gets denser toward zero.



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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



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Interesting Numbers

Description	exp	frac	Numeric Value
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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Special Properties of 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

Conceptual View

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

Rounding Modes (illustrate with \$ rounding)

	\$1.40	\$1.60	\$1.50	\$2.50	-\$1.50
■ Zero	\$1	\$1	\$1	\$2	-\$1
■ Round down (\rightarrow)	\$1	\$1	\$1	\$2	-\$2
■ Round up (\leftarrow)	\$2	\$2	\$2	\$3	-\$1
■ Nearest Even (default)	\$1	\$2	\$2	\$2	-\$2

Note:

1. Round down: rounded result is close to but no greater than true result.
2. Round up: rounded result is close to but no less than true result.

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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 under-estimated

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)

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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...

Examples

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

Value	Binary	Rounded	Action	Rounded Value
2 3/32	10.00011 ₂	10.00 ₂	(<1/2—down)	2
2 3/16	10.00110 ₂	10.01 ₂	(>1/2—up)	2 1/4
2 7/8	10.11100 ₂	11.00 ₂	(1/2—up)	3
2 5/8	10.10100 ₂	10.10 ₂	(1/2—down)	2 1/2

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FP Multiplication

Operands

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

Exact Result

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

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

Fixing

- If $M \geq 2$, shift M right, increment E
- If E out of range, overflow
- Round M to fit ϵ_{prec} precision

Implementation

- Biggest chore is multiplying significands

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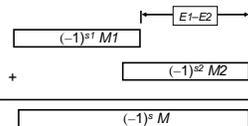
FP Addition

Operands

$$(-1)^{s1} M1 2^{E1}$$

$$(-1)^{s2} M2 2^{E2}$$

- Assume $E1 > E2$



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 ϵ_{prec} precision

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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
- 0 is additive identity? YES
- Every element has additive inverse ALMOST
 - Except for infinities & NaNs

Monotonicity

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

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Math. 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
- 1 is multiplicative identity? YES
- Multiplication distributes over addition? NO
 - Possibility of overflow, inexactness of rounding

Monotonicity

- $a < b \ \& \ c > 0 \Rightarrow a*c < b*c$? 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 between int, float, and double changes numeric values
- Double or float to int
 - Truncates fractional part
 - Like rounding toward zero
 - Not defined when out of range
 - » Generally saturates to TMin or TMax
- int to double
 - Exact conversion, as long as int has 53 bit word size
- int to float
 - Will round according to rounding mode

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Answers to Floating Point Puzzles

```
int x = ...;
float f = ...;
double d = ...;
```

Assume neither
d nor f is NAN

- `x == (int)(float) x` No: 24 bit significand
- `x == (int)(double) x` Yes: 53 bit significand
- `f == (float)(double) f` Yes: increases precision
- `d == (float) d` No: loses precision
- `f == -(-f)` Yes: Just change sign bit
- `2/3 == 2/3.0` No: $2/3 \neq 0$
- `d < 0.0 ⇒ ((d*2) < 0.0)` Yes!
- `d > f ⇒ -f < -d` Yes!
- `d * d >= 0.0` Yes!
- `(d+f) - d == f` No: Not associative

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Ariane 5

- Exploded 37 seconds after liftoff
- Cargo worth \$500 million

Why

- Computed horizontal velocity as floating point number
- Converted to 16-bit integer
- Worked OK for Ariane 4
- Overflowed for Ariane 5
 - Used same software



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Summary

IEEE Floating Point Has Clear Mathematical Properties

- Represents numbers of form $M \times 2^E$
- 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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