Computer Systems & Drogramming Lecture #04 -Floating Point KOC VERSITY Aykut Erdem // Koç University // End

Recap: Bitwise Operators

- You're already familiar with many operators in C:
 - Arithmetic operators: +, -, *, /, %
 - Comparison operators: ==, !=, <, >, <=, >=
 - Logical Operators: &&, ||, !

• Bitwise operators:

- Logical operators: $\&, |, \sim, \uparrow,$
- Bit shift operators: <<, >>

Plan For Today

- Representing real numbers
- Fixed Point
- Floating Point
- Example and Properties
- Floating Point Arithmetic
- Floating Point in C

Disclaimer: Slides for this lecture were borrowed from

- -Nick Troccoli's Stanford CS107 class
- -Randal E. Bryant and David R. O'Hallaron's CMU 15-213 class

<u>COMP201 Topic 2</u>: How can a computer represent real numbers in addition to integer numbers?

Learning Goals

Understand the design and compromises of the floating point representation, including:

- Fixed point vs. floating point
- How a floating point number is represented in binary
- Issues with floating point imprecision
- Other potential pitfalls using floating point numbers in programs

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Real Numbers

- We previously discussed representing integer numbers using two's complement.
- However, this system does not represent real numbers such as 3/5 or 0.25.
- How can we design a representation for real numbers?

Real Numbers

Problem: unlike with the integer number line, where there are a finite number of values between two numbers, there are an *infinite* number of real number values between two numbers!

Integers between 0 and 2: 1

Real Numbers Between 0 and 2: 0.1, 0.01, 0.001, 0.0001, 0.00001,...

We need a fixed-width representation for real numbers. Therefore, by definition, we will not be able to represent all numbers.

Real Numbers

Problem: every number base has un-representable real numbers.

Base 10: $1/6_{10} = 0.166666666..._{10}$ **Base 2:** $1/10_{10} = 0.00011001100110011..._2$

Therefore, by representing in base 2, *we will not be able to represent all numbers*, even those we can exactly represent in base 10.

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• Idea: Like in base 10, let's add binary decimal places to our existing number representation.



• Idea: Like in base 10, let's add binary decimal places to our existing number representation.

1 0 1 1 0 1 1 8s 4s 2s 1s 1/2s 1/4s 1/8s

• **Pros**: arithmetic is easy! And we know exactly how much precision we have.

• **Problem**: we have to fix where the decimal point is in our representation. What should we pick? This also fixes us to 1 place per bit.

. 0 1 1 0 0 1 1

1/2s 1/4s 1/8s ...

1 0 1 1 0 1 1 1 16s 8s 4s 2s 1s 1/2s 1/4s

• **Problem:** we have to fix where the decimal point is in our representation. What should we pick? This also fixes us to 1 place per bit.



To be able to store both these numbers using the same fixed point representation, the bitwidth of the type would need to be at least 207 bits wide!

Let's Get Real

What would be nice to have in a real number representation?

- Represent widest range of numbers possible
- Flexible "floating" decimal point
- Represent scientific notation numbers, e.g. 1.2 x 10⁶
- Still be able to compare quickly
- Have more predictable over/under-flow behavior

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IEEE Floating Point

Let's aim to represent numbers of the following scientific-notation-like format:

$\chi * 2^{\gamma}$

With this format, 32-bit floats represent numbers in the range $\sim 1.2 \times 10^{-38}$ to $\sim 3.4 \times 10^{-38}$! Is every number between those representable? **No**.

IEEE Single Precision Floating Point



Exponent

s exponent (8 bits)	fraction (23 bits)
Exponent (Binary)	Exponent (Base 10)
1111111	?
11111110	?
11111101	?
11111100	?
•••	?
00000011	?
00000010	?
0000001	?
0000000	?

