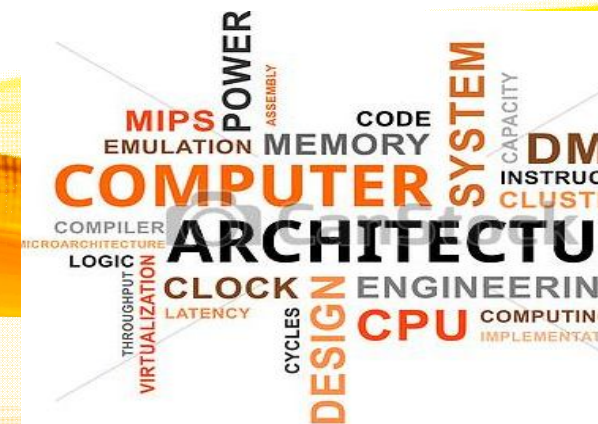


# UNIT II

## ARITHMETIC OPERATIONS

Addition and subtraction of signed numbers – Design of fast adders -  
Multiplication of positive numbers - Signed operand multiplication- fast  
multiplication – Integer division – **Floating point numbers and operations**



# Recap the previous Class



# Representing Fractional Numbers

- A binary number with fractional part

$$B = b_{n-1} b_{n-2} \dots b_1 b_0 . b_{-1} b_{-2} \dots b_{-m}$$

corresponds to the decimal number

$$D = \sum_{i=-m}^{n-1} b_i 2^i$$

*If the radix point is allowed to move, we call it a floating-point representation.*

- Also called *fixed-point numbers*.
  - The position of the radix point is fixed.

# Some Examples

$$1011.1 \rightarrow 1x2^3 + 0x2^2 + 1x2^1 + 1x2^0 + 1x2^{-1} = 11.5$$

$$101.11 \rightarrow 1x2^2 + 0x2^1 + 1x2^0 + 1x2^{-1} + 1x2^{-2} = 5.75$$

$$10.111 \rightarrow 1x2^1 + 0x2^0 + 1x2^{-1} + 1x2^{-2} + 1x2^{-3} = 2.875$$

## Some Observations:

- Shift right by 1 bit means divide by 2
- Shift left by 1 bit means multiply by 2
- Numbers of the form  $0.111111\dots_2$  has a value less than 1.0 (one).

# Limitations of Representation

- In the fractional part, we can only represent numbers of the form  $x/2^k$  exactly.
  - Other numbers have repeating bit representations (i.e. never converge).

- Examples:

$$3/4 = 0.11$$

$$7/8 = 0.111$$

$$5/8 = 0.101$$

$$1/3 = 0.101010101 [01] \dots$$

$$1/5 = 0.001100110011 [0011] \dots$$

$$1/10 = 0.0001100110011 [0011] \dots$$

- More the number of bits, more accurate is the representation.
- We sometimes see:  $(1/3)^*3 \neq 1$ .

# Floating-Point Number Representation (IEEE-754)

- For representing numbers with fractional parts, we can assume that the fractional point is somewhere in between the number (say,  $n$  bits in integer part,  $m$  bits in fraction part). → *Fixed-point representation*
  - Lacks flexibility.
  - Cannot be used to represent very small or very large numbers (for example:  $2.53 \times 10^{-26}$ ,  $1.7562 \times 10^{+35}$ , etc.).
- **Solution :: use floating-point number representation.**
  - A number  $F$  is represented as a triplet  $\langle s, M, E \rangle$  such that  $F = (-1)^s M \times 2^E$

$$F = (-1)^s M \times 2^E$$

- s is the *sign bit* indicating whether the number is negative (=1) or positive (=0).
- M is called the *mantissa*, and is normally a fraction in the range [1.0,2.0].
- E is called the *exponent*, which weights the number by power of 2.

### Encoding:

- Single-precision numbers: total 32 bits, E 8 bits, M 23 bits
- Double-precision numbers: total 64 bits, E 11 bits, M 52 bits



# Points to Note

- The number of *significant digits* depends on the number of bits in M.
  - 7 significant digits for 24-bit mantissa (23 bits + 1 implied bit).
- The *range* of the number depends on the number of bits in E.
  - $10^{38}$  to  $10^{-38}$  for 8-bit exponent.

