Floating Point

- Representation for non-integral numbers
 - Including very small and very large numbers

$$4,600,000,000$$
 or 4.6×10^9

- Like scientific notation

$$-2 \times 10^{-7}$$

$$- +0.002 \times 10^{-4}$$

Can't be represented by integer

Normalized (no leading zeros)

not normalized

- In binary
 - $-\pm 1.xxxxxxx_2 \times 2^{yyyy}$
- Types float and double in C

float a; // single precision double b; //double precision

Floating Point Standard- IEEE Std 754-1985

• Single precision - 32-bit

single: 8 bits single: 23 bits

S	Exponent	Fraction	
X =	$=(-1)^{S}\times(1)^{S}$	+Fraction)×2 ^(Exponent-Bias)	Sign

 $x = (-1)^{S} \times (Significand) \times 2^{(Exponent-Bias)}$

Significand= 1+fraction

- S: sign bit $(0 \Rightarrow \text{non-negative}, 1 \Rightarrow \text{negative})$
- Normalized number ±1.xxxxxxx₂ × 2^{yyyy}
 - Always has a leading 1, so no need to represent it explicitly (hidden bit)
- Exponent: actual exponent + Bias
 - Ensures exponent is unsigned
 - Single precision: Bias = 127, Double precision: Bias = 1023

Floating-Point Example – single-precision

 What number is represented by the following singleprecision float?

```
x=11000000101000...00_2 (32-bit)
```

- S = 1
- Fraction = $01000...00_2$
- Exponent = $10000001_2 = 129$

•
$$x = (-1)^1 \times (1 + .01_2) \times 2^{(129 - 127)}$$

= $(-1) \times (1 + 1/4) \times 2^2$
= -5.0

Floating-Point Example

 Represent –0.75 in single-precision floating point

•
$$-0.75 = -(1/4+1/2)=(-1)^1 \times 1.1_2 \times 2^{-1}$$

- S = 1
- Fraction = 1000...00 Hidden 1 is not represented
- Exponent = -1 + Bias=126=011111110₂

Answer: 1011111101000...00

Why uses bias in the exponents

- Easy to compare which number is larger
 - Just need to check the bit from left to right

8 bit	S	Bi	as=127	
127 126	01111111 01111110	254 253	11111110 11111101	
1	00000001	128	10000000	
0	00000000	127	01111111	
-1	111111111			
-126	10000010	1	00000001	
-127	10000001	0	00000000	reserved (discussed later)
-128	10000000	255	11111111	reserved (discussed later)

- Allow quick comparison of floating point numbers
 - from MSB to LSB (except the sign bit)

Floating Point Standard- IEEE Std 754-1985

Double precision (64-bit)

double: 11 bits double: 52 bits

S	Exponent	Fraction
---	----------	----------

$$x = (-1)^{S} \times (1 + Fraction) \times 2^{(Exponent-Bias)}$$

$$x = (-1)^{S} \times (Significand) \times 2^{(Exponent-Bias)}$$

- S: sign bit $(0 \Rightarrow \text{non-negative}, 1 \Rightarrow \text{negative})$
- Normalized number

$$\pm 1.xxxxxxx_2 \times 2^{yyyy}$$

- Have hidden 1
- Exponent: excess representation: actual exponent + Bias
 - Ensures exponent is unsigned
 - Double: Bias = 1023

Floating-Point Example – double-precision

 What number is represented by the following double float?

- S = 1
- Fraction = $1000...00_2$

•
$$x = (-1)^1 \times (1 + .1_2) \times 2^{(1021 - 1023)}$$

= $(-1) \times (1+1/2) \times 2^{-2}$
= $-3/8$

Floating-Point Example

 Represent –0.75 in double-precision floating point

•
$$-0.75 = (-1)^1 \times 1.1_2 \times 2^{-1}$$

- S = 1
- Fraction = 1000...00₂ Hidden 1 is not represented
- Exponent = -1 + Bias= -1+1023= $1022=011111111110_2$

