## Floating Point II

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http://www.smbc-comics.com/?id=2999

## Relevant Course Information

* hw5 due Monday (4/11) @ 11:59 pm
* Lab 1a due TONIGHT (4/11) @ 11:59 pm
- Submit pointer.c and lab1Asynthesis.txt
- Make sure you check the Gradescope autograder output!
- Can use late day tokens to submit up until Wed 11:59 pm
* Lab 1b, due 4/18
- Submit aisle_manager.c, store_client.c, and lab1Bsynthesis.txt


## Getting Help with 351

* Lecture recordings, readings, inked slides, textbook readings
* Form a study group!
- Good for everything but labs, which should be done in pairs
- Communicate regularly, use the class terminology, ask and answer each others' questions, show up to OH together
* Attend office hours
- Use the OH queue, but can also chat with other students there - help each other learn!
* Post on Ed Discussion
* Request a 1-on-1 meeting
- Available on a limited basis for special circumstances


## Reading Review

* Terminology:
- Special cases
- Denormalized numbers
- $\pm \infty$
- Not-a-Number (NaN)
- Limits of representation
- Overflow
- Underflow
- Rounding


## Review Questions

* What is the value of the following floats?
- 0x00000000
- 0xFF800000
* For the following code, what is the smallest value of $n$ that will encounter a limit of representation?

$$
\begin{aligned}
& \text { float } f=1.0 ; / / 2 \wedge 0 \\
& \text { for }(\text { int } i=0 ; i<n ;++i) \\
& f *=1024 ; / / 1024=2 \wedge 10 \\
& \text { printf("f }=\% \text { f } \backslash n ", f) ;
\end{aligned}
$$

## Floating Point Encoding Summary (Review)

| $\mathbf{E}$ | $\mathbf{M}$ | Meaning |
| :---: | :---: | :---: |
| $0 \times 00$ | 0 | $\pm 0$ |
| $0 \times 00$ | non-zero | $\pm$ denorm num |
| $0 \times 01-0 \times F E$ | anything | $\pm$ norm num |
| $0 \times F F$ | 0 | $\pm \infty$ |
| $0 \times F F$ | non-zero | NaN |

## Special Cases

* But wait... what happened to zero?
- Special case: E and M all zeros $=0$
- Two zeros! But at least $0 x 00000000=0$ like integers
* $\mathrm{E}=0 \times \mathrm{FFF}, \mathrm{M}=0: \pm \infty$
- e.g., division by 0
- Still work in comparisons!
$* E=0 x F F, \mathrm{M} \neq 0$ : Not a Number ( NaN )
- e.g., square root of negative number, $0 / 0, \infty-\infty$
- NaN propagates through computations
- Value of $M$ can be useful in debugging


## New Representation Limits

* New largest value (besides $\infty$ )?
- $\mathrm{E}=0 \times \mathrm{FF}$ has now been taken!
- $\mathrm{E}=0 \times \mathrm{FE}$ has largest: $1.1 \ldots 1_{2} \times 2^{127}=2^{128}-2^{104}$
* New numbers closest to 0:
- $\mathrm{E}=0 \times 00$ taken; next smallest is $\mathrm{E}=0 \times 01$
- $a=1.0 . . .00_{2} \times 2^{-126}=2^{-126}$
- $b=1.0 \ldots 01_{2} \times 2^{-126}=2^{-126}+2^{-149}$
- Normalization and implicit 1 are to blame
- Special case: $\mathrm{E}=0, \mathrm{M} \neq 0$ are denormalized numbers


## Denorm Numbers

* Denormalized numbers
- No leading 1
- Uses implicit exponent of -126 even though $E=0 x 00$
* Denormalized numbers close the gap between zero and the smallest normalized number
- Smallest norm: $\pm 1.0 . . .00_{\mathrm{two}} \times 2^{-126}= \pm 2^{-126}$

