# CS61C - Machine Structures <br> Lecture 16 - Floating Point Numbers II 

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## IEEE 754 Floating Point Standard (review)

${ }^{\circ}$ Biased Notation, where bias is number subtracted to get real number

- IEEE 754 uses bias of 127 for single precision
- Subtract 127 from Exponent field to get actual value for exponent
- 1023 is bias for double precision
${ }^{\circ}$ Summary (single precision):
$3130 \quad 2322$

| S | Exponent | Significand |
| :--- | :--- | :--- |

1 bit 8 bits 23 bits
$(-1)^{\mathrm{S}} \times\left(1+\right.$ Significand) $\times 2^{(\text {Exponent-127) }}$
Double precision identical, except with exponent bias of 1023

## Example: Converting Binary FP to Decimal

| 0 | 01101000 | 10101010100001101000010 |
| :--- | :--- | :--- | :--- |

Sign: 0 => positive
${ }^{\circ}$ Exponent:

- $01101000_{\text {two }}=104_{\text {ten }}$
- Bias adjustment: 104-127=-23
${ }^{\circ}$ Significand:
$1+1 \times 2^{-1}+0 \times 2^{-2}+1 \times 2^{-3}+0 \times 2^{-4}+1 \times 2^{-5}+\ldots$ $=1+2^{-1}+2^{-3}+2^{-5}+2^{-7}+2^{-9}+2^{-14}+2^{-15}+2^{-17}+2^{-22}$ $=1.0+0.666115$
${ }^{\circ}$ Represents: $1.666115_{\text {ten }}{ }^{*} 2^{-23} \sim 1.986 * 10^{-7}$
(about 2/10,000,000)


## Example: Converting Decimal to FP

$-2.340625 \times 10^{1}$

1. Denormalize: - 23.40625
2. Convert integer part:

$$
23=16+(7=4+(3=2+(1)))=10111_{2}
$$

3. Convert fractional part:
$.40625=.25+(.15625=.125+(.03125))=.01101_{2}$
4. Put parts together and normalize:
$10111.01101=1.011101101 \times 2^{4}$
5. Convert exponent: $127+4=10000011_{2}$

\section*{| 1 | 10000011 | 01110110100000000000000 |
| :--- | :--- | :--- |}

## Representation for +/- Infinity

${ }^{\circ}$ In FP, divide by zero should produce +/- infinity, not overflow.
${ }^{\circ}$ Why?

- OK to do further computations with infinity e.g., $\mathrm{X} / 0$ > Y may be a valid comparison
${ }^{\circ}$ IEEE 754 represents +/- infinity
- Largest positive exponent reserved for infinity
- Significands all zeroes


## Representation for 0

${ }^{\circ}$ Represent 0 ?

- exponent all zeroes
- significand all zeroes
- What about sign? Both cases valid.
+0: 00000000000000000000000000000000
-0: 10000000000000000000000000000000


## Special Numbers

${ }^{\circ}$ What have we defined so far? (Single Precision)

| Exponent | Significand | Object |
| :--- | :--- | :--- |
| 0 | 0 | 0 |
| 0 | nonzero | ??? |
| $1-254$ | anything | +/- fl. pt. \# |
| 255 | 0 | +/- infinity |
| 255 | nonzero | ??? |

" ${ }^{\text {"Wrofessor Kahan had clever ideas; }}$ "Waste not, want not"

- We'll talk about Exp=0,255 \& Sig!=0 later


## Precision and Accuracy

Don't confuse these two terms!
Precision is a count of the number bits in a computer word used to represent a value.

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;
pi will be represented using all 24 bits of the significant (highly precise), but is only an approximation (not accurate).

## Administrivia

${ }^{\circ}$ Midterm 1, 1 Pimentel, Tonight 6-8pm sharp

- Open Book/Notes, but no electronic devices of any kind!
© Don't forget to work on homework and start project 3 over the weekend.

Representation for Not a Number
${ }^{\circ}$ What do I get if I calculate sqrt(-4.0) or 0/0?