Ans: 10111111111101000...00

IEEE 754 Encoding of FP number

Encoding

- Exp. 00...00 and 111...11 reserved
- Exp.=00000000 and Fract.=00000...00 => 0
- Exp.=0, and Fract. != 0 => denormalized number (discuss later)
- Exp.=111..111 and Fract.= 000...000 => $\pm \infty$ (discuss later)
- Exp.=111...111 and Fract.!=0 => Non a Number (NaN) (discuss later)

Single precision		Double precision		Object r	epresented
Exponent	Fraction	Exponent	Fraction		
0	0	0	0		0
0	Nonzero	0	Nonzero	± denorm	alized number
1–254	Anything	1–2046	Anything	± floating	-point number
255	0	2047	0	±	infinity
255	Nonzero	2047	Nonzero	NaN (No	ot a Number)

Single-Precision Range (for EXP 1 to 254)

Smallest value

- **–** 000...01 00000000
- Fraction: $000...00 \Rightarrow$ significand = 1.0
- Exponent = 1 127 = -126
- Smallest value = $1.0 \times 2^{-126} \approx 1.2 \times 10^{-38}$
- Largest value
 - **-** 111...1101111111
 - Fraction: 111...11 \Rightarrow significand ≈ 2.0
 - Exponent = 254 127 = +127
 - − Largest value $\approx 2.0 \times 2^{+127} \approx 3.4 \times 10^{+38}$

S	Exponent	Fraction
---	----------	----------

Exponents
00000000 and
11111111 reserved

Double-Precision Range (for EXP 1 to 2046)

Smallest value

- 00000000010000....000
- Exponent = 1 1023 = -1022
- -1.0×2^{-1022}
- Largest value
 - **—** 11111111110111....11111
 - Exponent = 2046 1023 = +1023
 - Fraction: 111...11 ⇒ significand ≈ 2.0
 - $-2.0 \times 2^{+1023}$

S	Exponent	Fraction
---	----------	----------

Exponents 00000...00

and 11111...111 reserved

Denormalized Numbers

- (Review) Smallest normalized value
 - -00000010000000.....0000
 - Fraction: $000...00 \Rightarrow$ significand = 1.0
 - Exponent = 1 127 = -126
 - Smallest value = 1.0×2^{-126}
- How to represent number smaller than 1.0x2⁻¹²⁶?
- E.g. 0.5x2⁻¹²⁶ =>Use denormalized number

S	Exponent	Fraction
---	----------	----------

Denormalized Numbers (32-bit)

- Exponent = 00000000
- Fraction ⇒ hidden bit is 0

$$x = (-1)^{S} \times (Fraction) \times 2^{-126}$$

- Allow for gradual underflow
- Denormalized with fraction = 000...0 → 0

$$x = (-1)^{S} \times (0+0) \times 2^{-126} = \pm 0.0$$

Two representations of 0.0! +0.0 and -0.0

Example

Smallest positive single precision normalized number

```
1.00000000...00000_2x2^{-126}
```

• Smallest positive single precision denormalized no.

```
(Hint: Fraction is 23-bit)
S Exp Fraction
```

0 0000 0000 0000 0000 0000 0000 001

 2^{-126} $\times 1/2^{23}$

Special number: Infinities and NaNs

- Exponent = 111...1, Fraction = 000...0
 - $-\pm\infty$
 - Can be used in subsequent calculations, avoiding need for overflow check
 - E.g. $F+(+\infty)=+\infty$, or $F/\infty=0$
- Exponent = 111...1, Fraction ≠ 000...0
 - Not-a-Number (NaN)
 - Indicates illegal or undefined result
 - e.g., xx / 0.0
 - Can be used in subsequent calculations

Floating-Point Addition

- Consider a 4-digit decimal example
 - $-9.999 \times 10^{1} + 1.610 \times 10^{-1}$
- 1. Align decimal points
 - Shift number with smaller exponent $9.999 \times 10^1 + 0.016 \times 10^1$
- 2. Add significands $9.999 \times 10^1 + 0.016 \times 10^1 = 10.015 \times 10^1$
- 3. Normalize result & check for over/underflow 1.0015×10^2
- 4. Round and renormalize if necessary 1.002×10^2
 - Renormalize may be required due to rounding
 - E.g. rounding 9.9999 x 10²

Floating-Point Addition

Why do we shift the number with a smaller exponent?

