

# 15-213

## Floating Point Arithmetic August 25, 2007

### Topics

- IEEE Floating Point Standard
- Rounding
- Floating Point Operations
- Mathematical properties

# Floating Point Puzzles

■ For each of the following C expressions, either:

- Argue that it is true for all argument values
- Explain why not true

```
int x = ...;
float f = ...;
double d = ...;
```

Assume neither  
d nor f is NaN

- `x == (int)(float) x`
- `x == (int)(double) x`
- `f == (float)(double) f`
- `d == (float) d`
- `f == -(-f);`
- `2/3 == 2/3.0`
- `d < 0.0       ⇒     ((d*2) < 0.0)`
- `d > f         ⇒     -f > -d`
- `d * d >= 0.0`
- `(d+f) - d == f`

# IEEE Floating Point

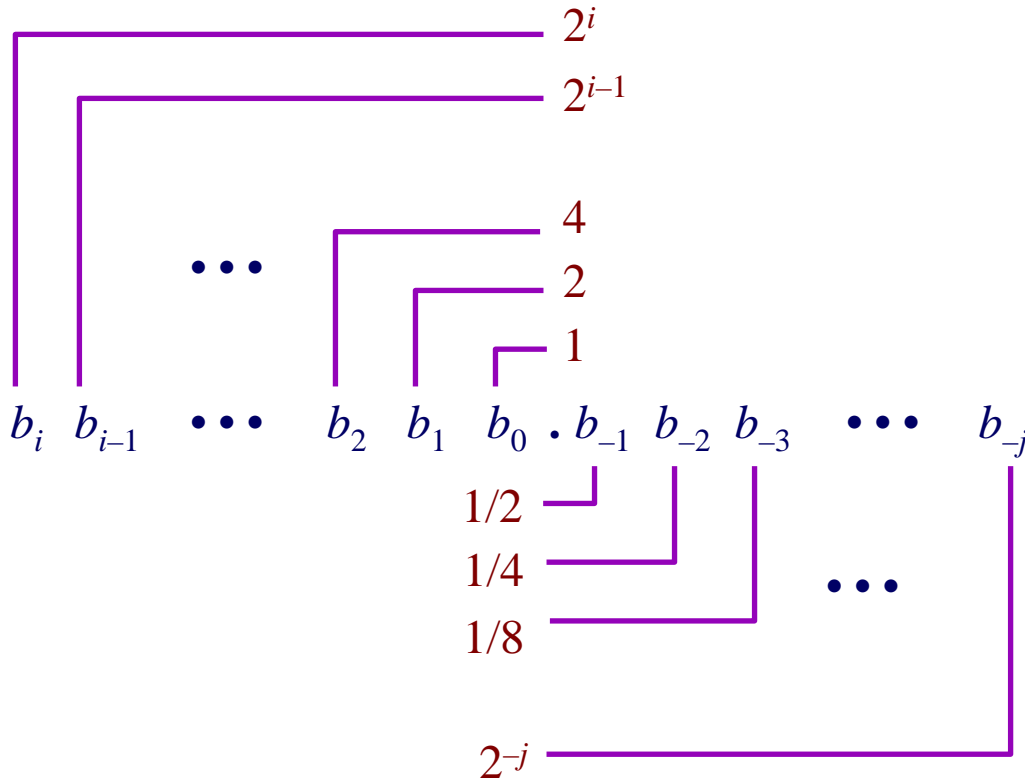
## IEEE Standard 754

- Established in 1985 as uniform standard for floating point arithmetic
  - Before that, many idiosyncratic formats
- Supported by all major CPUs

## Driven by Numerical Concerns

- Nice standards for rounding, overflow, underflow
- Hard to make go fast
  - Numerical analysts predominated over hardware types in defining standard

# Fractional Binary Numbers



## Representation

- Bits to right of “binary point” represent fractional powers of 2
- Represents rational number:

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

# Frac. Binary Number Examples

Value	Representation
5-3/4	$101.11_2$
2-7/8	$10.111_2$
63/64	$0.111111_2$

## Observations

- Divide by 2 by shifting right
- Multiply by 2 by shifting left
- Numbers of form  $0.111111\dots_2$  just below 1.0
  - $1/2 + 1/4 + 1/8 + \dots + 1/2^i + \dots \rightarrow 1.0$
  - Use notation  $1.0 - \epsilon$

