Computer Arithmetic

Troels Henriksen

Inspired by slides by Randal E. Bryant and David R. O'Hallaron. Some material by Michael Kirkedal Tomsen.

Agenda

Floating point arithmetic

Biased numbers

Background: Fractional binary numbers

IEEE floating point standard

Examples and properties

Rounding, addition, and multiplication

Floating point in C

Summary

Integers

Arithmetic

Conversion, casting

Expansion and truncation

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Biased numbers

Background: Fractional binary numbers IEEE floating point standard Examples and properties Rounding, addition, and multiplication Floating point in C

Summary

Integers

Arithmetic Conversion, casting Expansion and truncation

For *biased numbers*, the raw bits are interpreted as unsigned, and then a constant *bias* is subtracted.

Unsigned

Two's complement

Biased

$$B2U(X) = \sum_{i=0}^{w-1} x_i \cdot 2^i \quad B2S(X) = -x_{w-1} \cdot 2^{w-1} + \sum_{i=0}^{w-2} x_i \cdot 2^i \quad B2I(X) = \left(\sum_{i=0}^{w-1} x_i \cdot 2^i\right) - \text{Bias}$$

Typically

$$Bias = 2^{w-1} - 1$$

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$$i=0$$

i=

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	B2U	B2S	B2I
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01111111 ₂			

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	B2U	B2S	B21
000000002	0 ₁₀	0 ₁₀	-127_{10}
011111112	127 ₁₀		

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- 0	, Dia3 — 127			
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,,

Typically

$$Bias = 2^{w-1} - 1$$

Examples for $w = 0$,	Dia5 — 127			
		B2U	B2S	B2I
-	000000002	0 ₁₀	0 ₁₀	-127_{10}
	011111112	127 ₁₀	127 ₁₀	0 ₁₀
	11111111 ₂	255 ₁₀		

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Unsigned	Two's complement	Biased

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• Typically
$$Rias = 2^{w-1} - 1$$

Evamples for w = 9 Pies = 127

Floating point arithmetic

Biased numbers

Background: Fractional binary numbers

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Summary

Integers

Arithmetic Conversion, casting Expansion and truncation

Integral binary numbers

We have seen that

100101012

is basically interpreted like

 149_{10}

"Structure" is the same, just with 2 instead of 10.

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Can we do the same thing for fractional numbers?

1011.1012

Fractional numbers

$$123.456 = 1 \cdot 10^2 + 2 \cdot 10^1 + 3 \cdot 10^0 + 4 \cdot 10^{-1} + 5 \cdot 10^{-2} + 6 \cdot 10^{-3}$$

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Generally

$$a_{m-1}\cdots a_0.a_{-1}\cdots a_{-n}=\sum_{i=-n}^{m-1}a_i\cdot 10^i$$

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$$a_{m-1}\cdots a_0.a_{-1}\cdots a_{-n}=\sum_{i=-n}^{m-1}a_i\cdot 10^i$$

Even more generally, for radix r

$$a_{m-1}\cdots a_0.a_{-1}\cdots a_{-n} = \sum_{i=-n}^{m-1} a_i \cdot r^i$$

Fractional binary numbers

Representation

- Bits to the right of "binary point" represents fractional powers of 2.
- Represents rational number.

$$b_{m-1}\cdots b_0.b_{-1}\cdots b_{-n} = \sum_{i=-n}^{m-1} b_i \cdot 2^i$$

Examples of fractional binary numbers

Value $5\frac{3}{4}$	Representation 101.11 ₂
$2\frac{7}{8}$	10.111 ₂
$1\frac{7}{16}$	1.0111 ₂

Observations

- Divide by 2 by logical shifting right.
- Multiply by 2 by shifting left.
- Numbers of form 0.111... are just below 1.0.
 - $ightharpoonup 1/2 + 1/4 + 1/8 + \cdots 1/2^n + \cdots \sim 1.0.$
 - ▶ Use notation 1.0ϵ .

Representable numbers

Limitation #1

- Can only represent fractional part of form $x/2^k$
- Other rational numbers have repeating bit representation

Value $\frac{1}{3}$	$\begin{array}{c} \textbf{Representation} \\ 0.0101010101[01] \cdots_2 \end{array}$
<u>1</u> 5	$0.001100110011[0011] \cdots_{2}$
$\frac{1}{10}$	0.0001100110011[0011]2

Limitation #2

- Just one setting of binary point within the w bits.
 - Limited range of numbers—very small values? Very large?