## How many significant digits?

$$2^{24} = 10^x$$

$$24 \log_{10} 2 = x \log_{10} 10$$

$x = 7.2$  -- 7 significant decimal places

## Range of exponent?

$$2^{127} = 10^y$$

$$127 \log_{10} 2 = y \log_{10} 10$$

$y = 38.1$  -- maximum exponent value  
38 (in decimal)



# “Normalized” Representation

- We shall now see how  $E$  and  $M$  are actually encoded.
- Assume that the actual exponent of the number is  $EXP$  (i.e. number is  $M \times 2^{EXP}$ ).
- Permissible range of  $E$ :  $1 \leq E \leq 254$  (the all-0 and all-1 patterns are not allowed).
- Encoding of the exponent  $E$ :
  - The exponent is encoded as a biased value:  $E = EXP + BIAS$   
where  $BIAS = 127$  ( $2^{8-1} - 1$ ) for single-precision, and  $BIAS = 1023$  ( $2^{11-1} - 1$ ) for double-precision.

## • Encoding of the mantissa M:

–The mantissa is coded with an implied leading 1 (i.e. in 24 bits).

$$M = 1 . xxxx...x$$

–Here,  $xxxx...x$  denotes the bits that are actually stored for the mantissa. We get the extra leading bit for *free*.

–When  $xxxx...x = 0000...0$ , M is minimum (= 1.0).

–When  $xxxx...x = 1111...1$ , M is maximum (=  $2.0 - \epsilon$ ).

# An Encoding Example

- Consider the number  $F = 15335$   
 $15335_{10} = 11101111100111_2 = 1.1101111100111 \times 2^{13}$
- Mantissa will be stored as:  $M = 1101111100111\ 0000000000_2$
- Here,  $EXP = 13$ ,  $BIAS = 127$ .  $\rightarrow E = 13 + 127 = 140 = 10001100_2$

0	10001100	110111110011100000000000	466F9C00 in hex
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## Another Encoding Example

- Consider the number  $F = -3.75$

$$-3.75_{10} = -11.11_2 = -1.111 \times 2^1$$

- Mantissa will be stored as:  $M = 111000000000000000000000_2$
- Here,  $EXP = 1$ ,  $BIAS = 127$ .  $\rightarrow E = 1 + 127 = 128 = 10000000_2$



# Special Values

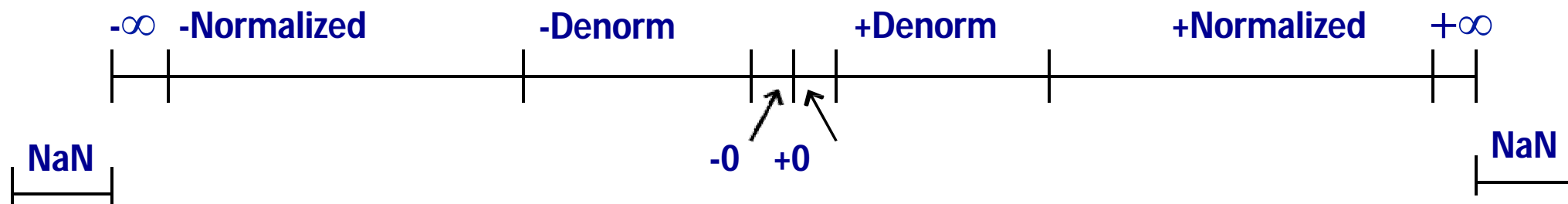
- When  $E = 000\dots 0$ 
  - $M = 000\dots 0$  represents the value 0.
  - $M \neq 000\dots 0$  represents numbers very close to 0.
- When  $E = 111\dots 1$ 
  - $M = 000\dots 0$  represents the value  $\infty$  (infinity).
  - $M \neq 000\dots 0$  represents *Not-a-Number* (NaN).

Zero is represented by the all-zero string.

Also referred to as *de-normalized* numbers.