- If infinity is not an error, these shouldn't be either.
- Called Not a Number (NaN)
- Exponent = 255, Significand nonzero
${ }^{\circ}$ Why is this useful?
- Hope NaNs help with debugging?
- They contaminate: $\mathrm{Op}(\mathrm{NaN}, \mathrm{X})=\mathrm{NaN}$


## Special Numbers (cont'd)

${ }^{\circ}$ What have we defined so far? (Single Precision)?

| Exponent | Significand | Object |
| :--- | :--- | :--- |
| 0 | 0 | 0 |
| 0 | nonzero | ??? |
| $1-254$ | anything | +/- fl. pt. \# |
| 255 | 0 | $+/-$ infinity |
| 255 | nonzero | NaN |

Representation for Denorms (1/2)
${ }^{\circ}$ Problem: There's a gap among representable FP numbers around 0

- Smallest representable pos num:

$$
a=1.0 \ldots 2^{*} 2^{-126}=2^{-126}
$$

- Second smallest representable pos num:

$$
\begin{aligned}
& b=1.000 \ldots \ldots .1_{2} * 2^{-126}=2^{-126}+2^{-149} \\
& a-0=2^{-126} \\
& b-a=2^{-149}
\end{aligned}
$$



## Representation for Denorms (2/2)

## ${ }^{\circ}$ Solution:

- We still haven't used Exponent = 0, Significand nonzero
- Denormalized number: no (implied) leading 1, exponent $=-126$.
- Smallest representable pos num:

$$
a=2^{-149}
$$

- Second smallest representable pos num:

$$
\begin{aligned}
& b=2^{-148} \\
& \quad-\infty \leftrightarrows+\infty
\end{aligned}
$$



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## Rounding

${ }^{\circ}$ When we perform math on real numbers, we have to worry about rounding to fit the result in the significant field.
${ }^{\circ}$ The FP hardware carries two extra bits of precision, and then round to get the proper value
${ }^{\circ}$ Rounding also occurs when converting: double to a single precision value, or floating point number to an integer

## IEEE FP Rounding Modes

${ }^{\circ}$ Round towards +infinity
-ALWAYS round "up": $2.001 \rightarrow 3$
-2.001 $\rightarrow$-2
${ }^{\circ}$ Round towards -infinity

- ALWAYS round "down": $1.999 \rightarrow$ 1,
$-1.999 \rightarrow-2$
${ }^{\circ}$ Truncate
- Just drop the last bits (round towards 0)
${ }^{\circ}$ Round to (nearest) even
- Normal rounding, almost


## Round to Even

${ }^{\circ}$ Round like you learned in grade school
${ }^{\circ}$ Except if the value is right on the borderline, in which case we round to the nearest EVEN number
$2.5 \rightarrow 2$
$3.5 \rightarrow 4$
${ }^{\circ}$ Insures fairness on calculation

- This way, half the time we round up on tie, the other half time we round down
- Tends to balance out inaccuracies

This is the default rounding mode

## Casting floats to ints and vice versa

(int) floating point expression
Coerces and converts it to the nearest integer (C uses truncation)

$$
i=(\text { int })(3.14159 * f) ;
$$

(float) expression converts integer to nearest floating point $\mathrm{f}=\mathrm{f}+$ (float) i ;

```
int }->\mathrm{ float }->\mathrm{ int
if (i == (int)((float) i)) {
    printf("true");
    }
    * Will not always print "true"
    `}\mathrm{ Most large values of integers don't
        have exact floating point
        representations
    * What about double?
```

float $\rightarrow$ int $\rightarrow$ float

$$
\begin{aligned}
& \text { if (f == (float) ((int) f)) \{ } \\
& \text { printf("true"); } \\
& \} \\
& { }^{\circ} \text { Will not always print "true" } \\
& \text { Small floating point numbers (<1) } \\
& \text { don't have integer representations } \\
& { }^{\circ} \text { For other numbers, rounding errors }
\end{aligned}
$$