The fixed-point dilemma

Consider
$$w = 8$$

1 bit for fraction

- Largest number: 1111111.1₂ = 127.5₁₀
- Increment: $0000000.1_2 = 0.5_{10}$

7 bits for fraction

- Largest number: 1.11111111₂ = 1.9921875₁₀
- Increment: $0.0000001_2 = 0.0078125_{10}$

4 bits for fraction

- Largest number: 1111.1111₂ = 15.9375₁₀
- Increment: $0000.0001_2 = 0.0625_{10}$

Fixed-point has same absolute precision everywhere, but this means relative precision is worse for numbers close to 0!

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IEEE Floating Point

IEEE Standard 754

- Established in 1985 as uniform standard for floating point.
 - Many idiosyncratic formats before then.
- Supported by all major CPUs, GPUs, and most other processors.

Driven by numerical concerns

- Nice standards for rounding, overflow, underflow.
- Hard to make fast in hardware.
 - Numerical analysts predominated over hardware designers in defining standard.
 - ... but (later) Turing Award winner William Kahan secretly knew that Intel had figured out how to make accurate computation fast.
 - Beware the wrath of Kahan!
 - http://people.eecs.berkeley.edu/~wkahan/

Essentially scientific notation

$$3.5 \times 10^2 = 2$$

- **Significand** is 3.5
 - ► Conventionally a number in range [1, 10), with sign.
- **Exponent** is 2.
 - Can also be negative.

Essentially scientific notation

$$3.5 \times 10^2 = 2$$

- Significand is 3.5
 - Conventionally a number in range [1, 10), with sign.
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To keep significand in range, adjust exponent:

$$\begin{array}{ccc} 35\times10^1 & \Rightarrow & 3.5\times10^2 \\ 0.35\times10^1 & \Rightarrow & 3.5\times10^2 \end{array}$$

IEEE 754 uses bits instead of digits, and specifies a fixed-size encoding, but idea is the same.

Floating Point Representation

Numerical form

$$(-1)^S \cdot M \cdot 2^E$$

- **Sign bit** *S* determines whether number negative or positive.
- **Significand** *M* normally a fractional value in range [1, 2).
- **Exponent** *E* weights value by power of two.

Encoding

- Most significant bit is sign bit.
- Exp field encodes *E* (but is not equal to *E*).
- Frac field encodes *M* (but is not equal to *M*).

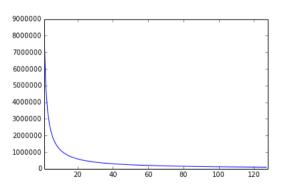
S Exp Frac

Why such a weird format?

The point is floating

- No fixed number of bits allocated to "fraction".
- More bits close to 0, fewer bits for numbers with large magnitude.
- Symmetric around 0.

Density of floats



https://stackoverflow.com/a/24179424/6131552

Precision options

32-bit single precision: float Exp Frac 1 bit 8 bits 23 bits 64-bit double precision: double Exp Frac 1 bit 11 bits 52 bits 80-bit Extended precision (Intel only, never use): long double Exp Frac 1 bit 15 bits 64 bits

Newer standards contain more variants (16 bits, decimal floats) that we will not cover.

Main problem: not all numbers are representable

Format trouble

For example the number

$$0.1_{10}$$

cannot be represented on the form

$$(-1)^S \cdot M \cdot 2^E$$

Precision trouble

- A fixed number of bits cannot represent all numbers.
- Integer types represent an interval of natural numbers.
- Any nontrivial interval of rational numbers contains infinity elements.
- "Neighbouring" floats are separated by a "step size" 2^E (we'll see).

Consequence

Rounding.

Rounding of floating point numbers

• Floating point arithmetic returns the floating point number *closest* to the mathematically correct result.

Example

Mathematically,

$$1/10 = 0.1$$

But since 0.1 cannot be represented in binary floating point, we instead get the number that is closest:

$$1/10 = 0.10000000000000000555111512312578270211815834045410156250$$

(With 64-bit floats).

We will return to this, but writing algorithms that are robust to roundoff errors is a *big topic* that is outside the scope of this course.

Normalised values when $\mathsf{Exp} \neq 0 \cdots 0$ and $\mathsf{Exp} \neq 1 \cdots 1$

$$v = (-1)^{S} \cdot M \cdot 2^{E}$$

S Exp

Frac

Exponent encoded as biased value

$$E = Exp - Bias$$

- Exp: unsigned value of Exp field.
- ▶ Bias = $2^{k-1} 1$, where k is number of Exp bits.
 - ► Single precision: 127 (Exp \in [1, 254], $E \in$ [-126, 127]).
 - ▶ Double precision: 1023 (Exp \in [1, 2046], $E \in$ [-1022, 1023]).