So much

- Smallest denorm: $\pm 0.0 \ldots 01_{\text {two }} \times 2^{-126}= \pm 2^{-149}$
- There is still a gap between zero and the smallest denormalized number


## Floating Point Interpretation Flow Chart


$\square$ = special case

## 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...


## Tiny Floating Point Representation

* We will use the following 8-bit floating point representation to illustrate some key points:

| $S$ | $E$ | $M$ |
| :---: | :---: | :---: |
| 1 | 4 | 3 |

* Assume that it has the same properties as IEEE floating point:
- bias =
- encoding of $-0=$
- encoding of $+\infty=$
- encoding of the largest (+) normalized \# =
- encoding of the smallest (+) normalized \# =


## Distribution of Values (Review)

* What ranges are NOT representable?
- Between largest norm and infinity Overflow (Exp too large)
- Between zero and smallest denorm Underflow (Exp too small)
- Between norm numbers? Rounding
* Given a FP number, what's the bit pattern of the next largest representable number?
- What is this "step" when Exp $=0$ ?
- What is this "step" when Exp $=100$ ?
* Distribution of values is denser toward zero



## Floating Point Rounding

This is extra (non-testable) material

* The IEEE 754 standard actually specifies different rounding modes:
- Round to nearest, ties to nearest even digit
- Round toward $+\infty$ (round up)
- Round toward $-\infty$ (round down)
- Round toward 0 (truncation)
* In our tiny example:
- Man = 1.00101 rounded to $\mathrm{M}=0 \mathrm{0b001}$
- Man = 1.00111 rounded to $\mathrm{M}=0 \mathrm{Ob} 010$
- Man = 1.00110 rounded to $M=0 b 010$
- Man = 1.00010 rounded to $\mathrm{M}=0 \mathrm{Ob00}$


## Floating Point Operations: Basic Idea

$$
\text { Value }=(-1)^{S} \times \text { Mantissa } \times 2^{\text {Exponent }}
$$

| $S$ | E | M |
| :---: | :---: | :---: |

$* x+_{f} Y=R o u n d(x+y)$
$* x *_{f} y=\operatorname{Round}(x * y)$

* Basic idea for floating point operations:
- First, compute the exact result
- Then round the result to make it fit into the specified precision (width of M)
- Possibly over/underflow if exponent outside of range


## Mathematical Properties of FP Operations

* Overflow yields $\pm \infty$ and underflow yields 0
* Floats with value $\pm \infty$ and NaN can be used in operations
- Result usually still $\pm \infty$ or NaN, but not always intuitive
* Floating point operations do not work like real math, due to rounding
- Not associative: (3.14+1e100)-1e100 !=3.14+(1e100-1e100)

0
3.14

- Not distributive: $100 *(0.1+0.2) \quad!=100 * 0.1+100 * 0.2$

$$
30.000000000000003553
$$

- Not cumulative
- Repeatedly adding a very small number to a large one may do nothing


## Floating Point Encoding Flow Chart



## Limits of Interest

* The following thresholds will help give you a sense of when certain outcomes come into play, but don't worry about the specifics:
- FOver $=2^{\text {bias+1 }}=2^{8}$
- This is just larger than the largest representable normalized number
- FDenorm $=2^{1-\text { bias }}=2^{-6}$
- This is the smallest representable normalized number
- FUnder $=2^{1-\text { bias-m }}=2^{-9}$
- $m$ is the width of the mantissa field
- This is the smallest representable denormalized number


## Floating Point in C

* Two common levels of precision:

| float | 1.0 f | single precision (32-bit) |
| :--- | :--- | :--- |
| double | 1.0 | double precision (64-bit) |

* \#include <math.h> to get INFINITY and NAN constants
*. Equality (==) comparisons between floating point numbers are tricky, and often return unexpected results, so just avoid them!


