Floating Point I CSE 351 Autumn 2022

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"IMITATOR" SUN NUMBERS GOLD (DONOTUSE) WAIT	ARTHENON 2.9299372 S	HODOX ("EVERY 7 YEARS", "SCHENCE HODOX ("EVERY 7 YEARS", "SCHENCE SAYS THERE ARE 7", ETC) HODOX UNEXPLORED , 7	IF YOU ENCOUNTER A NUMBER HIGHER THAN THIS, YOU'RE NOT DOING REAL MATH 9 10 8 9 10 1 EST J PRIME

http://xkcd.com/899/

Relevant Course Information

- hw5 due Wednesday, hw6 due Friday
- Don't change your poll answers after-the-fact!
 - Graded on completion; misrepresents your understanding
- Lab 1a due tonight at 11:59 pm
 - Submit pointer.c and lab1Asynthesis.txt
 - Make sure there are no lingering printf statements in your code!
 - Make sure you submit *something* to Gradescope before the deadline and that the file names are correct
 - Can use late days to submit up until Wed 11:59 pm
- Lab 1b due next Monday (10/17)
 - Submitaisle_manager.c, store_client.c, and lab1Bsynthesis.txt

Lab 1b Aside: C Macros

- C macros basics:
 - Basic syntax is of the form: #define NAME expression
 - Allows you to use "NAME" instead of "expression" in code
 - Does naïve copy and replace *before* compilation everywhere the characters "NAME" appear in the code, the characters "expression" will now appear instead
 - NOT the same as a Java constant
 - Useful to help with readability/factoring in code
- You'll use C macros in Lab 1b for defining bit masks
 - See Lab 1b starter code and Lecture 4 slides (card operations) for examples

Reading Review

- Terminology:
 - normalized scientific binary notation
 - trailing zeros
 - sign, mantissa, exponent ↔ bit fields S, M, and E
 - float, double
 - biased notation (exponent), implicit leading one (mantissa)
 - rounding errors
- Questions from the Reading?

Review Questions

- * Convert 11.375₁₀ to normalized binary scientific notation 8+2+1+6.25+0.125 $2^{3}+2^{4}+2^{9}+2^{-2}+2^{-3} = 1011.011, = 1.011011 \times 2^{3}$
- What is the value encoded by the following floating point number?

• bias =
$$2^{\frac{8}{W}-1}-1=2^{\frac{7}{2}}-1=12^{\frac{7}{2}}$$

- exponent = \vec{E} bias = 2^{7} 127 = 128 127 = 1
- mantissa = $1.M = 1.110...0_2$

 $(-1)^{\circ} \times 1.11_{2} \times 2^{1} = 11.1_{2} = [+3.5]$

Number Representation Revisited

- What can we represent in one word?
 - Signed and Unsigned Integers
 - Characters (ASCII)
 - Addresses
- How do we encode the following:
 - Real numbers (*e.g.*, 3.14159)
 - Very large numbers (*e.g.*, 6.02×10²³)
 - Very small numbers (*e.g.*, 6.626×10⁻³⁴)
 - Special numbers (e.g., ∞, NaN)

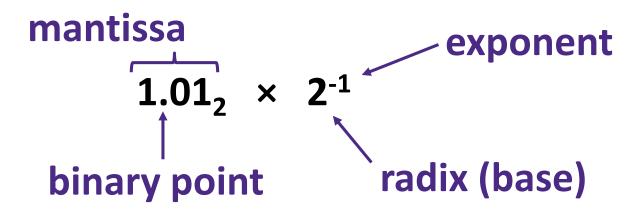
_Floating Point
Point

Floating Point Topics

- * IEEE floating-point standard
- Floating-point operations and rounding
- Floating-point in C

- There are many more details that we won't cover
 - It's a 58-page standard...