- Example
 - $-9.99999 \times 10^{40} + 1.610 \times 10^{-1}$
 - Shift 9.99999 x 10⁴⁰ to 10⁻¹→ significand may be too large → overflow
 - 1.610x10⁻¹ to 10⁴⁰ => significand may be very small, but still don't overflow, using round
 - Therefore, we shift the smaller number

Floating-Point Addition

Now consider a 4-digit binary example

$$-1.000_2 \times 2^{-1} + -1.110_2 \times 2^{-2} \quad (0.5 + -0.4375)$$

1. Align binary points

Shift number with smaller exponent

$$1.000_2 \times 2^{-1} + -0.111_2 \times 2^{-1}$$

2. Add significands

$$1.000_2 \times 2^{-1} + -0.111_2 \times 2^{-1} = 0.001_2 \times 2^{-1}$$

3. Normalize result & check for over/underflow

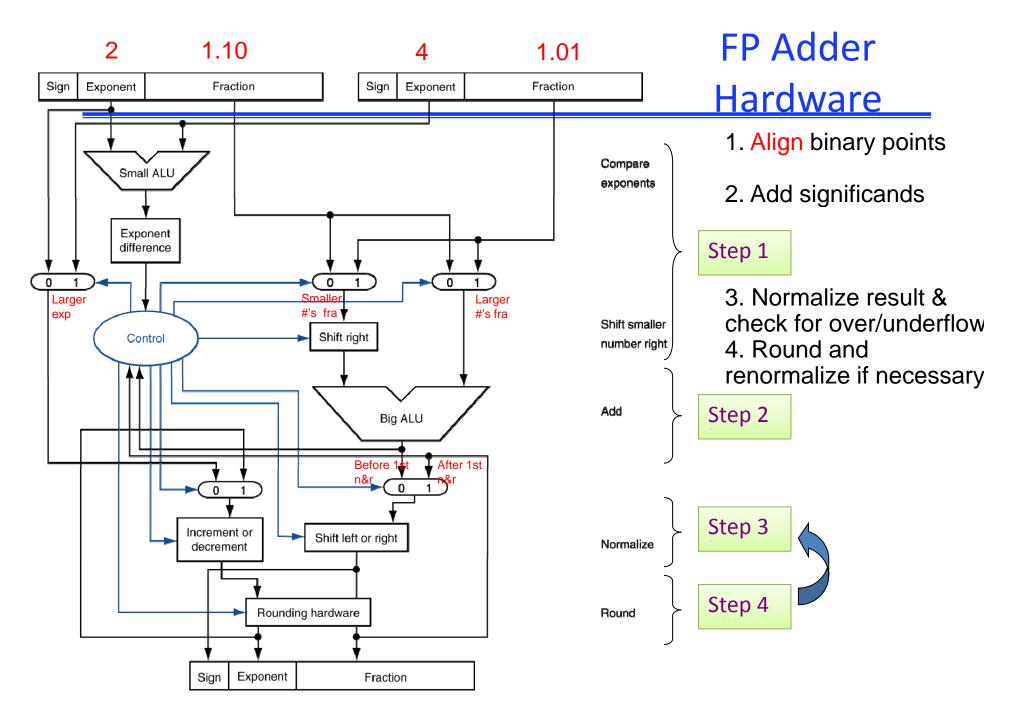
$$1.000_2 \times 2^{-4}$$
, with no over/underflow

4. Round and renormalize if necessary

$$1.000_2 \times 2^{-4}$$
 (no change) = 0.0625

FP Adder Hardware

- Much more complex than integer adder
 - Steps includes shift exponents and fraction, add fraction, ..., etc.
- Doing it in one clock cycle would take too long
 - Much longer than integer operations
 - Slower clock would penalize all instructions
- FP adder usually takes several cycles
 - Can be pipelined (see Chapter 4 about pipeline)



