

CS 33

Data Representation (Part 2)

Many of the slides in this lecture are either from or adapted from slides provided by the authors of the textbook “Computer Systems: A Programmer’s Perspective.” 2nd Edition and are provided from the website of Carnegie-Mellon University, course 15-213, taught by Randy Bryant and David O’Hallaron in Fall 2010. These slides are indicated “Supplied by CMU” in the notes section of the slides.

Why Should I Use Unsigned?

- **Don't use just because number nonnegative**

- easy to make mistakes

```
unsigned i;  
for (i = cnt-2; i >= 0; i--)  
    a[i] += a[i+1];
```

- can be very subtle

```
#define DELTA sizeof(int)  
int i;  
for (i = CNT; i-DELTA >= 0; i-= DELTA)  
    . . .
```

- **Do use when using bits to represent sets**

- logical right shift, no sign extension

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Note that “sizeof” returns an unsigned value. (Recall that, when mixing signed and unsigned items in an expression, the result will be unsigned.)

Word Size

- **(Mostly) obsolete term**
 - old computers had items of one size: the word size
- **Now used to express the number of bits necessary to hold an address**
 - 16 bits (really old computers)
 - 32 bits (old computers)
 - 64 bits (most current computers)

Byte Ordering

- **Four-byte integer**
 - 0x76543210
- **Stored at location 0x100**
 - which byte is at 0x100?
 - which byte is at 0x103?

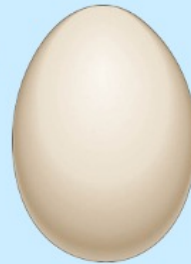


10	32	54	76
0x100	0x101	0x102	0x103

Little-endian

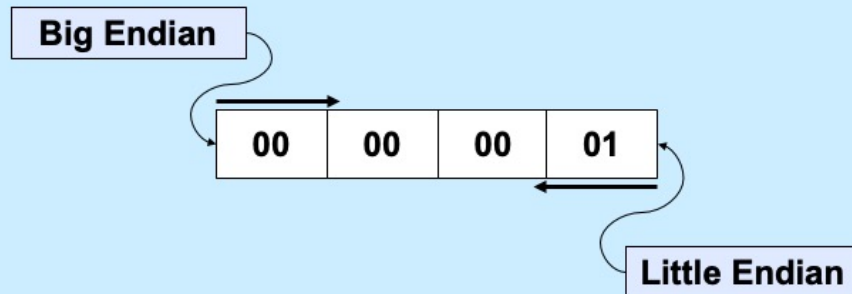
76	54	32	10
0x100	0x101	0x102	0x103

Big-endian



Read “Gulliver’s Travels” by Jonathan Swift for an explanation of the egg.

Byte Ordering (2)



Here we have a four-byte integer one. In the big-endian representation, the address of the integer is the address of the byte containing its most-significant bits (the big end), while in the little-endian representation, the address of the integer is the address of the byte containing its least-significant bits (the little end). Suppose we pass a pointer to this integer to some procedure. However, in a type-mismatch, the procedure assumes that what is passed it is a two-byte integer. On a big-endian system, it would think it was passed a zero, but on a little-endian system, it would think it was passed a one.

This is not an argument in favor of either approach, but simply an observation that behaviors could be different.

Quiz 1

```
int main() {  
    long x=1;  
    func((int *)&x);  
    return 0;  
}  
  
void func(int *arg) {  
    printf("%d\n", *arg);  
}
```

**What value is printed
on a big-endian 64-bit
computer?**

- a) 1
- b) 0
- c) 2^{32}
- d) $2^{32}-1$

Which Byte Ordering Do We Use?

```
int main() {  
    unsigned int x = 0x03020100;  
    unsigned char *xarray = (unsigned char *)&x;  
    for (int i=0; i<4; i++) {  
        printf("%02x", xarray[i]);  
    }  
    printf("\n");  
    return 0;  
}
```

Possible results:

00010203
03020100

This code prints out the value of `x`, one byte at a time, starting with the byte at the lowest address (little end). On x86-based computers, it will print:

00010203

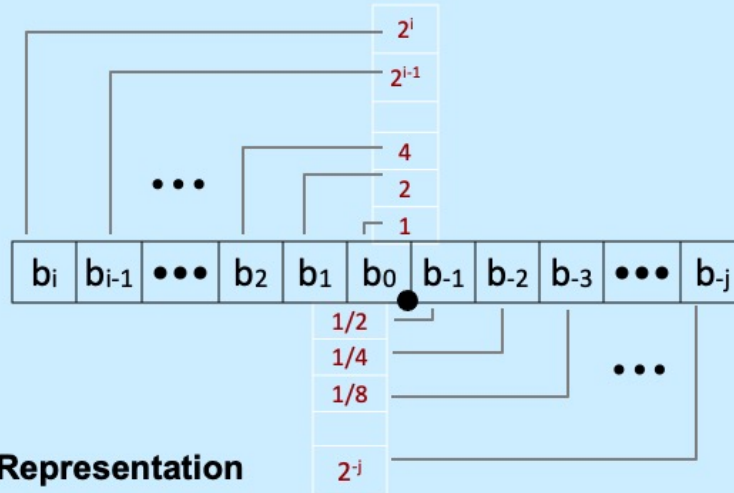
which means that the address of an `int` is the address of the byte containing its least significant digits (little endian).

Fractional binary numbers

- What is 1011.101_2 ?

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Fractional Binary Numbers



• Representation

- bits to right of “binary point” represent fractional powers of 2
- represents rational number:

$$\sum_{k=-j}^i b_k \times 2^k$$

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Representable Numbers

- **Limitation #1**

- can exactly represent only numbers of the form $n/2^k$
 - » other rational numbers have repeating bit representations
- value representation
 - » 1/3 0.0101010101[01] $_{\dots 2}$
 - » 1/5 0.001100110011[0011] $_{\dots 2}$
 - » 1/10 0.0001100110011[0011] $_{\dots 2}$

- **Limitation #2**

- just one setting of decimal point within the w bits
 - » limited range of numbers (very small values? very large?)

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IEEE Floating Point

- **IEEE Standard 754**
 - established in 1985 as uniform standard for floating point arithmetic
 - » before that, many idiosyncratic formats
 - supported on all major CPUs
- **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

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IEEE is the Institute for Electrical and Electronics Engineers (pronounced "eye triple e").

Floating-Point Representation

- Numerical Form:

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

- sign bit **s** determines whether number is negative or positive
- significand **M** normally a fractional value in range [1.0,2.0)
- exponent **E** weights value by power of two

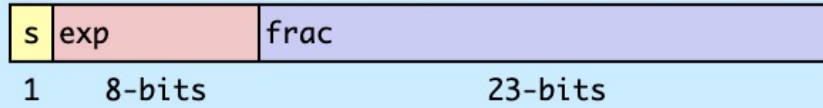
- Encoding

- MSB **s** is sign bit **s**
- exp field encodes **E** (but is not equal to E)
- frac field encodes **M** (but is not equal to M)

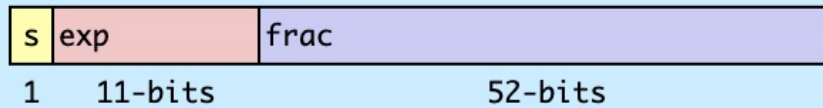


Precision options

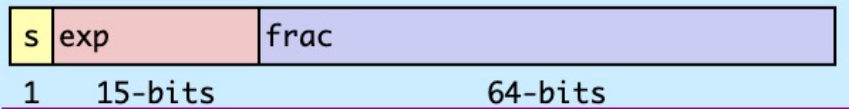
- **Single precision: 32 bits**



- **Double precision: 64 bits**



- **Extended precision: 80 bits (Intel only)**



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On x86 hardware, all floating-point arithmetic is done with 80 bits, then reduced to either 32 or 64 as required.