Significand coded with implied leading 1:

$$M = 1.xxx \cdots x_2$$

- xxx · · · x: bits of Frac field.
- Minimum when Frac = $0000 \cdots 0$ (M = 1).
- Maximum when Frac = $1111 \cdots 1$ ($M = 2 \epsilon$).
- Get extra leading bit for free.

Normalised encoding example

$$v = (-1)^S \cdot M \cdot 2^E$$
 $E = \text{Exp} - \text{Bias}$
Value: float $F = 15213.0$
 $15213_{10} = 11101101101101_2$

Significand

$$M = 1.1101101101101_2$$

Frac = 11011011011010000000000₂

 $= 1.1101101101101_2 \cdot 2^{13}$

Exponent

$$E = 13_{10}$$

Bias = 127_{10}
Exp = $E +$ Bias = $140_{10} = 10001100_2$

 Result
 0
 10001100
 11011011011010000000000

Denormal values

$$v = (-1)^S \cdot M \cdot 2^E$$
 $E = 1 - \text{Bias}$

Occur when $Exp = 000 \cdots 0_2$.

Exponent encoded as

$$E = 1 - \mathsf{Bias}$$

Significand coded with implied leading 0:

$$M = 0.xxx \cdots x_2$$

- Cases
 - ► $Exp = 000 \cdots 0_2$, $Frac = 000 \cdots 0_2$
 - Represents zero value.
 - Note distinct values -0, +0 when might that be useful?
 - \triangleright Exp = $000 \cdots 0_2$, Frac $\neq 000 \cdots 0_2$
 - Numbers closest to 0.0.
 - Called subnormal numbers.
 - ► Ensure that $x \neq y \Rightarrow x y \neq 0$, i.e. avoid underflow.

Special values

Occur when
$$Exp = 111 \cdots 1_2$$
.

When
$$Exp = 111 \cdots 1_2$$
, $Frac = 000 \cdots 0_2$

- Represents $\pm \infty$.
- Typically the result of overflow.
 - Overflow can be negative!
 - Underflow is when the result becomes zero due to rounding.
- Both positive and negative.
- Examples:

$$\frac{1}{0} = \frac{-1}{-0} = \infty \qquad \frac{1}{-0} = -\infty$$

When
$$Exp = 111 \cdots 1_2$$
, $Frac \neq 000 \cdots 0_2$

- Not A Number (NaN).
- Represents case when no numeric value can be determined.
- Examples:

$$\operatorname{sqrt}(-1)$$
 $\infty - \infty$ $\infty \cdot 0$

The floating point number line

NaN

Note that NaNs are unordered:

- NaN is different from everything even other NaNs!
 - ► NaN == NaN is false.
 - Floating-point equality is not reflexive!
- NaN > x and NaN < x is false for all x.

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Play the game

https://topps.diku.dk/compsys/floating-point.html

Tiny 8-bit floating point example

S	Exp	Frac
1b	4b	3b

8-bit floating point representation

- Sign bit is the most significant bit (leftmost).
- The next four bits are Exp with a bias of 7.
- The last three bits are Frac.

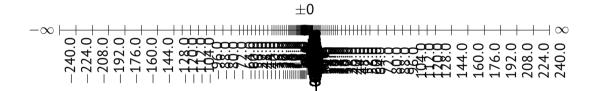
Same general form as IEEE Format

- Normalised, denormalised.
- Representation of 0, NaN, both infinities.

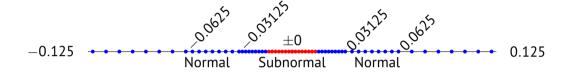
Let's look at their dynamic range.

	Sign	Exp	Frac	E	Value	
Denormalised	0	0000	000	-6	$0/8 \cdot 1/64 = 0/512$	zero
	0	0000	001	-6	$1/8 \cdot 1/64 = 1/512$	closest to zero
	0	0000	010	-6	$2/8 \cdot 1/64 = 2/512$	
	0	0000	111	-6	$7/8 \cdot 1/64 = 7/512$	largest denorm
Normalised	0	0001	000	-6	$8/8 \cdot 1/64 = 8/512$	smallest norm
	0	0001	001	-6	$9/8 \cdot 1/64 = 9/512$	
	0	0110	110	-1	$14/8 \cdot 1/2 = 14/16$	
	0	0110	111	-1	$15/8 \cdot 1/2 = 15/16$	closest to 1
	0	0111	000	0	$8/8 \cdot 1 = 8/8$	1
	0	0111	001	0	$9/8 \cdot 1 = 9/8$	closest to 1
	0	0111	010	0	$10/8 \cdot 1 = 10/8$	
	0	1110	110	7	$14/8 \cdot 128 = 224$	
	0	1110	111	7	$15/8 \cdot 128 = 240$	
	0	1111	000	N/A	∞	
	0	1111	001	N/A	NaN	
					NaN	

Distribution of values



Distribution of values (zooming in)



- Note how the distribution gets denser towards zero.
- Note the big gap there would be around 0 if we did not have subnormals.
- Each of the spans with same distance between neighbors corresponds to numbers with same Exp.