# Representable Numbers

## Limitation

- Can only exactly represent numbers of the form  $x/2^k$
- Other numbers have repeating bit representations

## Value

## Representation

1/3

0.0101010101 [01]...<sub>2</sub>

1/5

0.001100110011 [0011]...<sub>2</sub>

1/10

0.0001100110011 [0011]...<sub>2</sub>

# 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 is sign bit
- exp field encodes  $E$
- frac field encodes  $M$

# Floating Point Precisions

## Encoding



- MSB is sign bit
- exp field encodes  $E$
- frac field encodes  $M$

## Sizes

- Single precision: 8 exp bits, 23 frac bits
  - 32 bits total
- Double precision: 11 exp bits, 52 frac bits
  - 64 bits total
- Extended precision: 15 exp bits, 63 frac bits
  - Only found in Intel-compatible machines
  - Stored in 80 bits
    - » 1 bit wasted



# “Normalized” Numeric Values

## Condition

- $\text{exp} \neq 000\dots 0$  and  $\text{exp} \neq 111\dots 1$

## Exponent coded as *biased* value

$$E = \text{Exp} - \text{Bias}$$

- *Exp* : unsigned value denoted by *exp*
- *Bias* : Bias value
  - » Single precision: 127 (*Exp*: 1...254, *E*: -126...127)
  - » Double precision: 1023 (*Exp*: 1...2046, *E*: -1022...1023)
  - » in general:  $\text{Bias} = 2^{e-1} - 1$ , where *e* is number of exponent bits

## Significand coded with implied leading 1

$$M = 1.\text{xxx}\dots\text{x}_2$$

- *xxx...x*: bits of *frac*
- Minimum when 000...0 ( $M = 1.0$ )
- Maximum when 111...1 ( $M = 2.0 - \epsilon$ )
- Get extra leading bit for “free”

# Normalized Encoding Example

## Value

Float  $F = 15213.0;$

■  $15213_{10} = 11101101101101_2 = 1.1101101101101_2 \times 2^{13}$

## Significand

$M = 1.\underline{1101101101101}_2$

$\text{frac} = \underline{11011011011010000000000}_2$

## Exponent

$E = 13$

$\text{Bias} = 127$

$\text{Exp} = 140 = 10001100_2$

### Floating Point Representation:

Hex:	4	6	6	D	B	4	0	0
Binary:	0100	0110	0110	1101	1011	0100	0000	0000
140:	100	0110	0					
15213:				1110	1101	1011	01	

# Denormalized Values

## Condition

- $\text{exp} = 000\dots 0$

## Value

- Exponent value  $E = -\text{Bias} + 1$
- Significand value  $M = 0.\text{xxx}\dots\text{x}_2$ 
  - $\text{xxx}\dots\text{x}$ : bits of *frac*

## Cases

- $\text{exp} = 000\dots 0, \text{frac} = 000\dots 0$ 
  - Represents value 0
  - Note that have distinct values +0 and -0
- $\text{exp} = 000\dots 0, \text{frac} \neq 000\dots 0$ 
  - Numbers very close to 0.0
  - Lose precision as get smaller
  - “Gradual underflow”