Exponent

s exponent (8 bits)	fraction (23 bits)
Exponent (Binary)	Exponent (Base 10)
1111111	RESERVED
11111110	?
11111101	?
11111100	?
•••	?
00000011	?
00000010	?
0000001	?
0000000	RESERVED

Exponent

s exponent (8 bits)	fraction (23 bits)
Exponent (Binary)	Exponent (Base 10)
1111111	RESERVED
11111110	127
11111101	126
11111100	125
•••	•••
00000011	-124
00000010	-125
0000001	-126
0000000	RESERVED



s exponent (8 bits) fraction (23 bits)	S	exponent (8 bits)	fraction (23 bits)

- The exponent is **not** represented in two's complement.
- Instead, exponents are sequentially represented starting from 000...1 (most negative) to 111...10 (most positive). This makes bit-level comparison fast.
- Actual value = binary value 127 ("bias")

11111110	254 - 127 = 127
11111101	253 - 127 = 126
•••	
00000010	2 - 127 = -125
0000001	1 - 127 = -126

Fraction



• We could just encode whatever x is in the fraction field. But there's a trick we can use to make the most out of the bits we have.

An Interesting Observation

In Base 10:

 $42.4 \times 10^5 = 4.24 \times 10^6$

 $324.5 \times 10^5 = 3.245 \times 10^7$

 $0.624 \times 10^5 = 6.24 \times 10^4$

We tend to adjust the exponent until we get down to one place to the left of the decimal point.

In Base 2:

 $10.1 \times 2^5 = 1.01 \times 2^6$ $1011.1 \times 2^5 = 1.0111 \times 2^8$ $0.110 \times 2^5 = 1.10 \times 2^4$ **Observation:** in base 2, this means there is always a 1 to the left of the decimal point!

Fraction



 $\chi * 2^{y}$

- We can adjust this value to fit the format described previously. Then, x will always be in the format 1.XXXXXXXXX...
- Therefore, in the fraction portion, we can encode just what is *to the right* of the decimal point! This means we get one more digit for precision.

Value encoded = 1. [FRACTION BINARY DIGITS]_





Sign			Ехро	onent				Frac	ction	
0	0	•••	0	0	0	1	0	1	0	•••

Is this number:

- A) Greater than 0?
- B) Less than 0?





Sign			Ехро		Frac	ction				
0	0	•••	0	0	0	1	0	1	0	•••

Is this number:

- A) Greater than 0?
- B) Less than 0?

Is this number:

- A) Less than -1?
- B) Between -1 and 1?
- C) Greater than 1?

1.25 x 2^-126

Skipping Numbers

• We said that it's not possible to represent *all* real numbers using a fixedwidth representation. What does this look like?

Float Converter

<u>https://www.h-schmidt.net/FloatConverter/IEEE754.html</u>

Floats and Graphics

<u>https://www.shadertoy.com/view/4tVyDK</u>

Let's Get Real

What would be nice to have in a real number representation?

- Represent widest range of numbers possible \checkmark
- Flexible "floating" decimal point \checkmark
- Represent scientific notation numbers, e.g. 1.2 x 10⁶ **?**
- Still be able to compare quickly \checkmark
- Have more predictable over/under-flow behavior ?

Representing Zero

The float representation of zero is all zeros (with any value for the sign bit)

Sign	Exponent	Fraction
any	All zeros	All zeros

• This means there are two representations for zero! 🛞

Representing Small Numbers

If the exponent is all zeros, we switch into "<u>denormalized</u>" mode.

Sign	Exponent	Fraction
any	All zeros	Any

- We now treat the exponent as -126, and the fraction as <u>without the</u> <u>leading 1</u>.
- This allows us to represent the smallest numbers as precisely as possible.

Representing Exceptional Values

If the exponent is all ones, and the fraction is all zeros, we have +- infinity.

Sign	Exponent	Fraction
any	All ones	All zeros

- The sign bit indicates whether it is positive or negative infinity.
- Floats have built-in handling of over/underflow!
 - Infinity + anything = infinity
 - Negative infinity + negative anything = negative infinity
 - Etc.

Representing Exceptional Values

If the exponent is all ones, and the fraction is nonzero, we have **Not a Number (NaN)**

Sign			Ехро	Fraction			
any	1	•••	•••	•••	•••	1	Any nonzero

- NaN results from computations that produce an invalid mathematical result.
 - Sqrt(negative)
 - Infinity / infinity
 - Infinity + -infinity
 - Etc.

