CSCE 230J Computer Organization

Floating Point

Dr. Steve Goddard goddard@cse.unl.edu

http://cse.unl.edu/~goddard/Courses/CSCE230J

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

Assume neither d nor f is NaN

int x = ...;

float f = ...;
double 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

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 * d >= 0.0

• (d+f)-d == f

IEEE Floating Point

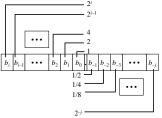
IEEE Standard 754

- 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

Fractional Binary Numbers



Representation

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

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

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Frac. Binary Number Examples

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

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 + ... + 1/2ⁱ + ... → 1.0
 - •Use notation 1.0 − ε

Representable Numbers

- Can only exactly represent numbers of the form x/2^k
- Other numbers have repeating bit representations

Representation 0.0101010101[01]...2 1/5 0.001100110011[0011]...2 1/10 0.0001100110011[0011]...2

Floating Point Representation

Numerical Form

- -1° M 2^E
- •Sign bit s determines whether number is negative or positive

frac

- Significand *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

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
 - Extended precision: 15 exp bits, 63 frac bits Only found in Intel-compatible machines
 - Stored in 80 bits
 - » 1 bit wasted

"Normalized" Numeric Values

Condition

■ exp ≠ 000...0 and exp ≠ 111...1

Exponent coded as biased value

- E = Exp 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: Bias = 2e-1 1, where e is number of exponent bits

Significand coded with implied leading 1

- ●Minimum when 000...0 (M = 1.0)
- •Maximum when 111...1 (M = 2.0 − ε) •Get extra leading bit for "free"

Normalized Encoding Example

Value

Float F = 15213.0;

■ 15213₁₀ = 11101101101101₂ = 1.1101101101101₂ X 2¹³

Significand

M = 1.<u>1101101101101</u>₂ frac=

Exponent

Bias = 127

140 = 10001100₂ Exp =

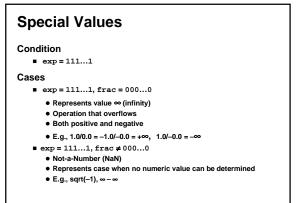
Floating Point Representation (Class 02):

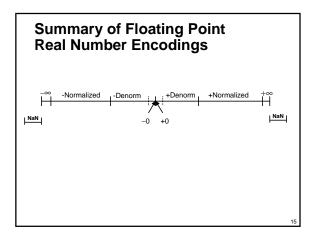
100 0110 0

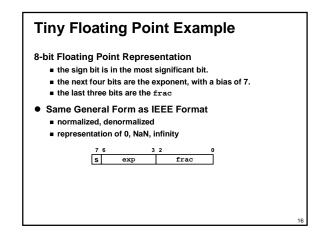
15213: *1*110 1101 1011 01

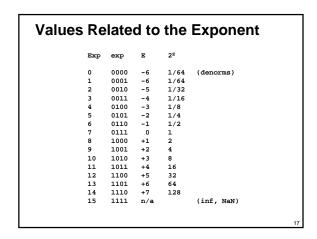
Denormalized Values Condition ■ exp = 000...0 Value ■ Exponent value E = -Bias + 1 ■ Significand value M = 0.xxx...x₂ ● xxx...x: bits of frac Cases ■ exp = 000...0, frac = 000...0 ● Represents value 0 ● Note that have distinct values +0 and -0 ■ exp = 000...0, frac ≠ 000...0 ● Numbers very close to 0.0

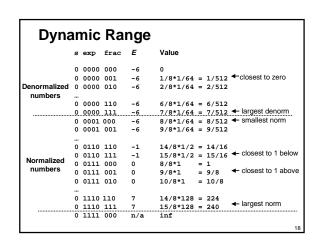
Lose precision as get smaller"Gradual underflow"



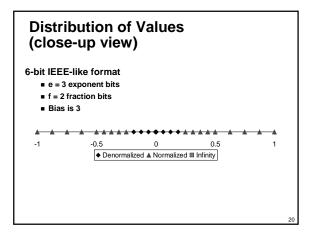








Distribution of Values 6-bit IEEE-like format • e = 3 exponent bits • f = 2 fraction bits • Bias is 3 Notice how the distribution gets denser toward zero. -15 -10 -5 0 5 10 15 • Denormalized A Normalized Infinity



Interesting Numbers Numeric Value 00...00 00...00 0.0 Smallest Pos. Denorm. 00...00 00...01 2- {23,52} X 2- {126,1022} ■ Single ≈ 1.4 X 10⁻⁴⁵ ■ Double $\approx 4.9 \times 10^{-324}$ Largest Denormalized 00...00 11...11 (1.0 - ε) X 2- {126,1022} ■ Single ≈ 1.18 X 10⁻³⁸ ■ Double ≈ 2.2 X 10⁻³⁰⁸ Smallest Pos. Normalized 00...01 00...00 1.0 X 2- {126,1022} ■ Just larger than largest denormalized 01...11 00...00 1.0 Largest Normalized 11...10 11...11 (2.0 – ε) X 2^{127,1023} ■ Single ≈ 3.4 X 10³⁸ ■ Double ≈ 1.8 X 10³⁰⁸

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

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 (-∞)	\$1	\$1	\$1	\$2	-\$2
■ Round up (+∞)	\$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.

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)

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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)</td>
 2

 2 3/16
 10.001102
 10.012
 (>1/2—up)
 2 1/4

 2 7/8
 10.111002
 11.002
 (1/2—up)
 3

 2 5/8
 10.101002
 10.102
 (1/2—down)
 2 1/2

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FP Multiplication

Operands

(-1)^{s1} M1 2^{E1} * (-1)^{s2} M2 2^{E2}

Exact Result

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

- Sign s: s1 ^ s2
- Significand M: M1 * M2
- Exponent *E*: *E*1 + *E*2

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

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FP Addition Operands (-1)s1 M1 2E1 E1-E2 (-1)^{s2} M2 2^{E2} ■ Assume *E1* > *E*2 **Exact Result** $(-1)^s M 2^E$ ■ 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

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 ⇒ a+c ≥ b+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

■ Round M to fit frac precision

- YES
- Multiplication Commutative?Multiplication is Associative?
- NO
- Possibility of overflow, inexactness of rounding
- 1 is multiplicative identity? YE
- Multiplication distributes over addition? NO
- Possibility of overflow, inexactness of rounding

Monotonicity

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

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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)(double) x • f == (float)(double) f • d == (float) d

x == (int)(float) x

No: 24 bit significand Yes: 53 bit significand Yes: increases precision No: loses precision

• f == -(-f); • 2/3 == 2/3.0 Yes: Just change sign bit No: 2/3 == 0

• $d < 0.0 \Rightarrow ((d*2) < 0.0)$ • d > f ⇒-f < -d • d * d >= 0.0

Yes! Yes! Yes!

• (d+f)-d == f

No: Not associative

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



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