MIPS floating-point arithmetic



- Floating-point computations are vital for many applications, but correct implementation of floating-point hardware and software is very tricky.
- Today we'll study the IEEE 754 standard for floating-point arithmetic.
 - Floating-point number representations are complex, but limited.
 - Addition and multiplication operations require several steps.
 - The MIPS architecture includes support for floating-point arithmetic.
- Machine Problem 2 will include some floating-point programming in MIPS.
- Sections this week will review the last three lectures on arithmetic.

Floating-point representation

IEEE numbers are stored using a kind of scientific notation.

 \pm mantissa × 2^{exponent}

 We can represent floating-point numbers with three binary fields: a sign bit s, an exponent field e, and a fraction field f.

| s e f | | S | e | f |
|-------|--|---|---|---|
|-------|--|---|---|---|

- The IEEE 754 standard defines several different precisions.
 - Single precision numbers include an 8-bit exponent field and a 23-bit fraction, for a total of 32 bits.
 - Double precision numbers have an 11-bit exponent field and a 52-bit fraction, for a total of 64 bits.
- There are also various extended precision formats. For example, Intel processors use an 80-bit format internally.

Sign

| s e f | |
|-------|--|
|-------|--|

- The sign bit is 0 for positive numbers and 1 for negative numbers.
- But unlike integers, IEEE values are stored in signed magnitude format.



Mantissa

| | S | e | f |
|--|---|---|---|
|--|---|---|---|

- The field **f** contains a binary fraction.
- The actual mantissa of the floating-point value is (1 + f).
 - In other words, there is an implicit 1 to the left of the binary point.
 - For example, if f is 01101..., the mantissa would be 1.01101...
- There are many ways to write a number in scientific notation, but there is always a *unique* normalized representation, with exactly one non-zero digit to the left of the point.

 $0.232 \times 10^3 = 23.2 \times 10^1 = 2.32 \times 10^2 = \dots$

 A side effect is that we get a little more precision: there are 24 bits in the mantissa, but we only need to store 23 of them.

Exponent

s e f

- The e field represents the exponent as a biased number.
 - It contains the actual exponent *plus* 127 for single precision, or the actual exponent *plus* 1023 in double precision.
 - This converts all single-precision exponents from -127 to +127 into unsigned numbers from 0 to 255, and all double-precision exponents from -1024 to +1023 into unsigned numbers from 0 to 2047.
- Two examples with single-precision numbers are shown below.
 - If the exponent is 4, the e field will be $4 + 127 = 131 (10000011_2)$.
 - If e contains 01011101 (93₁₀), the actual exponent is 93 127 = -34.
- Storing a biased exponent *before* a normalized mantissa means we can compare IEEE values as if they were signed integers.

Converting an IEEE 754 number to decimal

• The decimal value of an IEEE number is given by the formula:

 $(1 - 2s) \times (1 + f) \times 2^{e-bias}$

- Here, the s, f and e fields are assumed to be in decimal.
 - (1 2s) is 1 or -1, depending on whether the sign bit is 0 or 1.
 - We add an implicit 1 to the fraction field f, as mentioned earlier.
 - Again, the bias is either 127 or 1023, for single or double precision.

Example IEEE-decimal conversion

• Let's find the decimal value of the following IEEE number.

- First convert each individual field to decimal.
 - The sign bit s is 1.
 - The e field contains $01111100 = 124_{10}$.
 - The mantissa is $0.11000... = 0.75_{10}$.
- Then just plug these decimal values of s, e and f into our formula.

 $(1 - 2s) \times (1 + f) \times 2^{e-bias}$

• This gives us $(1 - 2) \times (1 + 0.75) \times 2^{124 - 127} = (-1.75 \times 2^{-3}) = -0.21875$.