## Floating Point Conversions in C

* Casting between int, float, and double changes the bit representation
- int $\rightarrow$ float
- May be rounded (not enough bits in mantissa: 23)
- Overflow impossible
- int or float $\rightarrow$ double
- Exact conversion (all 32-bit ints representable)
- long $\rightarrow$ double
- Depends on word size (32-bit is exact, 64-bit may be rounded)
- double or float $\rightarrow$ int
- Truncates fractional part (rounded toward zero)
- "Not defined" when out of range or NaN: generally sets to Tmin (even if the value is a very big positive)


## Exploration Question

* We execute the following code in C. How many bytes are the same (value and position) between $i$ and $f$ ?

```
int i = 384; // 2^8 + 2^7
float f = (float) i;
```

A. 0 bytes
B. 1 byte
C. 2 bytes
D. 3 bytes
E. We're lost...

## Floating Point Summary

* Floats also suffer from the fixed number of bits available to represent them
- Can get overflow/underflow
- "Gaps" produced in representable numbers means we can lose precision, unlike ints
- Some "simple fractions" have no exact representation (e.g. 0.2)
- "Every operation gets a slightly wrong result"
* Floating point arithmetic not associative or distributive
- Mathematically equivalent ways of writing an expression may compute different results
* Never test floating point values for equality!
* Careful when converting between ints and floats!


## Number Representation Really Matters

* 1991: Patriot missile targeting error
- clock skew due to conversion from integer to floating point
* 1996: Ariane 5 rocket exploded (\$1 billion)
- overflow converting 64-bit floating point to 16-bit integer
* 2000: Y2K problem
- limited (decimal) representation: overflow, wrap-around
* 2038: Unix epoch rollover
- Unix epoch = seconds since 12am, January 1, 1970
- signed 32-bit integer representation rolls over to TMin in 2038
* Other related bugs:
- 1982: Vancouver Stock Exchange 10\% error in less than 2 years
- 1994: Intel Pentium FDIV (floating point division) HW bug (\$475 million)
- 1997: USS Yorktown "smart" warship stranded: divide by zero
- 1998: Mars Climate Orbiter crashed: unit mismatch (\$193 million)


## Summary

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| $0 \times F F$ | 0 | $\pm \infty$ |
| $0 \times F F$ | non-zero | NaN |

* Floating point encoding has many limitations
- Overflow, underflow, rounding
- Rounding is a HUGE issue due to limited mantissa bits and gaps that are scaled by the value of the exponent
- Floating point arithmetic is NOT associative or distributive
* Converting between integral and floating point data types does change the bits


An example that applies the IEEE Floating Point concepts to a smaller (8-bit) representation scheme. These slides expand on material covered today, so while you don't need to read these, the information is "fair game."

## Tiny Floating Point Example



* 8-bit Floating Point Representation
- The sign bit is in the most significant bit (MSB)
- The next four bits are the exponent, with a bias of $2^{4-1}-1=7$
- The last three bits are the mantissa
* Same general form as IEEE Format
- Normalized binary scientific point notation
- Similar special cases for 0 , denormalized numbers, $\mathrm{NaN}, \infty$


## Dynamic Range (Positive Only)

|  | S | E | M | Exp | Value |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Denormalized numbers | 0 | 0000 | 000 | -6 | 0 | closest to zero |
|  | 0 | 0000 | 001 | -6 | $1 / 8 * 1 / 64=1 / 512$ |  |
|  | 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 | $9 / 8 * 1 / 64=9 / 512$ |  |
|  | $\ldots$ |  |  |  |  |  |
|  | 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 | $\mathrm{n} / \mathrm{a}$ | inf |  |

## Special Properties of Encoding

* Floating point zero ( $0^{+}$) exactly the same bits as integer zero
- All bits $=0$
* Can (Almost) Use Unsigned Integer Comparison
- Must first compare sign bits
- Must consider $0^{-}=0^{+}=0$
- NaNs problematic
- Will be greater than any other values
- What should comparison yield?
- Otherwise OK
- Denorm vs. normalized
- Normalized vs. infinity