Floating-Point Multiplication

- Consider a 4-digit decimal example
 - $-1.110 \times 10^{10} \times 9.200 \times 10^{-5}$
- 1. Add exponents
 - For biased exponents, subtract bias from sum (described later)
 - New exponent = 10 + -5 = 5
- 2. Multiply significands
 - $-1.110 \times 9.200 = 10.212 \implies 10.212 \times 10^{5}$
- 3. Normalize result & check for over/underflow
 - -1.0212×10^6
- 4. Round and renormalize if necessary
 - -1.021×10^6
- 5. Determine sign of result from signs of operands
 - $+1.021 \times 10^6$

Floating-Point Multiplication

- Now consider a 4-digit binary example
 - $-1.000_2 \times 2^{-1} \times -1.110_2 \times 2^{-2} (0.5 \times -0.4375)$
- 1. Add exponents
 - Unbiased: -1 + -2 = -3
 - Biased: (-1 + 127) + (-2 + 127) 127 = -3 + 254 127
- 2. Multiply significands

$$-1.000_2 \times 1.110_2 = 1.110_2 \implies 1.110_2 \times 2^{-3}$$

- 3. Normalize result & check for over/underflow
 - $-1.110_2 \times 2^{-3}$ (no change) with no over/underflow
- 4. Round and renormalize if necessary
 - $-1.110_2 \times 2^{-3}$ (no change)
- 5. Determine sign: $+ \times \Rightarrow -$

$$-1.110_2 \times 2^{-3} = -0.21875$$

Remove one bias

FP Instructions in MIPS

- FP hardware is coprocessor 1
 - Adjunct processor that extends the ISA
- Separate FP registers for single precision
 - 32 single-precision: \$f0, \$f1, ... \$f31
 - These registers are also used for double-precision computation (described later)
- FP instructions operate only on FP registers
- Single-precision FP load and store instructions
 - lwc1, swc1

- e.g., lwc1 \$f8, 32(\$sp)
- Single-precision arithmetic
- e.g., add.s \$f0, \$f1, \$f6

- add.s, sub.s, mul.s, div.s
- Single-precision comparison

- e.g. c.lt.s \$f3, \$f4
- c.xx.s (xx is eq, neq, lt, le, ...)
- Sets or clears FP condition-code bit

Branch on FP condition code true/false bc1t, bc1f e.g., bc1t TargetLabel

FP Example: °F to °C

C code:

```
float f2c (float F) {
  return ((5.0/9.0)*(F - 32.0));
}
```

- Assume \$gp can access three constant 5. 0, 9.0, and 32.0
- Assume F in \$f12, and put result in \$f0
- Compiled MIPS code:

```
lwc1 $f16, const5($gp)
lwc1 $f18, const9($gp)
div.s $f16, $f16, $f18  # f16 = 5.0/9.0
lwc1 $f18, const32($gp)
sub.s $f18, $f12, $f18  # f18 = F - 32.0
mul.s $f0, $f16, $f18  # f0 = (5/9)x (F-32)
jr $ra  • $f0 and $f1 are used for FP results, similar to
$v0 and $v1 for INT results
```

FP Instructions in MIPS for double-precision

- Use 32 FP registers \$f0 \$f31
 - Paired for double-precision: \$f0/\$f1, \$f2/\$f3, ...
- FP Double-precision load and store instructions
 - ldc1, sdc1
- Double-precision arithmetic
 mul.d \$f4, \$f4, \$f6
 - add.d, sub.d, mul.d, div.d
- Double-precision comparison c.lt.d \$f4, \$f6
 - c.xx.d (xx is eq, lt, le, ...)
 - Sets or clears FP condition-code bit
- Branch on FP condition code true or false
 - bc1t, bc1f
 - e.g., bc1t TargetLabel

FP Example: Array Multiplication

- X = X + Y × Z
 All 32 × 32 matrices, 64-bit double-precision elements
- C code:

FP Example: Array Multiplication

MIPS code:

```
Ιi
        $t1, 32  # $t1 = 32 (row size/loop end)
      \$s0, 0 # i = 0; initialize 1st for loop
L1: Ii \$\$1, 0 # j = 0; restart 2nd for loop
L2: Ii \$s2, 0 # k = 0; restart 3rd for loop
   sll $t2, $s0, 5 # $t2 = i * 32 (size of row of x)
   addu t2, t2, t2, t2 = i * size(row) + j
   sll $t2, $t2, 3 # $t2 = byte offset of [i][j]
   addu t2, a0, t2 \# t2 = byte address of <math>x[i][j]
   I.d f_4, f_4, f_4 = 8 bytes of f_4 = 8 bytes of f_4 [i][j]
L3: sll $t0, $s2, 5 # <math>$t0 = k * 32 (size of row of z)
   addu $t0, $t0, $s1 # $t0 = k * size(row) + j
   \$11 \$t0, \$t0, 3 \#\$t0 = byte offset of [k][j]
   addu t0, a2, t0 # t0 = byte address of <math>z[k][j]
   I.d f16, 0(f0) # f16 = 8 bytes of z[k][j]
```

the same as Idc1, but a pseudo instruction

FP Example: Array Multiplication

```
$11 $t0, $s0, 5 # $t0 = i*32 (size of row of y)
addu $t0, $t0, $s2 # $t0 = i * size(row) + k
sll $t0, $t0, 3 # $t0 = byte offset of [i][k]
addu $t0, $a1, $t0  # $t0 = byte address of y[i][k]
I.d f18, 0($t0) # f18 = 8 bytes of y[i][k]
mul.d f16, f18, f16 # f16 = y[i][k] * z[k][j]
add. d f_4, f_4, f_6 # f_4=x[i][j] + y[i][k]*z[k][j]
addi u \$s2, \$s2, 1 # \$k = k + 1
bne $s2, $t1, L3 # if (k != 32) go to L3
s. d $f4, 0($t2)
                    \# x[i][j] = $f4
addiu \$\$1, \$\$1, 1 # \$j = j + 1
bne $s1, $t1, L2 # if (j != 32) go to L2
addiu $50, $50, 1 # $i = i + 1
                    # if (i != 32) go to L1
bne $s0, $t1, L1
```

Improve Accuracy: Guard and Round Bits

- IEEE Std 754 specifies additional rounding control
 - Extra bits of precision (guard, round, sticky)
 - Choice of rounding modes
 - Allows programmer to fine-tune numerical behavior of a computation
- Consider the addition $2.56 \times 10^{0} + 2.34 \times 10^{2} = 2.3656 \times 10^{2}$
- Without guard and round bits $0.02 \times 10^2 + 2.34 \times 10^2 = 2.36 \times 10^2$
- With guard and round bits

$$0.0256 \times 10^{2} + 2.3400 \times 10^{2} = 2.3656 \times 10^{2} = 2.37 \times 10^{2}$$

closer to accurate answer

Improve Accuracy: Rounding Modes

- 4 rounding modes
 - Round towards + ∞
 - Round towards ∞
 - Round towards0
 - truncate
 - Round to nearest (even)
 - Default
 - if the number falls midway, it is rounded to the nearest value with an even (zero) least significant bit
 - The sticky bit can improve the accuracy here (see next slide)

Improve Accuracy: Sticky Bit

- Sticky bit: one bit is set when there are nonzero bits to the right of the round bit.
 - Allow computer to see the difference between $0.50000..0_{10}$ and $0.50000..1_{10}$

- Without Sticky bit
 - 2.3450000000001 will be stored as 2.345
- With Sticky bit
 - 2.345000000001 will be stored as 2.345 and sticky bit =1
- Used for rounding

2.345 with sticky bit=1 is larger than 2.345

Interpretation of Data

The BIG Picture

- Bits have no inherent meaning
 - Interpretation depends on the instructions applied
- Computer representations of numbers
 - Finite range and precision
 - Need to account for this in programs

Associativity

Is (x+y)+z equal to x+(y+z) ???

		(x+y)+z	x+(y+z)
X	-1.50E+38		-1.50E+38
у	1.50E+38	0.00E+00	
Z	1.0	1.0	1.50E+38
		1.00E+00	0.00E+00

- Parallel Programs may interleave operations in unexpected orders
 - Assumptions of associativity may fail
- Parallel execution strategies that work for integers may not work for floating-point numbers
 - Need to validate parallel programs under varying degrees of parallelism

Concluding Remarks

- ISAs support arithmetic
 - Signed and unsigned integers
 - Floating-point approximation to reals
- Bounded range and precision
 - Operations can overflow and underflow