Converting a decimal number to IEEE 754

- What is the single-precision representation of **347.625**?
 - 1. First convert the number to binary: $347.625 = 101011011.101_2$.
 - 2. Normalize the number by shifting the binary point until there is a single 1 to the left:

 $101011011.101 \times 2^{0} = 1.01011011101 \times 2^{8}$

- 3. The bits to the right of the binary point, 01011011101_2 , comprise the fractional field f.
- 4. The number of times you shifted gives the exponent. In this case, the field e should contain $8 + 127 = 135 = 10000111_2$.
- 5. The number is positive, so the sign bit is **0**.
- The final result is:

0 10000111 010110100000000000

Special values

- The smallest and largest possible exponents e=00000000 and e=11111111 (and their double precision counterparts) are reserved for special values.
- If the mantissa is always (1 + f), then how is 0 represented?
 - The fraction field f should be 0000...0000.
 - The exponent field e contains the value 0000000.
 - With signed magnitude, there are *two* zeroes: +0.0 and -0.0.
- There are representations of positive and negative infinity, which might sometimes help with instances of overflow.
 - The fraction f is 0000...0000.
 - The exponent field e is set to 11111111.
- Finally, there is a special "not a number" value, which can handle some cases of errors or invalid operations such as 0.0/0.0.
 - The fraction field **f** is set to any non-zero value.
 - The exponent e will contain 11111111.

 $(1 - 2s) \times (1 + f) \times 2^{e-127}$.

- The largest possible "normal" number is $(2 2^{-23}) \times 2^{127} = 2^{128} 2^{104}$.
 - The largest possible e is 11111110 (254).
- And the smallest positive non-zero number is 1 × 2⁻¹²⁶ = 2⁻¹²⁶.
 - The smallest e is 0000001 (1).
- In comparison, the smallest and largest possible 32-bit integers in two's complement are only -2³¹ and 2³¹ 1
- How can we represent so many more values in the IEEE 754 format, even though we use the same number of bits as regular integers?



Finiteness

- There aren't more IEEE numbers.
- With 32 bits, there are 2³²-1, or about 4 billion, different bit patterns.
 - These can represent 4 billion integers *or* 4 billion reals.
 - But there are an infinite number of reals, and the IEEE format can only represent *some* of the ones from about -2^{128} to $+2^{128}$.
- This causes enormous headaches in doing floating-point arithmetic.
 - Not all values between -2^{128} to $+2^{128}$ can be represented.
 - Small roundoff errors can quickly accumulate with multiplications or exponentiations, resulting in big errors.
 - Rounding errors can invalidate many basic arithmetic principles such as the associative law, (x + y) + z = x + (y + z).
- The IEEE 754 standard guarantees that all machines will produce the same results—but those results may not be mathematically correct!

Limits of the IEEE representation

- Even some integers cannot be represented in the IEEE format.
 - int x = 33554431; float y = 33554431; printf("%d\n", x); printf("%f\n", y);
- Some simple decimal numbers cannot be represented exactly in binary to begin with.

 $0.10_{10} = 0.0001100110011..._{2}$

- During the Gulf War in 1991, a U.S. Patriot missile failed to intercept an Iraqi Scud missile, and 28 Americans were killed.
- A later study determined that the problem was caused by the inaccuracy of the binary representation of 0.10.
 - The Patriot incremented a counter once every 0.10 seconds.
 - It multiplied the counter value by 0.10 to compute the actual time.
- However, the (24-bit) binary representation of 0.10 actually corresponds to 0.09999904632568359375, which is off by 0.000000095367431640625.
- This doesn't seem like much, but after 100 hours the time ends up being off by 0.34 seconds—enough time for a Scud to travel 500 meters!
- Professor Skeel wrote a short article about this.

Roundoff Error and the Patriot Missile. SIAM News, 25(4):11, July 1992.



Floating-point addition example

- To get a feel for floating-point operations, we'll do an addition example.
 - To keep it simple, we'll use base 10 scientific notation.
 - Assume the mantissa has four digits, and the exponent has one digit.
- The text shows an example for the addition:

```
99.99 + 0.161 = 100.151
```

• As normalized numbers, the operands would be written as:

 9.999×10^{1} 1.610×10^{-1}

1. Equalize the exponents.

The operand with the smaller exponent should be rewritten by increasing its exponent and shifting the point leftwards.

```
1.610 \times 10^{-1} = 0.0161 \times 10^{1}
```

With four significant digits, this gets rounded to 0.016×10^{1} .

This can result in a loss of least significant digits—the rightmost 1 in this case. But rewriting the number with the larger exponent could result in loss of the *most* significant digits, which is much worse.