“Normalized” Values

- **When:** $\text{exp} \neq 000\dots 0$ and $\text{exp} \neq 111\dots 1$
- **Exponent coded as biased value:** $E = \text{Exp} - \text{Bias}$
 - exp : unsigned value exp
 - $\text{bias} = 2^{k-1} - 1$, where k is number of exponent bits
 - » single precision: 127 (Exp: 1...254, E: -126...127)
 - » double precision: 1023 (Exp: 1...2046, E: -1022...1023)
- **Significand coded with implied leading 1:** $M = 1.\text{xxx}\dots\text{x}_2$
 - $\text{xxx}\dots\text{x}$: bits of frac
 - minimum when $\text{frac} = 000\dots 0$ ($M = 1.0$)
 - maximum when $\text{frac} = 111\dots 1$ ($M = 2.0 - \epsilon$)
 - get extra leading bit for “free”

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Normalized Encoding Example

- **Value:** float $F = 15213.0$;

$$\begin{aligned} - 15213_{10} &= 11101101101101_2 \\ &= 1.1101101101101_2 \times 2^{13} \end{aligned}$$

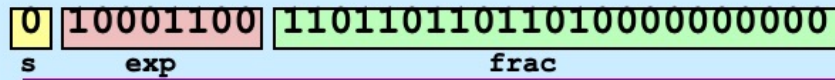
- **Significand**

$$\begin{aligned} M &= 1.\underline{1101101101101}_2 \\ \text{frac} &= \underline{1101101101101}0000000000_2 \end{aligned}$$

- **Exponent**

$$\begin{aligned} E &= 13 \\ \text{bias} &= 127 \\ \text{exp} &= 140 = 10001100_2 \end{aligned}$$

- **Result:**



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Denormalized Values

- **Condition:** $\text{exp} = 000\dots 0$
- **Exponent value:** $E = -\text{Bias} + 1$ (instead of $E = 0 - \text{Bias}$)
- **Significand coded with implied leading 0:**
 $M = 0.\text{xxx}\dots\text{x}_2$
 - $\text{xxx}\dots\text{x}$: bits of frac
- **Cases**
 - $\text{exp} = 000\dots 0, \text{frac} = 000\dots 0$
 - » represents zero value
 - » note distinct values: $+0$ and -0 (why?)
 - $\text{exp} = 000\dots 0, \text{frac} \neq 000\dots 0$
 - » numbers closest to 0.0
 - » equispaced

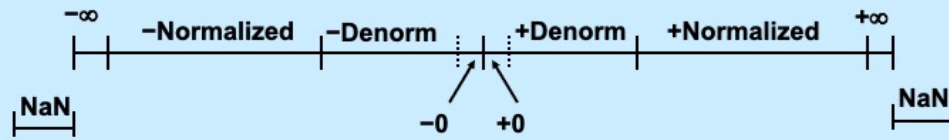
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Special Values

- **Condition: $\text{exp} = 111\dots 1$**
- **Case: $\text{exp} = 111\dots 1, \text{frac} = 000\dots 0$**
 - represents value ∞ (infinity)
 - operation that overflows
 - both positive and negative
 - e.g., $1.0/0.0 = -1.0/-0.0 = +\infty$, $1.0/-0.0 = -\infty$
- **Case: $\text{exp} = 111\dots 1, \text{frac} \neq 000\dots 0$**
 - not-a-number (NaN)
 - represents case when no numeric value can be determined
 - e.g., $\text{sqrt}(-1)$, $\infty - \infty$, $\infty \times 0$

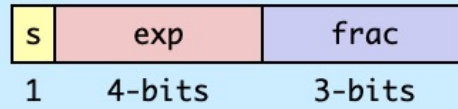
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Visualization: Floating-Point Encodings



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Tiny Floating-Point Example



- **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
 - the last three bits are the *frac*
- **Same general form as IEEE Format**
 - normalized, denormalized
 - representation of 0, NaN, infinity

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Dynamic Range (Positive Only)