Number Ranges

- 32-bit integer (type **int**):
 - > -2,147,483,648 to 2147483647

> Every integer in that range can be represented

- 64-bit integer (type long):
 > -9,223,372,036,854,775,808 to 9,223,372,036,854,775,807
- 32-bit floating point (type **float**):
 - ~1.2 x10⁻³⁸ to ~3.4 x10³⁸
 - Not all numbers in the range can be represented (not even all integers in the range can be represented!)
 - Gaps can get quite large! (larger the exponent, larger the gap between successive fraction values)
- 64-bit floating point (type **double**):
 - ~2.2 x10⁻³⁰⁸ to ~1.8 x10³⁰⁸

Precision options

• Single precision: 32 bits

S	ехр	frac
1	8-bits	23-bits

• Double precision: 64 bits

S	exp	frac
1	11-bits	52-bits

• Extended precision: 80 bits (Intel only)

S	ехр	frac
1	15-bits	63 or 64-bits

Visualization: Floating Point Encodings



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Tiny Floating Point Example

S	exp	frac		
1	4-bits	3-bits		

- 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 (= $2^{(4-1)}-1$)
 - the last three bits are the **frac**
- Same general form as IEEE Format
 - normalized, denormalized
 - representation of 0, NaN, infinity

Dynamic Range (Positive Only)

	S	exp	frac	Е	Value
	0	0000	000	-6	0
	0	0000	001	-6	1/8*1/64 = 1/512 closest to zero
Denormalized	0	0000	010	-6	$2/8 \times 1/64 = 2/512$
numbers	••••				
	0	0000	110	-6	6/8*1/64 = 6/512
	0	0000	111	-6	7/8*1/64 = 7/512 largest 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	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 norm
	0	1111	000	n/a	inf

 $v = (-1)^{s} M 2^{E}$ n: E = Exp - Bias d: E = 1 - BiasBias = 2⁽⁴⁻¹⁾-1 = 7

Distribution of Values

- 6-bit IEEE-like format
 - e = 3 exponent bits
 - f = 2 fraction bits
 - Bias is $2^{3-1}-1 = 3$

S	exp	frac		
1	3-bits	2-bits		

• Notice how the distribution gets denser toward zero.



Distribution of Values (close-up view)

- 6-bit IEEE-like format
 - e = 3 exponent bits
 - f = 2 fraction bits
 - Bias is 3

S	exp	frac		
1	3-bits	2-bits		



Special Properties of the IEEE 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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Demo: Float Arithmetic



float_arithmetic.c

Is this just overflowing? It turns out it's more subtle.

float a = 3.14; float b = 1e20; printf("(3.14 + 1e20) - 1e20 = %g\n", (a + b) - b); // prints 0 printf("3.14 + (1e20 - 1e20) = %g\n", a + (b - b)); // prints 3.14 Let's look at the binary representations for 3.14 and 1e20:





To add real numbers, we must align their binary points:

3.14

What does this number look like in 32-bit IEEE format?

Step 1: convert from base 10 to binary

Step 2: find most significant 1 and take the next 23 digits for the fractional component, rounding if needed.

1 01011010111100011101100

Step 3: find how many places we need to shift **left** to put the number in 1.xxx format. This fills in the exponent component.

66 shifts -> 66 + 127 = 193

Step 4: if the sign is positive, the sign bit is 0. Otherwise, it's 1.

Sign bit is 0.

The binary representation for 1e20 + 3.14 thus equals the following:

31	30		23	22 0
0		11000001		01011010111100011101100

This is the **same** as the binary representation for 1e20 that we had before!

We didn't have enough bits to differentiate between 1e20 and 1e20 + 3.14.

Is this just overflowing? It turns out it's more subtle.

```
float a = 3.14;
float b = 1e20;
printf("(3.14 + 1e20) - 1e20 = %g\n", (a + b) - b); // prints 0
printf("3.14 + (1e20 - 1e20) = %g\n", a + (b - b)); // prints 3.14
```

Floating point arithmetic is not associative. The order of operations matters!

- The first line loses precision when first adding 3.14 and 1e20, as we have seen.
- The second line first evaluates 1e20 1e20 = 0, and then adds 3.14

Demo: Float Equality



float_equality.c

Float arithmetic is an issue with most languages, not just C!