# Special Values

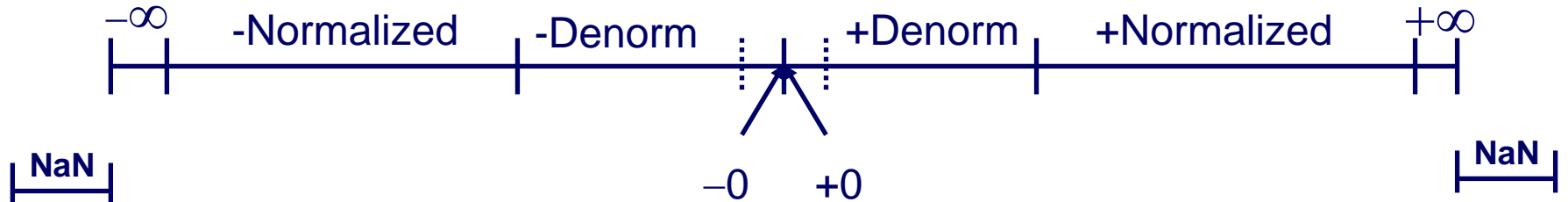
## Condition

- $\text{exp} = 111\dots 1$

## Cases

- $\text{exp} = 111\dots 1, \text{frac} = 000\dots 0$ 
  - Represents value  $\infty$  (infinity)
  - Operation that overflows
  - Both positive and negative
  - E.g.,  $1.0/0.0 = -1.0/-0.0 = +\infty$ ,  $1.0/-0.0 = -\infty$
- $\text{exp} = 111\dots 1, \text{frac} \neq 000\dots 0$ 
  - Not-a-Number (NaN)
  - Represents case when no numeric value can be determined
  - E.g.,  $\text{sqrt}(-1)$ ,  $\infty - \infty$ ,  $\infty * 0$

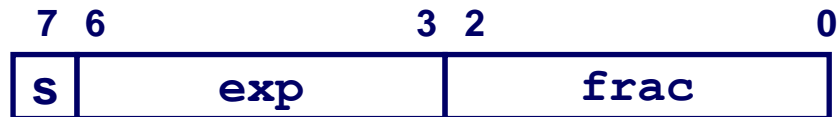
# Summary of Floating Point Real Number Encodings



# 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



# Values Related to the Exponent

Exp	exp	E	$2^E$	
0	0000	-6	1/64	(denorms)
1	0001	-6	1/64	
2	0010	-5	1/32	
3	0011	-4	1/16	
4	0100	-3	1/8	
5	0101	-2	1/4	
6	0110	-1	1/2	
7	0111	0	1	
8	1000	+1	2	
9	1001	+2	4	
10	1010	+3	8	
11	1011	+4	16	
12	1100	+5	32	
13	1101	+6	64	
14	1110	+7	128	
15	1111	n/a		(inf, NaN)

# Dynamic Range

	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	
	.....						
	0	0001	000	-6	$8/8 * 1/64 = 8/512$	← smallest norm	
	0	0001	001	-6	$9/8 * 1/64 = 9/512$		
Normalized numbers	...						
	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			

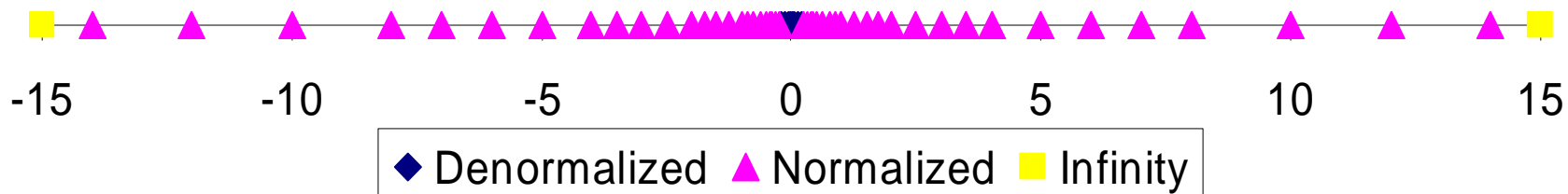


# Distribution of Values

## 6-bit IEEE-like format

- $e = 3$  exponent bits
- $f = 2$  fraction bits
- Bias is 3

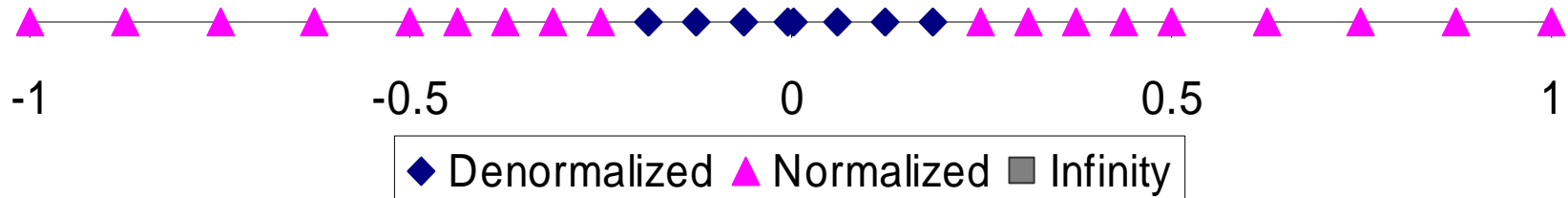
Notice how the distribution gets denser toward zero.