NaN represents cases when no numeric value can be determined, like uninitialized values,  $\infty * 0$ ,  $\infty - \infty$ , square root of a negative number, etc.

# Summary of Number Encodings



Denormal numbers have very small magnitudes (close to 0) such that trying to normalize them will lead to an exponent that is below the minimum possible value.

- Mantissa with leading 0's and exponent field equal to zero.
- Number of significant digits gets reduced in the process.

# Rounding

- Suppose we are adding two numbers (say, in single-precision).
  - We add the mantissa values after shifting one of them right for exponent alignment.
  - We take the first 23 bits of the sum, and discard the residue R (beyond 32 bits).
- IEEE-754 format supports four rounding modes:
  - a) Truncation
  - b) Round to  $+\infty$  (similar to ceiling function)
  - c) Round to  $-\infty$  (similar to floor function)
  - d) Round to nearest

- To implement rounding, two temporary bits are maintained:
  - *Round Bit (r)*: This is equal to the MSB of the residue  $R$ .
  - *Sticky Bit (s)*: This is the logical OR of the rest of the bits of the residue  $R$ .
- Decisions regarding rounding can be taken based on these bits:

a)  $R > 0$ :      If  $r + s = 1$

b)  $R = 0.5$ :      If  $r \cdot s' = 1$                       // '+' is logical OR, '.' is logical AND

c)  $R > 0.5$ :      If  $r \cdot s = 1$

- Renormalization after Rounding:

- If the process of rounding generates a result that is not in normalized form, then we need to re-normalize the result.





# Floating Point Addition/Subtraction

- Two numbers:  $M1 \times 2^{E1}$  and  $M2 \times 2^{E2}$ , where  $E1 > E2$  (say).
- Basic steps:
  - Select the number with the smaller exponent (i.e.  $E2$ ) and shift its mantissa right by  $(E1-E2)$  positions.
  - Set the exponent of the result equal to the larger exponent (i.e.  $E1$ ).
  - Carry out  $M1 \pm M2$ , and determine the sign of the result.
  - Normalize the resulting value, if necessary.



## Addition Example

- Suppose we want to add  $F1 = 270.75$  and  $F2 = 2.375$

$$F1 = (270.75)_{10} = (100001110.11)_2 = 1.0000111011 \times 2^8$$

$$F2 = (2.375)_{10} = (10.011)_2 = 1.0011 \times 2^1$$

- Shift the mantissa of F2 right by  $8 - 1 = 7$  positions, and add:

1000 0111 0110 0000 0000 0000

1 0011 0000 0000 0000 0000 000

---

1000 1000 1001 0000 0000 0000 0000 000

**Residue**

- Result:  $1.00010001001 \times 2^8$

## Subtraction Example

- Suppose we want to subtract  $F2 = 224$  from  $F1 = 270.75$

$$F1 = (270.75)_{10} = (100001110.11)_2 = 1.0000111011 \times 2^8$$

$$F2 = (224)_{10} = (11100000)_2 = 1.111 \times 2^7$$

- Shift the mantissa of F2 right by  $8 - 7 = 1$  position, and subtract:

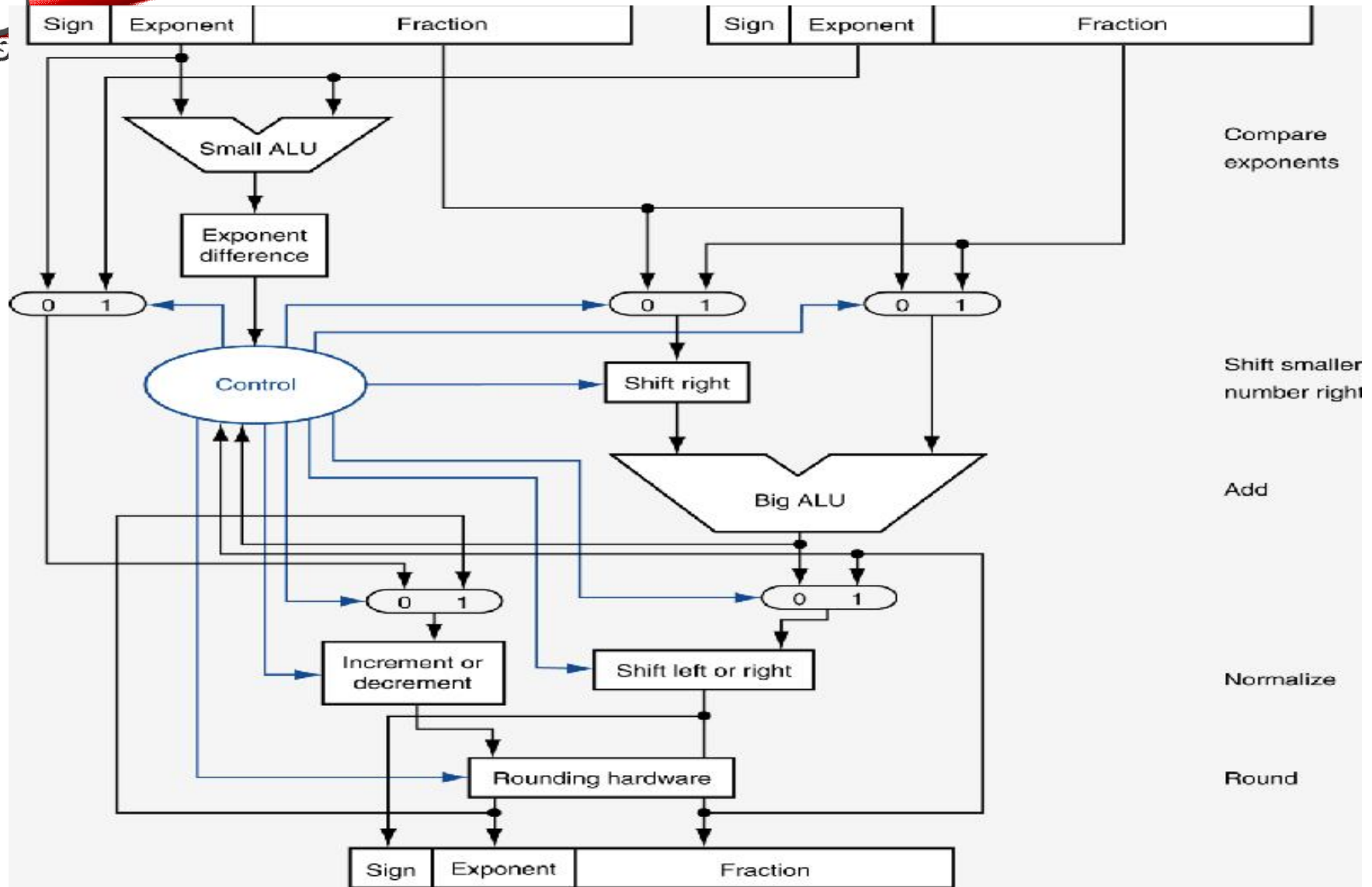
1000 0111 0110 0000 0000 0000

111 0000 0000 0000 0000 0000 000

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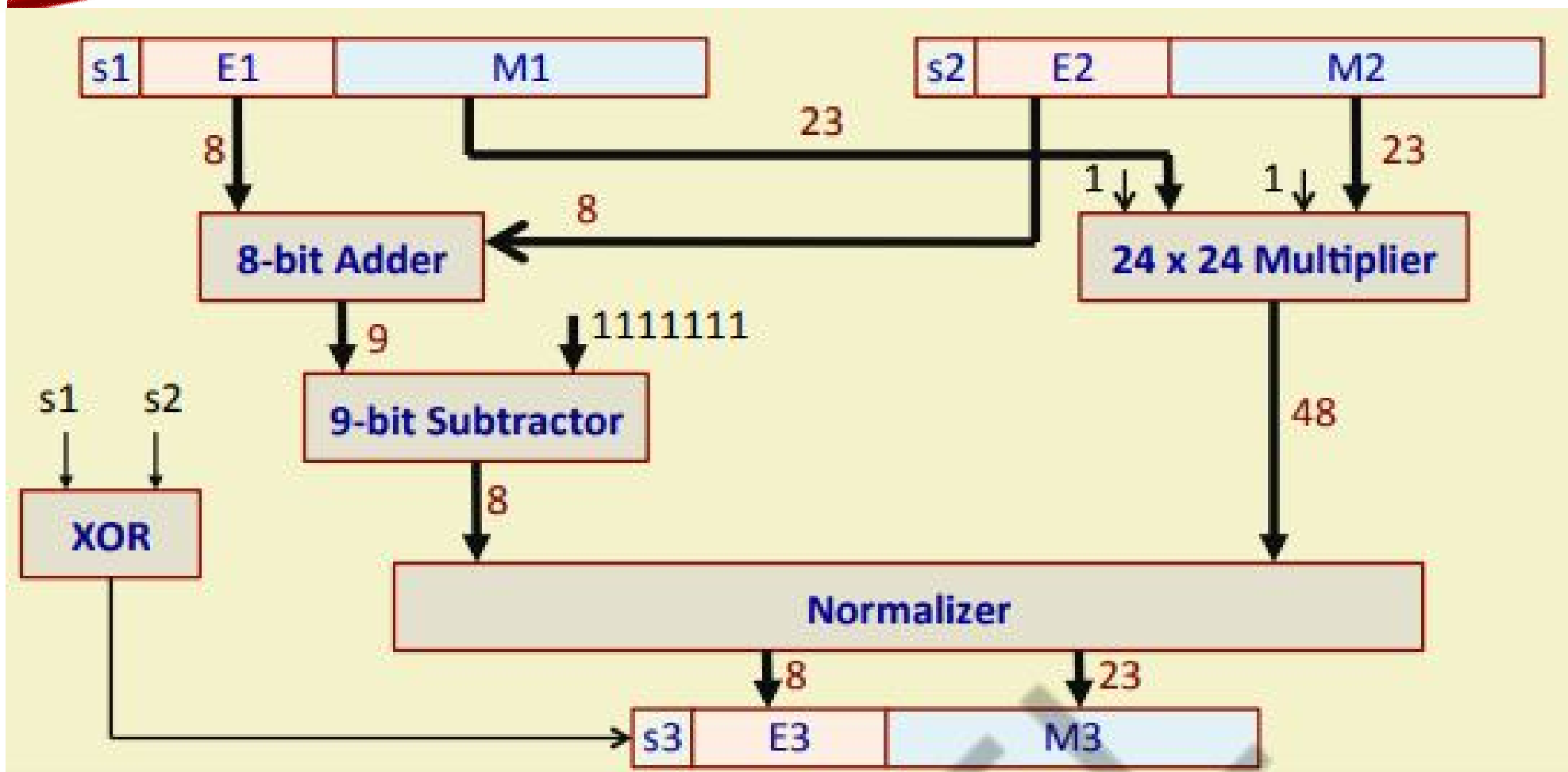
0001 0111 0110 0000 0000 0000 000

- For normalization, shift mantissa left 3 positions, and decrement E by 3.
- Result:  $1.01110110 \times 2^5$



# Floating-Point Multiplication

- Two numbers:  $M1 \times 2^{E1}$  and  $M2 \times 2^{E2}$
- Basic steps:
  - Add the exponents  $E1$  and  $E2$  and subtract the  $BIAS$ .
  - Multiply  $M1$  and  $M2$  and determine the sign of the result.
  - Normalize the resulting value, if necessary.



# Floating-Point Division

- Two numbers:  $M1 \times 2^{E1}$  and  $M2 \times 2^{E2}$
- Basic steps:
  - Subtract the exponents  $E1$  and  $E2$  and add the  $BIAS$ .
  - Divide  $M1$  by  $M2$  and determine the sign of the result.
  - Normalize the resulting value, if necessary.