Binary Scientific Notation (Review)



- Normalized form: exactly one digit (non-zero) to left of binary point
- Computer arithmetic that supports this called floating point due to the "floating" of the binary point
 - Declare such variable in C as float (or double)

IEEE Floating Point

- ✤ IEEE 754 (established in 1985)
 - Standard to make numerically-sensitive programs portable
 - Specifies two things: representation scheme and result of floating point operations
 - Supported by all major CPUs
- Driven by numerical concerns
 - Scientists/numerical analysts want them to be as real as possible
 - Engineers want them to be easy to implement and fast competing souls
 - Scientists mostly won out:
 - Nice standards for rounding, overflow, underflow, but...
 - Hard to make fast in hardware
 - Float operations can be an order of magnitude slower than integer ops FLOPs used in computer benchmarks

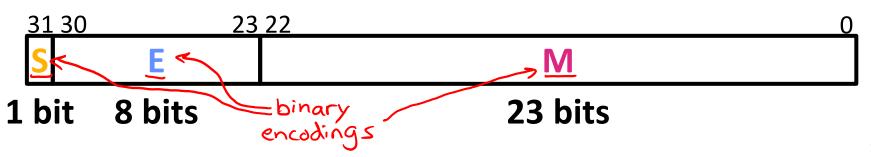
Floating Point Encoding (Review)

- Use normalized, base 2 scientific notation:
 - Value: ±1 × Mantissa × 2^{Exponent}
 - Bit Fields: $(-1)^{S} \times 1.M \times 2^{(E-bias)}$
- * Representation Scheme: (3 separate fields within 32 bits)

Sign bit (0 is positive, 1 is negative)

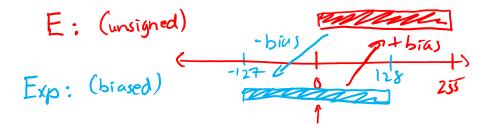
Mantissa (a.k.a. significand) is the fractional part of the number in normalized form and encoded in bit vector M

Exponent weights the value by a (possibly negative) power of 2 and encoded in the bit vector E



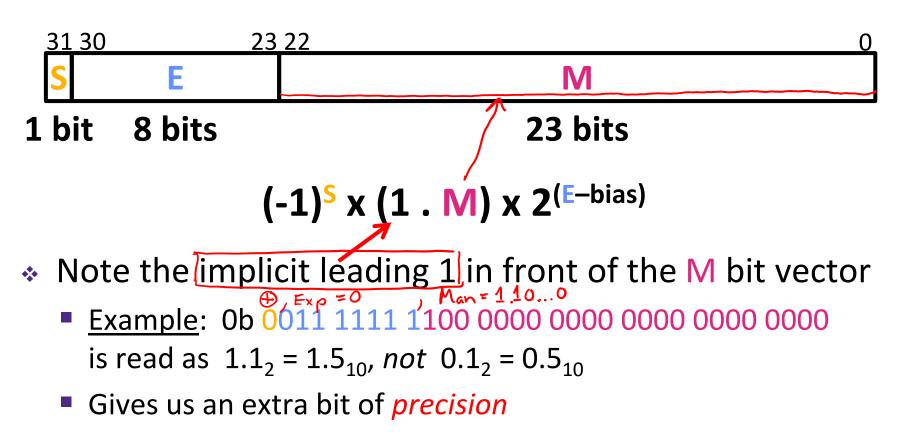
The Exponent Field (Review)

- ♦ Use biased notation w=8, can encode 2⁸=256 exponents
 - Read exponent as unsigned, but with bias of 2^{w-1}-1 = 127
 - Representable exponents roughly ½ positive and ½ negative
 - $Exp = E bias \leftrightarrow E = Exp + bias$
 - Exponent 0 (Exp = 0) is represented as $E = 0b 0111 1111 = 2^{7} 1$



- Why biased?
 - Now it's a sign-and-magnitude representation!
 - Makes floating point arithmetic easier (somewhat compatible with two's complement hardware)

The Mantissa (Fraction) Field (Review)



Mantissa "limits"

$$> 2^{E_{x}p} \times 1.0.0 = 2^{E_{x}p}$$

- Low values near M = 0b0...0 are close to 2^{Exp}
- High values near M = 0b1...1 are close to 2^{Exp+1} $2^{Exp} \times 1.1$ $1 = 2^{Exp}(2-2^{-23}) = 2^{Exp+1} - 2^{Exp-23}$

Normalized Floating Point Conversions

- ♦ FP → Decimal
 - Append the bits of M to implicit leading 1 to form the mantissa.
 - 2. Multiply the mantissa by 2^{E-bias} .
 - **3**. Multiply the sign (-1)^S.
 - Multiply out the exponent by shifting the binary point.
 - 5. Convert from binary to decimal.