# Distribution of Values (close-up view)

## 6-bit IEEE-like format

- $e = 3$  exponent bits
- $f = 2$  fraction bits
- Bias is 3



# Interesting Numbers

Description	exp	frac	Numeric Value
Zero	00...00	00...00	0.0
Smallest Pos. Denorm.	00...00	00...01	$2^{-\{23,52\}} \times 2^{-\{126,1022\}}$
<ul style="list-style-type: none"> <li>■ Single <math>\approx 1.4 \times 10^{-45}</math></li> <li>■ Double <math>\approx 4.9 \times 10^{-324}</math></li> </ul>			
Largest Denormalized	00...00	11...11	$(1.0 - \epsilon) \times 2^{-\{126,1022\}}$
<ul style="list-style-type: none"> <li>■ Single <math>\approx 1.18 \times 10^{-38}</math></li> <li>■ Double <math>\approx 2.2 \times 10^{-308}</math></li> </ul>			
Smallest Pos. Normalized	00...01	00...00	$1.0 \times 2^{-\{126,1022\}}$
<ul style="list-style-type: none"> <li>■ Just larger than largest denormalized</li> </ul>			
One	01...11	00...00	1.0
Largest Normalized	11...10	11...11	$(2.0 - \epsilon) \times 2^{\{127,1023\}}$
<ul style="list-style-type: none"> <li>■ Single <math>\approx 3.4 \times 10^{38}</math></li> <li>■ Double <math>\approx 1.8 \times 10^{308}</math></li> </ul>			

# Special Properties of Encoding

## FP Zero Same as Integer Zero

- All bits = 0

## Can (Almost) Use Unsigned Integer Comparison

- Must first compare sign bits
- Must consider  $-0 = 0$
- NaNs problematic
  - Will be greater than any other values
  - What should comparison yield?
- Otherwise OK
  - Denorm vs. normalized
  - Normalized vs. infinity

# Floating Point Operations

## Conceptual View

- First compute exact result
- Make it fit into desired precision
  - Possibly overflow if exponent too large
  - Possibly round to fit into `frac`

## Rounding Modes (illustrate with \$ rounding)

	\$1.40	\$1.60	\$1.50	\$2.50	-\$1.50
■ Zero	\$1	\$1	\$1	\$2	-\$1
■ Round down ( $-\infty$ )	\$1	\$1	\$1	\$2	-\$2
■ Round up ( $+\infty$ )	\$2	\$2	\$2	\$3	-\$1
■ Nearest Even (default)	\$1	\$2	\$2	\$2	-\$2

### Note:

1. Round down: rounded result is close to but no greater than true result.
2. Round up: rounded result is close to but no less than true result.

# Closer Look at Round-To-Even

## Default Rounding Mode

- Hard to get any other kind without dropping into assembly
- All others are statistically biased
  - Sum of set of positive numbers will consistently be over- or underestimated

## Applying to Other Decimal Places / Bit Positions

- When exactly halfway between two possible values
  - Round so that least significant digit is even
- E.g., round to nearest hundredth

1.2349999	1.23	(Less than half way)
1.2350001	1.24	(Greater than half way)
1.2350000	1.24	(Half way—round up)
1.2450000	1.24	(Half way—round down)

# Rounding Binary Numbers

## Binary Fractional Numbers

- “Even” when least significant bit is 0
- Half way when bits to right of rounding position =  $100\dots_2$

## Examples

- Round to nearest  $1/4$  (2 bits right of binary point)