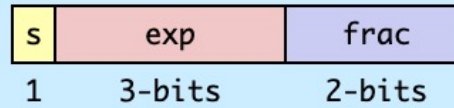
	s	exp	frac	E	Value	
Denormalized numbers	0	0000	000	-6	0	
	0	0000	001	-6	$1/8 * 1/64 = 1/512$	closest to zero
	0	0000	010	-6	$2/8 * 1/64 = 2/512$	
	...					
	0	0000	110	-6	$6/8 * 1/64 = 6/512$	
	0	0000	111	-6	$7/8 * 1/64 = 7/512$	largest denorm
Normalized numbers	0	0001	000	-6	$8/8 * 1/64 = 8/512$	smallest 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
	0	0111	000	0	$8/8 * 1 = 1$	
	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	

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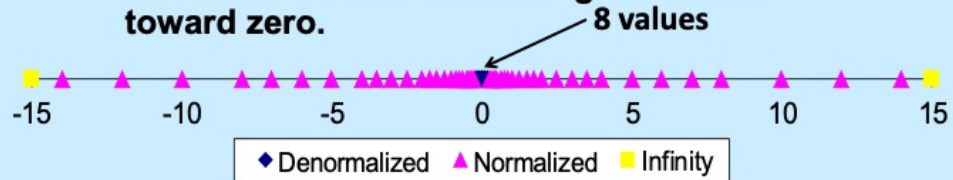
Distribution of Values

- **6-bit IEEE-like format**

- $e = 3$ exponent bits
- $f = 2$ fraction bits
- bias is $2^{3-1}-1 = 3$



- **Notice how the distribution gets denser toward zero.**

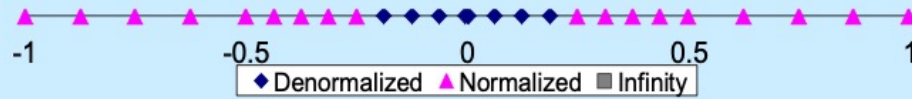
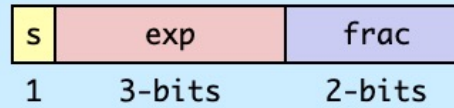


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Distribution of Values (close-up view)

- **6-bit IEEE-like format**

- **e = 3** exponent bits
- **f = 2** fraction bits
- **bias is 3**



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Quiz 1

- **6-bit IEEE-like format**

- e = 3 exponent bits
- f = 2 fraction bits
- bias is 3

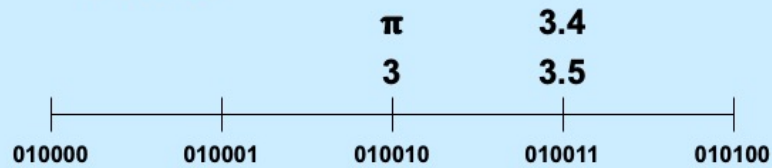


What number is represented by 0 010 10?

- a) 3
- b) 1.5
- c) .75
- d) none of the above

Mapping Real Numbers to Float

- The real number 3 is represented as
0 100 10
- The real number 3.5 is represented as
0 100 11
- How is the real number 3.4 represented?
0 100 11
- How is the real number π represented?
0 100 10



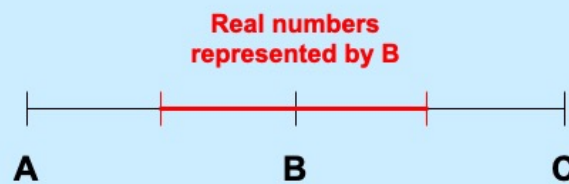
We're assuming here the six-bit floating-point format.

Mapping Real Numbers to Float

- If R is a real number, it's mapped to the floating-point number whose value is closest to R
- What if it's midway between two values?
 - rounding rules coming up soon!

Floats are Sets of Values

- If A, B, and C are successive floating-point numbers
 - e.g., 010001, 010010, and 010011
- B represents all real numbers from midway between A and B through midway between B and C



Note that we still have to discuss rounding so as to accommodate values that are equidistant from A and B or from B and C.

A special case is 0. Positive 0 represents a range of values that are greater than or equal to 0. Negative 0 represents a range of values that are less than or equal to zero.

