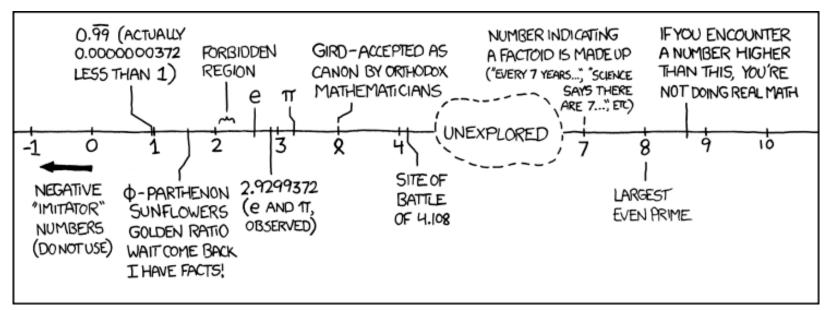
Floating Point I

CSE 351 Winter 2021

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http://xkcd.com/899/

Administrivia

- hw5 due Wednesday, hw6 due Friday
 Malay No Class
- Lab 1a due tonight at 11:59 pm
 - Submit pointer.c and lab1Areflect.txt
 - Make sure you submit *something* to Gradescope before the deadline and that the file names are correct
 - Can use late day tokens to submit up until Mon 11:59 pm
- Section worksheet due tonight at 11:59 pm!
- Lab 1b due next Friday (1/22)
 - Submit aisle_manager.c, store_client.c, and lab1Breflect.txt

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
- Floating Point Simulator
 - https://www.h-schmidt.net/FloatConverter/IEEE754.html
- Questions from the Reading?

Review Questions

* Convert 11.375₁₀ to normalized binary scientific $2^{-3} = 125$ notation $2^{3} + 2' + 2^{\circ} + 2^{-2} + 2^{-3} = 1011 \circ 011 z$

- - bias = $2^{w-1} 1 = 127$ $(-1)^{o} \times 1 10 = \times 2^{v} 5$
 - exponent = $E bias |_{28} |_{22} = 1$ |
 - mantissa = 1.M

2'+2° + 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
$\overline{\leq}$

Floating Point Topics

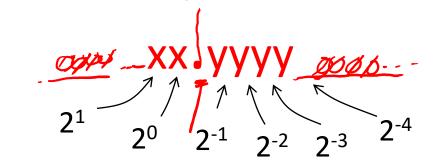
- * Fractional binary numbers
- * IEEE floating-point standard
- Floating-point operations and rounding
- Floating-point in C

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

Representation of Fractions

 "Binary Point," like decimal point, signifies boundary between integer and fractional parts:

Example 6-bit representation:



* <u>Example</u>: $10.1010_2 = 1 \times 2^1 + 1 \times 2^{-1} + 1 \times 2^{-3} = 2.625_{10}$

Representation of Fractions

* "Binary Point," like decimal point, signifies boundary between integer and fractional parts:

Example 6-bit
representation:
$$2^{1}$$
 2^{0} 2^{-1} 2^{-2} 2^{-3} 2^{-4}

- In this 6-bit representation: *
 - $\sim \sim \sim \sim$ What is the encoding and value of the smallest (most negative) number?
 - What is the encoding and value of the largest (most positive) number?
 - What is the smallest number greater than 2 that we can represent?

$$|1.111 = 4 - 2^{-4} = 3.9375$$

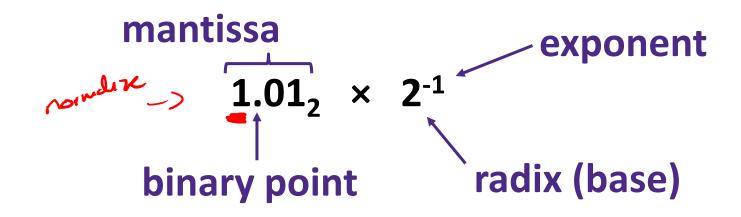
$$2' + 2^{0} + \frac{2^{-1} + 2^{2} + 2^{3} + 2^{4}}{1000}$$

$$|0.0000 = 10.0001 = 2.0625$$

$$\frac{1}{2^{4}}$$
8

8

Scientific Notation (Binary)



- 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)