S	Exp	Frac
1b	4b	3b

Useful properties of the IEEE encoding

S Exp Frac

- Floating-point zero same as integer zero
 - ► All bits 0.
 - ...but negative zero is different.
- Can almost compare floats with unsigned integer comparisons
 - Must first compare sign bit.
 - ► Must consider -0 = 0.
 - NaNs problematic:
 - Greater than any other value (because $Exp = 111 \cdots 1_2$).
 - What should comparison yield?
 - Otherwise OK:
 - Normalised and denormalised compare as expected.
 - Infinities ordered properly relative to finities.

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Basic idea behind floating point operations

$$x +_f y = \text{Round}(x + y)$$

 $x \times_f y = \text{Round}(x \times y)$

Basic idea

- ► First compute exact result!
- ► Then round it to fit into desired precision.
 - Overflow if exponent too large.
 - Round to fit into Frac.

	1.40	1.60	1.50	2.50	-1.50
Towards zero	1	1	1	2	-1
Towards $-\infty$					

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Nearest even	'				

• There's more than one way to round a number, here to an integer.

	1.40	1.60	1.50	2.50	-1.50
Towards zero	1	1	1	2	-1
Towards $-\infty$	1	1	1	2	-2
Towards ∞	2	2	2	3	-1
Nearest even ∞	1	2	2	2	-2

• "Round to nearest, ties to even" is the default rounding mode.

Default rounding mode

- But can be changed dynamically.
 - https:

```
//www.gnu.org/software/libc/manual/html_node/Rounding.html
```

- Never do this.
- All others are statistically biased.
 - Biased: Sum of set of positive numbers will consistently be over- or under-estimated.

- When exactly halfway between two possible values:
 - Round so that least significant digit is even.
- E.g. rounding to nearest hundredth:
 - **7.8949999**:

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 - **7.8990001**:

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- E.g. rounding to nearest hundredth:
 - **7.8949999: 7.89**
 - **7.8990001: 7.90**
 - **7.8950000: 7.90**
 - **7.8850000**:

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 - Round so that least significant digit is even.
- E.g. rounding to nearest hundredth:
 - **7.8949999: 7.89**
 - **7.8990001: 7.90**
 - **7.8950000: 7.90**
 - **7.8850000: 7.88**

- Binary fractional numbers
 - "Even" when least significant bit is 0.
 - \blacktriangleright "Half way" when bits to right of rounding position are $100\cdots_2$.
- Examples
 - Round to nearest 1/4 (2 bits right of binary point).

Value	Binary	Rounded	Action	Rounded value

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Value	Binary	Rounded	Action	Rounded value
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Examples

Value	Binary	Rounded	Action	Rounded value
2 3/32	10.000112	10.002	(< 1/2-down)	2
2 3/16				

Binary fractional numbers

- "Even" when least significant bit is 0.
- \blacktriangleright "Half way" when bits to right of rounding position are $100\cdots_2$.

Examples

Value	Binary	Rounded	Action	Rounded value
2 3/32	10.000112	10.002	(< 1/2-down)	2
2 3/16	10.00 <mark>110</mark> 2			

Binary fractional numbers

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Examples

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2 3/32	10.000112	10.002	(< 1/2 - down)	2
2 3/16	10.00 <mark>110</mark> 2	10.012	(>1/2-up)	2 1/4

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2 3/16	10.00 <mark>110</mark> 2	10.012	(>1/2-up)	2 1/4
2 7/8				

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Examples

Value	Binary	Rounded	Action	Rounded value
2 3/32	10.000112	10.002	(< 1/2 - down)	2
2 3/16	10.00 <mark>110</mark> 2	10.012	(>1/2-up)	2 1/4
2 7/8	10.11 <mark>100</mark> 2			

Binary fractional numbers

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- \blacktriangleright "Half way" when bits to right of rounding position are $100\cdots_2$.