Significance

- **Normalized numbers**
 - for a particular exponent value E and an S -bit significand, the range from 2^E up to 2^{E+1} is divided into 2^S equi-spaced floating-point values
 - » thus each floating-point value represents $1/2^S$ of the range of values with that exponent
 - » all bits of the significand are important
 - » we say that there are S significant bits – for reasonably large S , each floating-point value covers a rather small part of the range
 - high accuracy
 - for $S=23$ (32-bit float), accurate to one in 2^{23} (.0000119% accuracy)

Significance

- **Unnormalized numbers**
 - high-order zero bits of the significand aren't important
 - in 8-bit floating point, 0 0000 001 represents 2^{-9}
 - » it is the only value with that exponent: 1 significant bit (either 2^{-9} or 0)
 - 0 0000 010 represents 2^{-8}
0 0000 011 represents $1.5 \cdot 2^{-8}$
 - » only two values with exponent -8: 2 significant bits (encoding those two values, as well as 2^{-9} and 0)
 - fewer significant bits mean less accuracy
 - 0 0000 001 represents a range of values from $.5 \cdot 2^{-9}$ to $1.5 \cdot 2^{-9}$
 - 50% accuracy

Recall that the bias for the exponent of 8-bit IEEE FP is 7, thus for unnormalized numbers the actual exponent is -6 (-bias+1). The significand has an implied leading 0, thus 0 0000 001 represents $2^{-6} \cdot 2^{-3}$.

With 8-bit IEEE FP, the value 0 0000 01 is interpreted as 2^{-9} , But the number represented could be 50% or 50% more.

Floating-Point Operations: Basic Idea

- $x \oplus_f y = \text{Round}(x \oplus y)$
- $x \otimes_f y = \text{Round}(x \otimes y)$
- **Basic idea**
 - first **compute exact result**
 - make it fit into desired precision
 - » possibly overflow if exponent too large
 - » possibly **round to fit into frac**

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Rounding

- Rounding modes (illustrated with \$ rounding)

	\$1.40	\$1.60	\$1.50	\$2.50	-\$1.50
towards zero	\$1	\$1	\$1	\$2	-\$1
round down ($-\infty$)	\$1	\$1	\$1	\$2	-\$2
round up ($+\infty$)	\$2	\$2	\$2	\$3	-\$1
nearest integer	\$1	\$2	?	?	?
nearest even (default)	\$1	\$2	\$2	\$2	-\$2

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Floating-Point Multiplication

- $(-1)^{s_1} M_1 2^{E_1} \times (-1)^{s_2} M_2 2^{E_2}$
- **Exact result:** $(-1)^s M 2^E$
 - sign s : $s_1 \wedge s_2$
 - significand M : $M_1 \times M_2$
 - exponent E : $E_1 + E_2$
- **Fixing**
 - if $M \geq 2$, shift M right, increment E
 - if E out of range, overflow (or underflow)
 - round M to fit *frac* precision
- **Implementation**
 - biggest chore is multiplying significands

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Note that to compute E , one must first convert exp_1 and exp_2 to E_1 and E_2 , then add them together and check for underflow or overflow (corresponding to $-\infty$ and $+\infty$), and then convert to exp .

Floating-Point Addition

- $(-1)^{s_1} M_1 2^{E_1} + (-1)^{s_2} M_2 2^{E_2}$

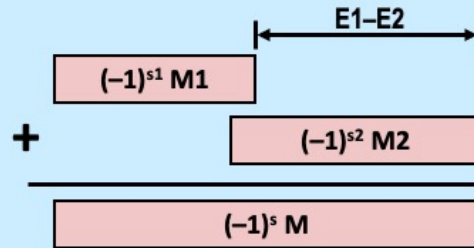
–assume $E_1 > E_2$

- **Exact result:** $(-1)^s M 2^E$

–sign s , significand M :

» result of signed align & add

–exponent E : E_1



- **Fixing**

–if $M \geq 2$, shift M right, increment E

–if $M < 1$, shift M left k positions, decrement E by k

–overflow if E out of range

–round M to fit **frac** precision

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Note that, by default, overflow results in either $+\infty$ or $-\infty$.