Value	Binary	Rounded	Action	Rounded Value
$2 \frac{3}{32}$	$10.00011_2$	$10.00_2$	( $<1/2$ —down)	2
$2 \frac{3}{16}$	$10.00110_2$	$10.01_2$	( $>1/2$ —up)	$2 \frac{1}{4}$
$2 \frac{7}{8}$	$10.11100_2$	$11.00_2$	( $1/2$ —up)	3
$2 \frac{5}{8}$	$10.10100_2$	$10.10_2$	( $1/2$ —down)	$2 \frac{1}{2}$

# FP Multiplication

## Operands

$$(-1)^{s1} M1 2^{E1} \quad * \quad (-1)^{s2} M2 2^{E2}$$

## Exact Result

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

- Sign  $s$ :  $s1 \wedge s2$
- Significand  $M$ :  $M1 * M2$
- Exponent  $E$ :  $E1 + E2$

## Fixing

- If  $M \geq 2$ , shift  $M$  right, increment  $E$
- If  $E$  out of range, overflow
- Round  $M$  to fit `frac` precision

## Implementation

- Biggest chore is multiplying significands



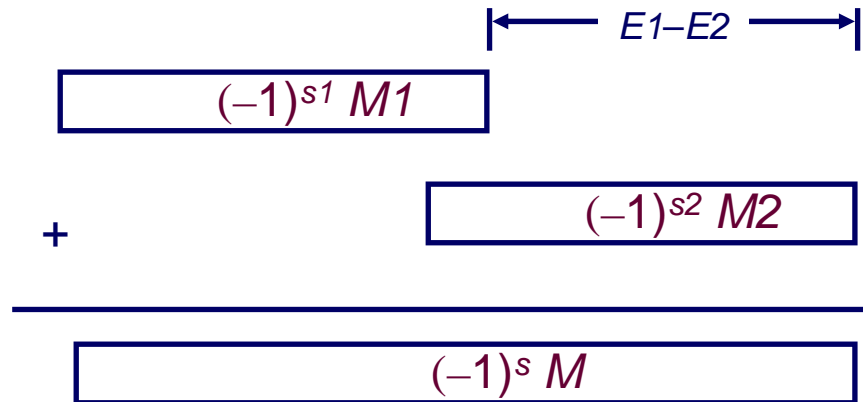
# FP Addition

## Operands

$$(-1)^{s1} M1 2^{E1}$$

$$(-1)^{s2} M2 2^{E2}$$

- Assume  $E1 > E2$



## Exact Result

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

- Sign  $s$ , significand  $M$ :
  - Result of signed align & add
- Exponent  $E$ :  $E1$

## 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  $\text{frac}$  precision

# Mathematical Properties of FP Add

## Compare to those of Abelian Group

- Closed under addition? YES
  - But may generate infinity or NaN
- Commutative? YES
- Associative? NO
  - Overflow and inexactness of rounding
- 0 is additive identity? YES
- Every element has additive inverse ALMOST
  - Except for infinities & NaNs

## Monotonicity

- $a \geq b \Rightarrow a+c \geq b+c$  ALMOST
  - Except for infinities & NaNs

# Math. Properties of FP Mult

## Compare to Commutative Ring

- Closed under multiplication? YES
  - But may generate infinity or NaN
- Multiplication Commutative? YES
- Multiplication is Associative? NO
  - Possibility of overflow, inexactness of rounding
- 1 is multiplicative identity? YES
- Multiplication distributes over addition? NO
  - Possibility of overflow, inexactness of rounding

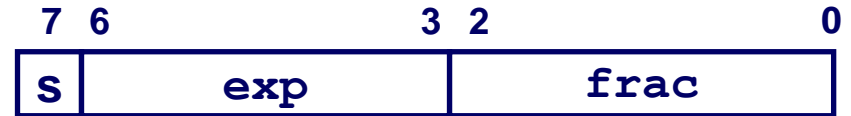
## Monotonicity

- $a \geq b \ \& \ c \geq 0 \Rightarrow a * c \geq b * c?$  ALMOST
  - Except for infinities & NaNs

# Creating Floating Point Number

## Steps

- Normalize to have leading 1
- Round to fit within fraction
- Postnormalize to deal with effects of rounding