## Floating Point Fallacy

${ }^{\circ} \mathrm{FP}$ add associative: FALSE!
$\cdot x=-1.5 \times 10^{38}, y=1.5 \times 10^{38}$, and $z=1.0$
$\cdot x+(y+z)=-1.5 \times 10^{38}+\left(1.5 \times 10^{38}+1.0\right)$
$=-1.5 \times 10^{38}+\left(1.5 \times 10^{38}\right)=\underline{0.0}$
$\cdot(x+y)+z=\left(-1.5 \times 10^{38}+1.5 \times 10^{38}\right)+1.0$
$=(0.0)+1.0=1.0$

## ${ }^{\circ}$ Therefore, Floating Point add is not associative!

- Why? FP result approximates real result!
- This example: $1.5 \times 10^{38}$ is so much larger than 1.0 that $1.5 \times 10^{38}+1.0$ in floating point representation is still $1.5 \times 10^{38}$


## FP Addition

${ }^{\circ}$ More difficult than with integers
${ }^{\circ}$ Can't just add significands
${ }^{\circ}$ How do we do it?

- De-normalize to match exponents
- Add significands to get resulting one
- Keep the same exponent
- Normalize (possibly changing exponent)
${ }^{\circ}$ Note: If signs differ, just perform a subtract instead.


## MIPS Floating Point Architecture (1/4)

${ }^{\circ}$ MIPS has special instructions for floating point operations:

- Single Precision:
add.s, sub.s, mul.s, div.s
- Double Precision:
add.d, sub.d, mul.d, div.d
${ }^{\circ}$ These instructions are far more complicated than their integer counterparts. They require special hardware and usually so they can take much longer to compute.


## MIPS Floating Point Architecture (2/4)

${ }^{\circ}$ Problems:

- It's inefficient to have different instructions take vastly differing amounts of time.
- Generally, a particular piece of data will not change from FP to int, or vice versa, within a program. So only one type of instruction will be used on it.
- Some programs do no floating point calculations
- It takes lots of hardware relative to integers to do Floating Point fast


## MIPS Floating Point Architecture (3/4)

${ }^{\circ} 1990$ Solution: Make a completely separate chip that handles only FP.
${ }^{\circ}$ Coprocessor 1: FP chip

- contains 32 32-bit registers: $\$ \mathrm{f} 0, \$ \mathrm{f} 1, \ldots$
- most registers specified in . s and .d instruction refer to this set
- separate load and store: lwc1 and swc1 ("load word coprocessor 1", "store ...")
- Double Precision: by convention, even/odd pair contain one DP FP number: \$f0/\$f1, \$f2/\$f3, .., \$f30/\$f31


## MIPS Floating Point Architecture (4/4)

## - 1990 Computer actually contains multiple separate chips:

- Processor: handles all the normal stuff
- Coprocessor 1: handles FP and only FP;
- more coprocessors?... Yes, later
-Today, cheap chips may leave out FP HW
${ }^{\circ}$ Instructions to move data between main processor and coprocessors:
$\cdot \mathrm{mfc} 0, \mathrm{mtc} 0, \mathrm{mfc} 1, \mathrm{mtc} 1$, etc.
${ }^{\circ}$ Appendix pages A-70 to A-74 contain many, many more FP operations.


## Things to Remember

${ }^{\circ}$ Floating Point lets us:

- Represent numbers containing both integer and fractional parts; makes efficient use of available bits.
- Store approximate values for very large and very small numbers.
${ }^{\circ}$ IEEE 754 Floating Point Standard is most widely accepted attempt to standardize interpretation of such numbers
${ }^{\circ}$ New MIPS registers(\$ $\mathbf{f 0} \mathbf{- \$ £ 3 1 )}$, instruct.:
- Single Precision ( 32 bits, $2 \times 10^{-38} \ldots 2 \times 10^{38}$ ):
add.s, sub.s, mul.s, div.s
- Double Precision ( 64 bits , $2 \times 10^{-308} \ldots 2 \times 10^{308}$ ):
$\underset{\text { Point II (26) }}{\text { add }}$ d sub. d, mu1. d, div.d