Examples

Value	Binary	Rounded	Action	Rounded value
2 3/32	10.00 <mark>011</mark> 2	10.00 ₂	(< 1/2 - down)	2
2 3/16	10.00 <mark>110</mark> 2	10.012	(>1/2-up)	2 1/4
2 7/8	10.11 <mark>100</mark> 2	11.00 ₂	(1/2-up)	3

Binary fractional numbers

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Examples

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2 3/32	10.000112	10.002	(< 1/2 - down)	2
2 3/16	10.00 <mark>110</mark> 2	10.012	(>1/2-up)	2 1/4
2 7/8	10.11 <mark>100</mark> 2	11.00 ₂	(1/2-up)	3
2 5/8			. ,	

Binary fractional numbers

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2 3/32	10.000112	10.002	(< 1/2-down)	2
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2 7/8	10.11 <mark>100</mark> 2	11.00 ₂	(1/2-up)	3
2 5/8	10.10 <mark>100</mark> ₂		•	

Binary fractional numbers

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2 3/32	10.000112	10.002	(< 1/2-down)	2
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2 7/8	10.11 <mark>100</mark> 2	11.00 ₂	(1/2-up)	3
2 5/8	10.10 <mark>100</mark> 2	10.102	(1/2-down)	2 1/2

Floating point multiplication (assuming operands are numbers)

$$((-1)^{S_3} \cdot M_3 \cdot 2^{E_3}) = ((-1)^{S_1} \cdot M_1 \cdot 2^{E_1}) \cdot ((-1)^{S_2} \cdot M_2 \cdot 2^{E_2})$$

Exact result

$$S_3 = S_1 \oplus S_2$$

$$M_3 = M_1 \cdot M_2$$

$$E_3 = E_1 + E_2$$

where \oplus is exclusive-or.

Fixing

- ▶ If $M_3 \ge 2$, shift M_3 right and increment E_e .
- ▶ If E_3 out of range, overflow to ∞ .
- Round M_3 to fit Frac precision.

Implementation

Biggest chore is multiplying significands.

Floating point addition (assuming operands are numbers)

$$((-1)^{S_3}\cdot M_3\cdot 2^{E_3})=((-1)^{S_1}\cdot M_1\cdot 2^{E_1})+((-1)^{S_2}\cdot M_2\cdot 2^{E_2})$$

Approach

- Assume without loss of generality that $E_1 \geq E_2$.
- ightharpoonup Rewrite smaller number such that its exponent matches E_1 :

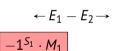
$$((-1)^{S_3} \cdot M_3 \cdot 2^{E_3}) = ((-1)^{S_1} \cdot M_1 \cdot 2^{E_1}) + ((-1)^{S_2} \cdot M_2' \cdot 2^{E_1})$$

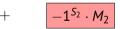
Exact result

- Sign S_3 , significant M_3 :
 - Result of signed addition.

Fixing

- If $M_3 > 2$, shift M_3 right and increment E_3 .
- ▶ If M_3 < 1, shift M left k positions and decrement E_3 by k.
- ▶ If E_3 out of range, overflow to ∞ .
- Round *M* to fit Frac precision.







Example of floating-point addition with a 2-bit significand

$$(-1.01 \cdot 2^2) + (1.1 \cdot 2^4)$$

= $(-1.01 \cdot 2^2) + (110.0 \cdot 2^2)$ Align exponents
= $(-1.01 + 110.0) \cdot 2^2$ Distributivity
= $100.11 \cdot 2^2$ Add significands
= $1.0011 \cdot 2^4$ Normalise
= $1.01 \cdot 2^4$ Perform rounding

- Compared to those of Abelian Group
 - Closed under addition?

- Compared to those of Abelian Group
 - Closed under addition? Yes
 - But may generate infinity or NaN.
 - Commutative?

- Compared to those of Abelian Group
 - Closed under addition? Yes
 - But may generate infinity or NaN.
 - Commutative? Yes
 - Associative?

Compared to those of Abelian Group

- Closed under addition? Yes
 - But may generate infinity or NaN.
- Commutative? Yes
- Associative? No
 - Due to overflow and inexactness of rounding.
 - (3.14 + 1e10) 1e10 = 0
 - \triangleright 3.14 + (1e10-1e10) = 3.14
- 0 is additive identity?

Compared to those of Abelian Group

- Closed under addition? Yes
 - But may generate infinity or NaN.
- Commutative? Yes
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 - Due to overflow and inexactness of rounding.
 - (3.14 + 1e10) 1e10 = 0
 - \triangleright 3.14 + (1e10-1e10) = 3.14
- 0 is additive identity? Yes
- Does every element have an additive inverse?

Compared to those of Abelian Group

- Closed under addition? Yes
 - But may generate infinity or NaN.
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 - Due to overflow and inexactness of rounding.
 - (3.14 + 1e10) 1e10 = 0
 - \triangleright 3.14 + (1e10-1e10) = 3.14
- 0 is additive identity? Yes
- Does every element have an additive inverse? Almost
 - Infinities and NaN do not have inverses.