## Case Study

- Convert 8-bit unsigned numbers to tiny floating point format
- Example Numbers

128      10000000

15        00001101

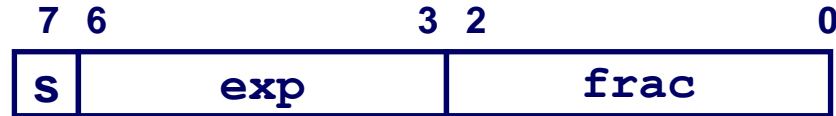
33        00010001

35        00010011

138      10001010

63        00111111

# Normalize

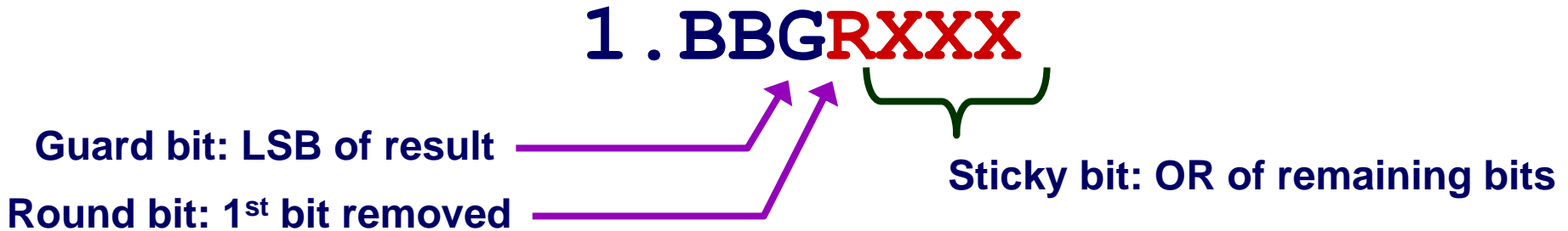


## Requirement

- Set binary point so that numbers of form 1.xxxxx
- Adjust all to have leading one
  - Decrement exponent as shift left

Value	Binary	Fraction	Exponent
128	10000000	1.0000000	7
15	00001101	1.1010000	3
17	00010001	1.0001000	5
19	00010011	1.0011000	5
138	10001010	1.0001010	7
63	00111111	1.1111100	5

# Rounding



## Round up conditions

- Round = 1, Sticky = 1 → > 0.5
- Guard = 1, Round = 1, Sticky = 0 → Round to even

Value	Fraction	GRS	Incr?	Rounded
128	1.000 <b>0000</b>	000	N	1.000
15	1.101 <b>0000</b>	100	N	1.101
17	1.000 <b>1000</b>	010	N	1.000
19	1.001 <b>1000</b>	110	Y	1.010
138	1.000 <b>1010</b>	111	Y	1.001
63	1.111 <b>1100</b>	111	Y	10.000

# Postnormalize

## Issue

- Rounding may have caused overflow
- Handle by shifting right once & incrementing exponent

Value	Rounded	Exp	Adjusted	Result
128	1.000	7		128
15	1.101	3		15
17	1.000	4		16
19	1.010	4		20
138	1.001	7		134
63	10.000	5	1.000/6	64

# Floating Point in C

## C Guarantees Two Levels

`float`      single precision  
`double`     double precision

## Conversions

- Casting between `int`, `float`, and `double` changes numeric values
- Double or `float` to `int`
  - Truncates fractional part
  - Like rounding toward zero
  - Not defined when out of range or NaN
    - » Generally sets to TMin
- `int` to `double`
  - Exact conversion, as long as `int` has  $\leq 53$  bit word size
- `int` to `float`
  - Will round according to rounding mode



# Curious Excel Behavior

	Number	Subtract 16	Subtract .3	Subtract .01
Default Format	16.31	0.31	0.01	-1.2681E-15
Currency Format	\$16.31	\$0.31	\$0.01	(\$0.00)

- Spreadsheets use floating point for all computations
- Some imprecision for decimal arithmetic
- Can yield nonintuitive results to an accountant!

# Summary

## IEEE Floating Point Has Clear Mathematical Properties

- Represents numbers of form  $M \times 2^E$
- Can reason about operations independent of implementation
  - As if computed with perfect precision and then rounded
- Not the same as real arithmetic
  - Violates associativity/distributivity
  - Makes life difficult for compilers & serious numerical applications programmers