## Division Example

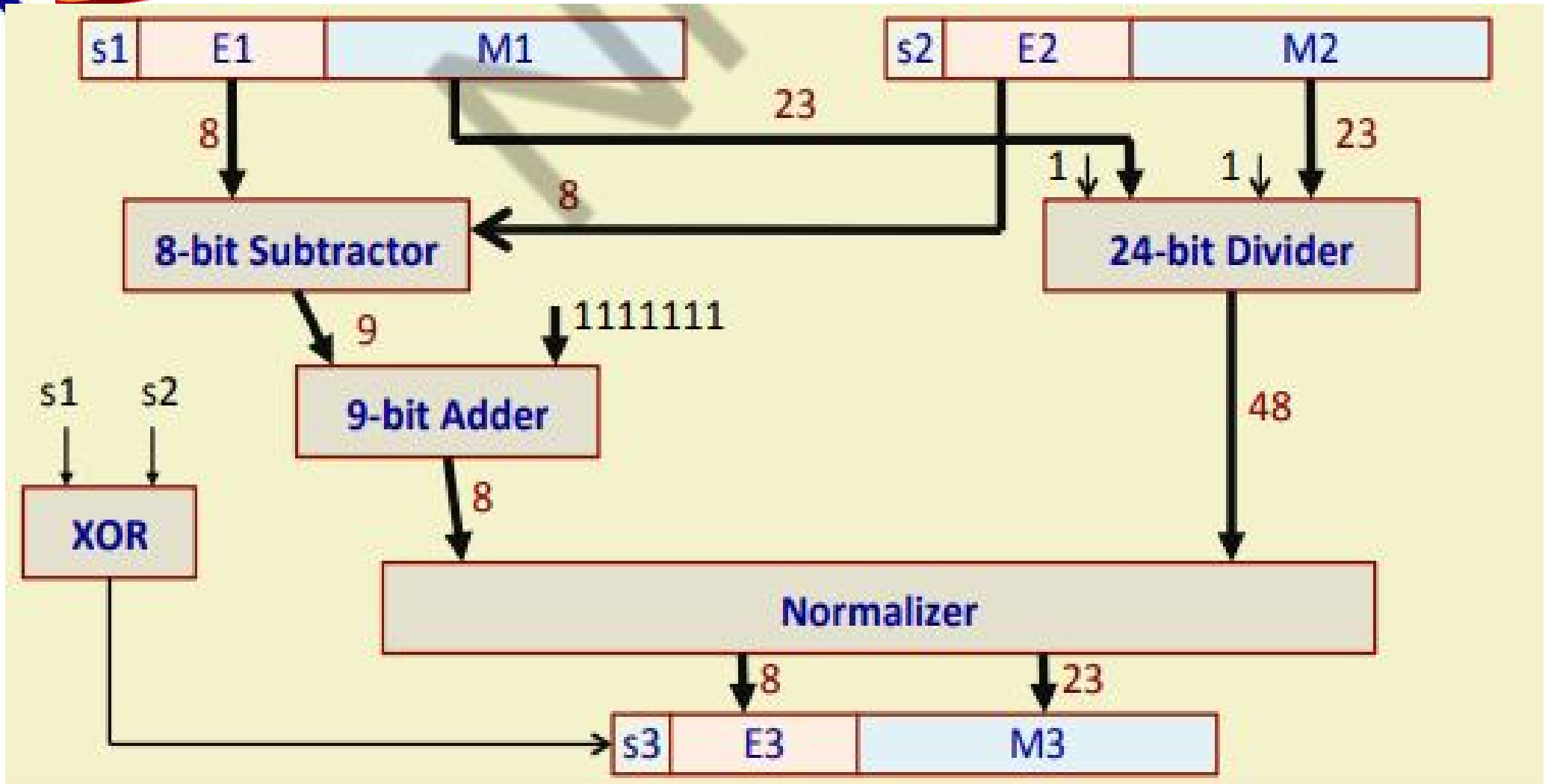
- Suppose we want to divide  $F1 = 270.75$  by  $F2 = -2.375$

$$F1 = (270.75)_{10} = (100001110.11)_2 = 1.0000111011 \times 2^8$$

$$F2 = (-2.375)_{10} = (-10.011)_2 = -1.0011 \times 2^1$$

- Subtract the exponents:  $8 - 1 = 7$
- Divide the mantissas:  $0.1110010$
- Result:  $0.1110010 \times 2^7$
- After normalization:  $1.110010 \times 2^6$









## TEXT BOOK

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# THANK YOU