Monotonicity

$$\triangleright$$
 $a \ge b \Rightarrow a + c \ge b + c?$

Compared to those of Abelian Group

- Closed under addition? Yes
 - But may generate infinity or NaN.
- Commutative? Yes
- Associative? No
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 - (3.14 + 1e10) 1e10 = 0
 - \triangleright 3.14 + (1e10-1e10) = 3.14
- 0 is additive identity? Yes
- Does every element have an additive inverse? Almost
 - Infinities and NaN do not have inverses.

Monotonicity

- ▶ $a \ge b \Rightarrow a + c \ge b + c$? Almost
 - Infinities and NaNs are the exception.

- Compared to those of a commutative ring
 - Closed under multiplication?

- Compared to those of a commutative ring
 - ► Closed under multiplication? **Yes**
 - But may generate infinity or NaN.
 - Commutative?

- Compared to those of a commutative ring
 - Closed under multiplication? Yes
 - But may generate infinity or NaN.
 - Commutative? Yes
 - Associative?

Compared to those of a commutative ring

- Closed under multiplication? Yes
 - But may generate infinity or NaN.
- Commutative? Yes
- ► Associative? **No**
 - Due to overflow and inexactness of rounding.
 - $(1e20*1e20)*1e-20=\infty$
 - ► 1e20*(1e20*1e-20) = 1e20
- 1 is multiplicative identity?

Compared to those of a commutative ring

- Closed under multiplication? Yes
 - But may generate infinity or NaN.
- Commutative? Yes
- Associative? No
 - Due to overflow and inexactness of rounding.
 - $(1e20 * 1e20) * 1e-20=\infty$
 - ► 1e20*(1e20*1e-20) = 1e20
- 1 is multiplicative identity? Yes
- Multiplication distributes over addition?

Compared to those of a commutative ring

- Closed under multiplication? Yes
 - But may generate infinity or NaN.
- Commutative? Yes
- Associative? No
 - Due to overflow and inexactness of rounding.
 - $(1e20*1e20)*1e-20=\infty$
 - ► 1e20*(1e20*1e-20) = 1e20
- 1 is multiplicative identity? Yes
- Multiplication distributes over addition? No
 - Overflow and rounding again.
 - ightharpoonup 1e20* (1e20-1e20) = 0.0
 - ► 1e20*1e20 1e20*1e20 = NaN

Floating point arithmetic

Biased numbers

Background: Fractional binary numbers

IEEE floating point standard

Examples and properties

Rounding, addition, and multiplication

Floating point in C

Summary

Integers

Arithmetic

Conversion, casting

Expansion and truncation

Floating point in C

C guarantees two types

- float: 32-bit single precision.
- double: 64-bit single precision.

Conversions/casting

- ▶ Casting between int, float, and double changes bit represensation.
- double/float to int
 - Truncates fractional part.
 - Like rounding toward zero.
 - Not defined when out of range or NaN: generally sets to SMin.
- int to double
 - Exact conversion as long as int fits in 53 bits.
- ▶ int to float
 - Will round according to rounding mode.

Floating point is exciting!



First "flight" of the Ariane 5 in 1996.

Floating point is exciting!



First "flight" of the Ariane 5 in 1996.

- A double storing horizontal velocity of the rocket was converted to a 16-bit signed integer.
- The number was larger than 32767 so the conversion failed, causing an exception, crashing the guidance module.

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

Assume neither d nor t is NaN.

Assume int is 32 bits.

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

```
 = x == (int) (float) x
```

Assume neither d nor t is NaN.

Assume int is 32 bits.

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

```
 = x == (int) (float) x
```

$$x == (int) (double) x$$

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

Assume neither ${\tt d}$ nor ${\tt t}$ is NaN.

Assume int is 32 bits.

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

```
x == (int) (float) x
x == (int) (double) x
f == (float) (double) f
```

```
int x = \dots;
float f = \dots;
double d = \dots;
```

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

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

```
" x == (int) (float) x
" x == (int) (double) x
" f == (float) (double) f
" d == (double) (float) d
```

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

```
x == (int) (float) x
x == (int) (double) x
f == (float) (double) f
d == (double) (float) d
f == -(-f)
```

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

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

```
" x == (int) (float) x
" x == (int) (double) x
" f == (float) (double) f
" d == (double) (float) d
" f == -(-f)
" 2/3 == 2/3.0
```

For each of the following C expressions, either

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- Explain why it's not.

```
" x == (int) (float) x
" x == (int) (double) x
" f == (float) (double) f
" d == (double) (float) d
" f == -(-f)
" 2/3 == 2/3.0
" d < 0.0 ⇒ (d*2) < 0.0</pre>
```

For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

Assume neither d nor t is NaN. Assume int is 32 bits.