<u>http://geocar.sdf1.org/numbers.html</u>

Let's Get Real

What would be nice to have in a real number representation?

- Represent widest range of numbers possible \checkmark
- Flexible "floating" decimal point \checkmark
- Represent scientific notation numbers, e.g. 1.2 x $10^6 \checkmark$
- Still be able to compare quickly \checkmark
- Have more predictable over/under-flow behavior \checkmark

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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 ightarrow int
 - Truncates fractional part
 - Like rounding toward zero
 - Not defined when out of range or NaN: Generally sets to TMin
 - int ightarrow double
 - Exact conversion, as long as int has \leq 53 bit word size
 - int \rightarrow float
 - Will round according to rounding mode

Ariane 5: A Bug and A Crash

- On June 4, 1996, Ariane 5 rocket self destructed just after 37 seconds after liftoff
- Cost: \$500 million
- **Cause:** An overflow in the conversion from a 64 bit floating point number to a 16 bit signed integer
- A design flaw:
 - 5 times faster than Ariane 4
 - Reused same software specifications from Ariane 4
 - Ariane 4 assumes horizontal velocity would never overflow a 16-bit number



[©] Fourmy/REA/SABA/Corbis

0x5F3759DF or The Fast Inverse Square Root

```
float Q_rsqrt( float number )
```

```
return y;
```

The fast inverse square root implementation from <u>Quake III Arena</u>, including the exact original comment text

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	False				
• x == (int	• x == (int)(double) x				
• f == (flo	at)(da	ouble) f	True		
• d == (flo	oat) d		False		
• f == -(-f	• f == -(-f);				
• 2/3 == 2/	3.0		False		
• d < 0.0	⇒	((d*2) < 0.0)	True (OF?)		
• d > f	\Rightarrow	-f > -d	True		
• d * d >=	0.0		True (OF?)		
• (d+f)-d =	False				

Floats Summary

- IEEE Floating Point is a carefully-thought-out standard. It's complicated, but engineered for their goals.
- Floats have an extremely wide range, but cannot represent every number in that range.
- Some approximation and rounding may occur! This means you definitely don't want to use floats e.g. for currency.
- Associativity does not hold for numbers far apart in the range
- Equality comparison operations are often unwise.

Additional Video

Thanks Turan Berke Yakarer for the suggestion ☺



https://www.youtube.com/watch?v=dQhj5RGtag0

Additional Reading

What Every Computer Scientist Should Know About Floating-Point Arithmetic

DAVID GOLDBERG

Xerox Palo Alto Research Center, 3333 Coyote Hill Road, Palo Alto, California 94304

Floating-point arithmetic is considered an esotoric subject by many people. This is rather surprising, because floating-point is ubiquitous in computer systems: Almost every language has a floating-point datatype; computers from PCs to supercomputers have floating-point accelerators; most compilers will be called upon to compile floating-point algorithms from time to time; and virtually every operating system must respond to floating-point exceptions such as overflow This paper presents a tutorial on the aspects of floating-point that have a direct impact on designers of computer systems. It begins with background on floating-point representation and rounding error, continues with a discussion of the IEEE floating-point standard, and concludes with examples of how computer system builders can better support floating point.

Categories and Subject Descriptors: (Primary) C.0 [Computer Systems Organization]: General—instruction set design; D.3.4 [Programming Languages]: Processors—compilers, optimization; G.1.0 [Numerical Analysis]: General—computer arithmetic, error analysis, numerical algorithms (Secondary) D.2.1 [Software Engineering]: Requirements/Specifications—languages; D.3.1 [Programming Languages]: Formal Definitions and Theory—semantics D.4.1 [Operating Systems]:

Process Management-synchronization

General Terms: Algorithms, Design, Languages

Additional Key Words and Phrases: denormalized number, exception, floating-point, floating-point standard, gradual underflow, guard digit, NaN, overflow, relative error, rounding error, rounding mode, ulp, underflow

What Every Computer Scientist Should Know About Floating-Point Arithmetic, David Goldberg, ACM Computing Surveys, 23(1), 1991

Recap

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Next time: How can a computer represent and manipulate more complex data like text?