2. Add the mantissas.

| | 9.999 | × | 10 ¹ | |
|---|--------|---|-----------------|--|
| + | 0.016 | × | 10 ¹ | |
| | 10.015 | × | 10 ¹ | |

3. Normalize the result if necessary.

```
10.015 \times 10^1 = 1.0015 \times 10^2
```

This step may cause the point to shift either left or right, and the exponent to either increase or decrease.

4. Round the number if needed.

 1.0015×10^2 gets rounded to 1.002×10^2 .

Repeat Step 3 if the result is no longer normalized.
 We don't need this in our example, but it's possible for rounding to add digits—for example, rounding 9.9995 yields 10.000.

Our result is 1.002×10^2 , or 100.2. The correct answer is 100.151, so we have the right answer to four significant digits, but there's a small error already.

- As we saw, rounding errors in addition can occur if one argument is much smaller than the other, since we need to match the exponents.
- An extreme example with 32-bit IEEE values is the following.

 $(1.5 \times 10^{38}) + (1.0 \times 10^{0}) = 1.5 \times 10^{38}$

The number $1.0 \times 10^{\circ}$ is much smaller than 1.5×10^{38} , and it basically gets rounded out of existence.

This has some nasty implications. The order in which you do additions can affect the result, so (x + y) + z is not always the same as x + (y + z)!

```
float x = -1.5e38;
float y = 1.5e38;
printf( "%f\n", (x + y) + 1.0 );
printf( "%f\n", x + (y + 1.0) );
```

Multiplication

 To multiply two floating-point values, first multiply their magnitudes and add their exponents.

 9.999×10^{1} $\times 1.610 \times 10^{-1}$ 16.098×10^{0}

- You can then round and normalize the result, yielding 1.610×10^{1} .
- The sign of the product is the exclusive-or of the signs of the operands.
 - If two numbers have the same sign, their product is positive.
 - If two numbers have different signs, the product is negative.

 $0 \oplus 0 = 0$ $0 \oplus 1 = 1$ $1 \oplus 0 = 1$ $1 \oplus 1 = 0$

This is one of the main advantages of using signed magnitude.

The history of floating-point computation

- In the past, each machine had its own implementation of floating-point arithmetic hardware and/or software.
 - It was impossible to write portable programs that would produce the same results on different systems.
 - Many strange tricks were needed to get correct answers out of some machines, such as Crays or the IBM System 370.
- It wasn't until 1985 that the IEEE 754 standard was adopted.
 - The standard is very complex and difficult to implement efficiently.
 - But having a standard at least ensures that all compliant machines will produce the same outputs for the same program.



Floating-point hardware

- Intel introduced the 8087 coprocessor around 1981.
 - The main CPU would call the 8087 for floating-point operations.
 - The 8087 had eight separate 80-bit floating-point registers that could be accessed in a stack-like fashion.
 - Some of the IEEE standard is based on the 8087.
- Intel's 80486, introduced in 1989, included floating-point support in the main processor itself.
- The MIPS floating-point architecture and instruction set still reflect the old coprocessor days, with separate floating-point registers and special instructions for accessing those registers.



MIPS floating-point architecture

- MIPS includes a separate set of 32 floating-point registers, \$f0-\$f31.
 - Each register is 32 bits long and can hold a single-precision value.
 - Two registers can be combined to store a double-precision number.
 You can have up to 16 double-precision values in registers \$f0-\$f1, \$f2-\$f3, ..., \$f30-\$f31.
 - \$f0 is *not* hardwired to the value 0.0!
- There are also separate instructions for floating-point arithmetic. The operands *must* be floating-point registers, and not immediate values.

add.s \$f1, \$f2, \$f3 # Single-precision \$f1 = \$f2 + \$f3 add.d \$f2, \$f4, \$f6 # Double-precision \$f2 = \$f4 + \$f6

- There are other basic operations as you would expect.
 - sub.s and sub.d for subtraction
 - mul.s and mul.d for multiplication
 - div.s and div.d for division

Floating-point register transfers

- mov.s and mov.d copy data between floating-point registers.
- Use mtc1 and mfc1 to transfer data between the integer registers \$0-\$31 and the floating-point registers \$f0-\$f31.
 - These are "raw" data transfers that do not convert between integer and floating-point representations.
 - Be careful with the order of the operands in these instructions.

| mtc1 | \$t0, | \$f0 | # | \$f0 | = | \$t0 |
|------|-------|-------------|---|------|---|-------------|
| mfc1 | \$t0, | \$f0 | # | \$t0 | = | \$f0 |

 There are also special loads and stores for transferring data between the floating-point registers and memory. (The base address is still given in an integer register.)

```
lwc1 $f2, 0($a0) # $f2 = M[$a0]
swc1 $f4, 4($sp) # M[$sp+4] = $f4
```

• The "c1" in the instruction names stands for "coprocessor 1."