 $\blacksquare d > f \Rightarrow -f > -d$

For each of the following C expressions, either

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For each of the following C expressions, either

- Argue that it is true for all argument values.
- Explain why it's not.

Assume neither d nor t is NaN. Assume int is 32 bits.

• d * d >= 0.0• (d+f)-d == f

Floating point arithmetic

Biased numbers

Background: Fractional binary numbers

IEEE floating point standard

Examples and properties

Rounding, addition, and multiplication

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Summary

- IEEE floating point has clear properties.
 - But they may not match your intuition.
- Represents numbers of the form $M \cdot 2^E$.
- One can reason about operations independent of implementation.
 - Computed with perfect precision and then rounded.
 - But rounded after every "primitive" operation (e.g. addition, multiplication).
- Not the same as \mathbb{Q}/\mathbb{R} arithmetic.
 - Violates associativity and distributivity, mostly due to rounding.
 - Sometimes makes life difficult for heavy-duty numerical programming.
 - ▶ But carefully designed such that "naive" use mostly does what one expects.

```
Also try this tool: https://evanw.github.io/float-toy/
And read this: https://moyix.blogspot.com/2022/09/
someones-been-messing-with-my-subnormals.html
```

Floating point arithmetic

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Example: Decimal addition

Example: Binary addition

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Summary

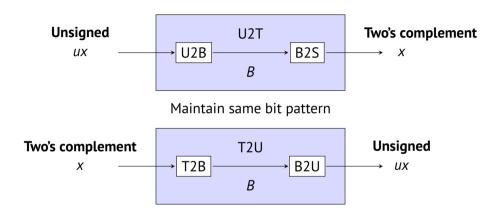
Integers

Arithmetic

Conversion, casting

Expansion and truncation

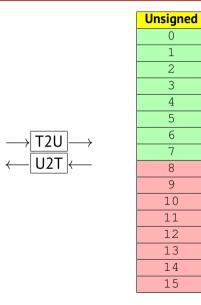
Mapping between signed and unsigned



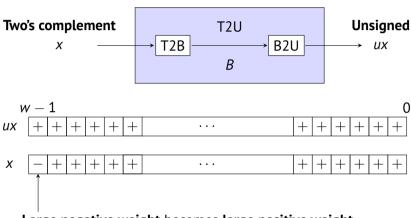
Mapping between unsigned and two's complement numbers: **Keep bit representations and reinterpret.**

Mapping signed ⇔ unsigned

Bits	Signed
0000	0
0001	1
0010	2
0011	3
0100	4
0101	5
0110	6
0111	7
1000	-8
1001	-7
1010	-6
1011	-5
1100	-4
1101	-3
1110	-2
1111	-1



Relation between signed and unsigned

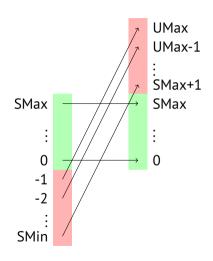


Large negative weight becomes large positive weight.

Conversion (that is, reinterpretation) visualized

Two's complement to unsigned

- Ordering inversion.
- Negative numbers become large positive numbers.



Signed versus unsigned in C

C makes working with this more error-prone than it should be.

Types

- Signedness part of type: unsigned int, int32_t, uint32_t.
- **Constants** By default are considered signed integers.
 - Unsigned with U suffix: 0U, 4294967259U

Casting

Explicit casting between signed and unsigned:

```
int tx, ty;
unsigned int ux, uy;
tx = (int) ux;
uy = (unsigned int) ty;
```

Implicit casting due to assignments and other expressions:

```
tx = ux;
uy = ty;
```

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
- Including comparison operations <, >, ==, <=, >=.
- Examples for w = 32: TMIN = -2, 147, 483, 648, <math>TMAX = 2, 147, 483, 647:

Const LHS	Relation	Const RHS	Evaluation
0	==	0U	

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
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Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned

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Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
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Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
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- Examples for w = 32: TMIN = -2, 147, 483, 648, <math>TMAX = 2, 147, 483, 647:

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-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	

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Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned

Evaluation

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w = 32: TMIN = -2, 147, 483, 648, TMAX = 2, 147, 483, 647:

Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned
-1	>	-2	

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
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- Examples for w = 32: TMIN = -2, 147, 483, 648, <math>TMAX = 2, 147, 483, 647:

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2147483647U	<	-2147483647-1	unsigned
-1	>	-2	signed

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- Examples for w = 32: TMIN = -2, 147, 483, 648, <math>TMAX = 2, 147, 483, 647:

Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned
-1	>	-2	signed
(unsigned int)-1	>	-2	

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
- Including comparison operations <, >, ==, <=, >=.
- Examples for $w = 32 \cdot TMIN = -2 \cdot 147 \cdot 483 \cdot 648 \cdot TM$

$\omega = 32 \cdot TMINI -$	2 147 493 649	TMAX = 2, 147, 483, 647:
VV = 32.1141111 =	-2, 177, 700, 070,	IMAX = 2, 171, 700, 071.

Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned
-1	>	-2	signed
(unsigned int)-1	>	-2	unsigned

Evaluation

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- Examples forX = 32: TMINI

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Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned
-1	>	-2	signed
(unsigned int)-1	>	-2	unsigned
2147483647	<	2147483648U	

Evaluation

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
- Including comparison operations <, >, ==, <=, >=.
- Examples for
 w 32: TMIN 2 147 4

w = 32: TMIN = -2, 147, 483, 648, <math>TMAX = 2, 147, 483, 647:

Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned
-1	>	-2	signed
(unsigned int)-1	>	-2	unsigned
2147483647	<	2147483648U	unsigned

Evaluation

- If there is a mix of unsigned and signed in single expression, *signed* values implicitly cast to unsigned.
- Including comparison operations <, >, ==, <=, >=.
- Examples for

w = 32: TMIN = -2, 147, 483, 648, <math>TMAX = 2, 147, 483, 647:

Const LHS	Relation	Const RHS	Evaluation
0	==	0U	unsigned
-1	<	0	signed
-1	>	0U	unsigned
2147483647	>	-2147483647-1	signed
2147483647U	<	-2147483647-1	unsigned
-1	>	-2	signed
(unsigned int)-1	>	-2	unsigned
2147483647	<	2147483648U	unsigned
2147483647	>	(int) 2147483648U	

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Casting between signed and unsigned: basic rules

- Bit pattern is maintained.
- ...but reinterpreted.
- Can have unexpected effects: adding or subtracting 2^w .
- Expression containing signed and unsigned int:
 - int is cast to unsigned int!
 - When can this go bad?

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for (unsigned int i = n-1; i \ge 0; i--) {

// do something with x[i]
}
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```
for (unsigned int i = n-1; i >= 0; i--) {
   // do something with x[i]
}
```

Advice: Never do arithmetic on unsigned types—only use them for bit operations.

But: Some C operators (sizeof) and many functions return unsigned types (e.g. size_t).

Floating point arithmetic

Biased numbers

Background: Fractional binary numbers

IEEE floating point standard

Examples and properties

Rounding, addition, and multiplication

Floating point in C

Summary

Integers

Arithmetic

Conversion, casting

Expansion and truncation

Truncation

Task

- Given k + w-bit signed integer x.
- Convert it to w-bit integer x' with same value i possible.

Approach

- Remove the *k* most significant bits.
- Equivalent to computing $x' = x \mod 2^w$.
- Can cause numerical change if number has no representation in w bits.
- Otherwise safe.

W	Bits	Two's complement
8	11111111 ₂	-1_{10}
4	1111 ₂	-1_{10}
8	100000002	-128_{10}
4	00002	0 ₁₀

Sign extension

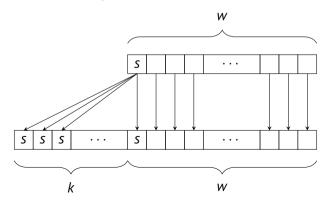
Task

- Given *w*-bit signed integer *x*.
- Convert it to w + k-bit integer x' with same value.

Approach

• Make *k* copies of sign bit (most significant bit):

$$x' = \underbrace{x_{w-1}, \dots, x_{w-1}}_{k \text{ copies of sign bit.}}, x_{w-1}, \dots, x_0$$



Sign extension example

```
short int x = 15213;
int         ix = (int) x;
short int y = -15213;
int         iy = (int) y;
```

	Decimal	Hex	Binary
Х	15213	3B 6D	0011 1011 0110 1101
ix	15213	00 00 3B 6D	0000 0000 0000 0000 0011 1011 0110 1101
У	-15213	C4 93	1100 0100 1001 0011
iy	-15213	FF FF C4 93	1111 1111 1111 1111 1100 0100 1001 0011

Summary: basic rules for expanding and truncating

Expanding (e.g. short to int)

- Unsigned: zeros added.
- Signed: sign extension.
- Both yield expected result.

Truncating (e.g. unsigned int to unsigned short)

- Bits are truncated.
- Result reinterpreted.
- Unsigned: modulo operation.
- Signed: similar to a modulo operation.
- For small numbers yield expected behaviour.