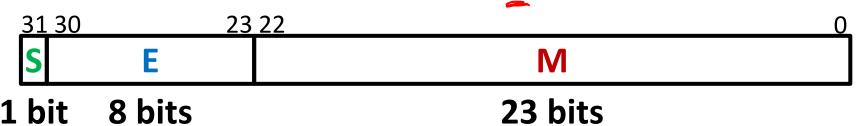
IEEE Floating Point

- ✤ 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
 - 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



Floating Point Encoding

- Use normalized, base 2 scientific notation:
 - Value: ±1 × Mantissa × 2^{Exponent}
- (s, E, M) Bit Fields: $(-1)^{S} \times 1.M \times 2^{(E-bias)}$
 - Representation Scheme:
 - Sign bit (0 is positive, 1 is negative) -> 5
 - 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

- Use biased notation
- 7 Floot. W= 8 Read exponent as unsigned, but with bias of 2^{w-1}-1 = 127
 - Representable exponents roughly ½ positive and ½ negative

= 127

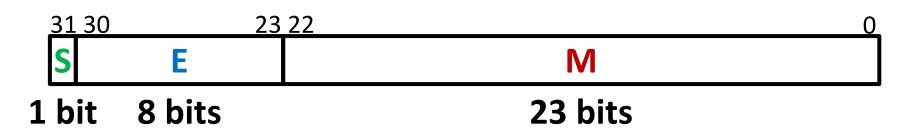
 $E_{77}p = 127 - 127$ $E_{77}p = 128 - 127 \rightarrow E = 00 1000 0000$ $-63 = 64 - 127 \rightarrow E = 00 0100 0000$

- $Exp = E bias \leftrightarrow E = Exp + bias$
 - Exponent 0 (Exp = 0) is represented as E = 0b 0111 1111

Why biased?

- Makes floating point arithmetic easier
- Makes somewhat compatible with two's complement hardware

The Mantissa (Fraction) Field



Note the implicit leading 1 in front of the M bit vector

- Gives us an extra bit of *precision*
- Mantissa "limits"
 - Low values near M = 0b0...0 are close to 2^{Exp}
 - High values near M = 0b1...1 are close to 2^{Exp+1} 2^{I+P}

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

- 1. 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

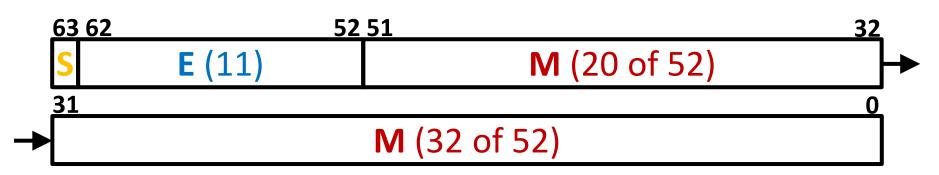
Convert the decimal number -7.375 into floating point representation = -(4+2+1+0.25+0.15) $= - 111.011_{2}$ $(-1)_{10}^{1} \times (1.11011_{2} \times 2^{2})$ 5=1 M= (10)10...\$ E=2+127=129 Challenge Question Find the sum of the following binary numbers in normalized scientific binary notation: 0.0101 +1.1100 $1.01_2 \times 2^0 + 1.11_2 \times 2^2$ 0.0101×22 + 1.11 ×22 10.000 $10.0001_2 \times 2^2$ 1.00001×2^{3}

Precision and Accuracy

- Precision is a count of the number of bits in a computer word used to represent a value
 - Capacity for accuracy
- Accuracy is a measure of the difference between the actual value of a number and its computer representation
 - High precision permits high accuracy but doesn't guarantee it. It is possible to have high precision but low accuracy.
 - Example: float pi = 3.14; 3.14
 - pi 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 $2^{10}-1 = 1023$ $2^{\omega-1}-1 \rightarrow \omega = 1$
- Advantages: greater precision (larger mantissa), greater range (larger exponent)
- Disadvantages: more bits used, slower to manipulate

expanded of -127

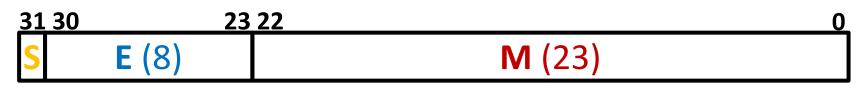
> M bits

Current Limitations

- Smallest magnitude we can represent? E ↓ M JU Ø
 - Limited range due to width of E field
- What happens if we try to represent $2^0 + 2^{-30}$?
 - Rounding due to limited *precision*: stores 2⁰
- There is a need for *special cases*
 - How do we represent the value zero?
 - 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*