♦ Decimal → FP

- Convert decimal to binary.
- Convert binary to normalized scientific notation.
- 3. Encode sign as S (0/1).
- Add the bias to exponent and encode E as unsigned.
- 5. The first bits after the leading 1 that fit are encoded into M.

Practice Question

Convert the decimal number -7.375 = -1.11011 x 2²
into floating point representation.

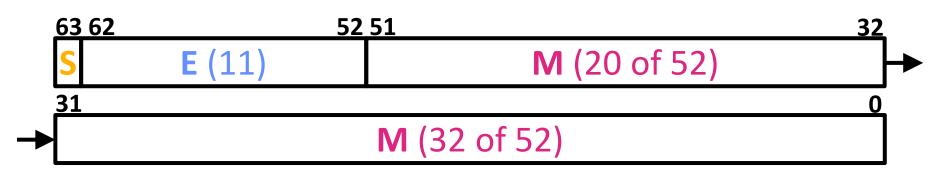
S = 1, E = 2 + 127 = 129 = 0610000001, M = 06100140...00611000001110100...0 = 0x COEC 00001

Precision and Accuracy

- Accuracy is a measure of the difference between the actual value of a number and its computer representation
- Precision is a count of the number of bits in a computer word used to represent a value
 - Capacity for accuracy
- High precision permits high accuracy but doesn't guarantee it
 - Example: float pi = 3.14; will be represented using all 24 bits of the mantissa (highly precise), but is only an approximation (not accurate)

Need Greater Precision?

Double Precision (vs. Single Precision) in 64 bits



- C variable declared as double
- Exponent bias is now $2^{10}-1 = 1023$, bias = $2^{\omega -1}-1$
- Advantages: greater precision (larger mantissa), greater range (larger exponent)
- Disadvantages: more bits used, slower to manipulate

L> Exp=-127

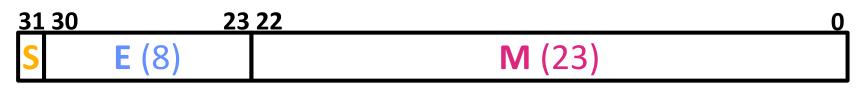
29 zeros

Current Limitations

- ^{Exp=128} م Largest magnitude we can represent? <u>E=061111 1111 , M=061...1</u>
- Smallest magnitude we can represent? E= 06000 000, M=060...0
 - Limited range due to width of E field
- What happens if we try to represent $2^0 + 2^{-30}$?= 1.0.01
 - Rounding due to limited *precision*: stores 2⁰
- There is a need for *special cases*
 - How do we represent the value zero? $0 \neq \pm 1.M \times 2^{E-bias}$
 - What about ∞ and NaN? ???

Summary

Floating point approximates real numbers:



- Handles large numbers, small numbers, special numbers
- Exponent in biased notation (bias = 2^{w-1} 1)
 - Size of exponent field determines our representable range
 - Outside of representable exponents is overflow and underflow
- Mantissa approximates fractional portion of binary point
 - Size of mantissa field determines our representable *precision*
 - Implicit leading 1 (normalized) except in special cases
 - Exceeding length causes *rounding*

Preview Question

* Find the sum of the following binary numbers in normalized scientific binary notation: 0 match exponents(2) sum mantissas $0 1.01_2 \times 2^{\cancel{0}} + 1.11_2 \times 2^2$ (3) normalize 1.0001×2^2 1.0001×2^2 1.00001×2^3