Floating-point comparisons

 We also need special instructions for comparing floating-point values, since slt and sltu only apply to signed and unsigned integers.

| c.le.s | \$f2, | \$f 4 |
|--------|-------|--------------|
| c.eq.s | \$f2, | \$f 4 |
| c.lt.s | \$f2, | \$f 4 |

- The comparison result is stored in a *special* coprocessor register.
- You can then branch based on whether this register contains 1 or 0.

| bc1t | Label | # | branch | if | true |
|------|-------|---|--------|----|-------|
| bc1f | Labe1 | # | branch | if | false |

Here is how you can branch to the label Exit if \$f2 = \$f4.

c.eq.s \$f2, \$f4 bc1t Exit

Floating-point functions

- There are conventions for passing data to and from functions.
 - Floating-point arguments are placed in \$f12-\$f15.
 - Floating-point return values go into \$f0-\$f1.
- We also split the register-saving chores, just like earlier.
 - \$f0-\$f19 are caller-saved.
 - \$f20-\$f31 are callee-saved.
- These are the same basic ideas as before because we still have the same problems to solve—now it's just with different registers.

Floating-point constants

- MIPS does not support immediate floating-point arithmetic instructions, so you must load constant values into a floating-point register first.
- One solution is to store floating-point constants in the data segment, and to load them with a l.s or l.d pseudo-instruction.

| alpha: | .data .float | 0.55555 | # 5.0 / 9.0 |
|--------|-----------------|-------------|------------------|
| | .text l.s | \$f6, alpha | # \$f6 = 0.55555 |

 Newer versions of SPIM also support the li.s and li.d pseudo-instructions, which make life much easier.

li.s \$f6, 0.55555 # \$f6 = 0.55555

 You can also cast integers to floating-point values using the MIPS type conversion instructions.



 Possible types for conversions are integers (w), single-precision (s) and double-precision (d) floating-point.

| li | \$t0, 3 | 32 | # | \$t0 = | 32 |
|---------|----------|------|---|--------|------|
| mtc1 | \$t0, \$ | \$f2 | # | f2 = | 32 |
| cvt.s.w | \$f4, 9 | \$f2 | # | \$f4 = | 32.0 |

A complete example

 Here is a slightly different version of the textbook example of converting single-precision temperatures from Fahrenheit to Celsius.

```
celsius = (fahrenheit - 32.0) × 5.0 / 9.0
```

| celsius: li mtc1 cvt.s.w li.s sub.s mul.s | \$t0, \$t0, \$f4, \$f6, \$f0, \$f0, | 32 \$f4 \$f4 0.55555 \$f12, \$f4 \$f0, \$f6 | #### | <pre>\$f4 = \$f6 = \$f0 = \$f0 = \$f0 =</pre> | 32.0 5.0 / 9.0 \$f12 - 32.0 \$f0 * 5.0/9.0 |
|---|--|--|------|---|---|
| mul.s jr | \$f0, \$ra | \$f0, \$f6 | # | \$f0 = | \$f0 * 5.0/9.0 |

- This example demonstrates a couple of things.
 - The argument is passed in 12, and the return value is placed in 10.
 - We use two different ways of loading floating-point constants.
 - We used only caller-saved floating-point registers.

Summary

- The IEEE 754 standard defines number representations and operations for floating-point arithmetic.
- Having a finite number of bits means we can't represent all possible real numbers, and errors will occur from approximations.
- MIPS processors implement the IEEE 754 standard.
 - There is a separate set of floating-point registers, $\frac{1}{10}$
 - New instructions handle basic floating-point operations, comparisons and branches. There is also support for transferring data between the floating-point registers, main memory and the integer registers.
 - We still have to deal with issues of argument and result passing, and register saving and restoring in function calls.