

# **Module 11**

# **Arithmetic Operations**

**S. Lakshmivarahan**  
**School of Computer Science**  
**University of Oklahoma**  
**USA-73019**  
**Varahan@ou.edu**

# Arithmetic Operations

---

- Our goal is to examine faster algorithms to add and multiply two  $n$ -bit integers
- To this end, we start by examining the grade school method for add and multiply

# Binary Addition

## Grade School Method

---

- Recall the basic binary addition table

X	Y	Sum	Carry
0	0	0	0
0	1	1	0
1	0	1	0
1	1	1	1

# Binary Addition

## Grade School Method

---

- Let  $n=2^k$  and let  $a$  and  $b$  be two  $n$ -bit integers given by

$$a = a_{n-1}a_{n-2} \dots a_1a_0$$

$$b = b_{n-1}b_{n-2} \dots b_1b_0$$

- Let  $s = s_ns_{n-1} \dots s_1s_0$  be the  $(n+1)$  bit integer that represents the sum  $s = a+b$

# Binary Addition

## Grade School Method

---

- Then,

	$c_n$	$c_{n-1}$	$c_{n-2}$		$c_2$	$c_1$	$c_0$
a=		$a_{n-1}$	$a_{n-2}$	...	$a_2$	$a_1$	$a_0$
b=		$b_{n-1}$	$b_{n-2}$	...	$b_2$	$b_1$	$b_0$
<hr/>							
s=	$s_n$	$s_{n-1}$	$s_{n-2}$	...	$s_2$	$s_1$	$s_0$

- Where  $c_0=0$  and  $s_n=c_n$ . The rest of the  $s_i$ ,  $0 \leq i \leq n - 1$ , are computed using the binary addition table.
-

# Binary Addition

## Grade School Method

---

- The total number of bit additions needed to compute  $s$  is  $2n = O(n)$  and is known as the Boolean complexity.
- Note: In the unit cost model, the unit was taken as the time required to perform one basic operation-  $+$ ,  $-$ ,  $*$ ,  $\%$ , and  $\geq$ .
- However, in here our basic unit is the time taken to perform a one bit binary- AND, OR, NOT operation. Thus, the unit in this case (boolean complexity) is much smaller

# Binary Addition

## Grade School Method

---

- Example: Let  $n=4$ ,  $a=11$ ,  $b=15$ . Then  $s=26$

	1	1	1	1	0
a=		1	0	1	1
b=		1	1	1	1
<hr/>					
s=	1	1	0	1	0

# Binary Multiply Grade School Method

---

- The basic binary multiplication table

X	Y	Product
0	0	0
0	1	0
1	0	0
1	1	1



# Binary Multiply Grade School Method

- Let  $n=2^k$  and let  $a$  and  $b$  be two  $n$  bit integers
- Example:  $n=4$ ,  $a=11$ ,  $b=15$ ,  $ab=165$

			a=		1	0	1	1	
			b=		1	1	1	1	
					1	0	1	1	← n-bit multiply
				1	0	1	1		← n-bit multiply
			1	0	1	1			← n-bit multiply
		1	0	1	1				← n-bit multiply
ab=	1	0	1	0	0	1	0	1	← (n-1) addition of the above n n-bit integers

# Binary Multiply Grade School Method

---

- Thus, it takes  $n^2$  bit multiplications and  $(n-1)$  bit additions, each of which takes  $O(n)$  bit additions
- Thus, it takes  $O(n^2)$  bit operations to multiply two  $n$ -bit integers

# A faster method for multiplication of integers

---

- Let  $n=2^k$  and let  $a = a_{n-1}a_{n-2} \dots a_2a_1a_0$  and  $b=b_{n-1}b_{n-2}\dots b_1b_0$  be the  $n$  bit integers to be multiplied.
- Consider the polynomials  $A(x)$  and  $B(x)$  in  $x$  of degree  $n-1$ , given by

$$A(x) = \sum_{i=0}^{n-1} a_i x^i, \quad a_i \in \{0,1\} \quad \rightarrow a = A(2)$$

$$B(x) = \sum_{i=0}^{n-1} b_i x^i, \quad b_i \in \{0,1\} \quad \rightarrow b = B(2)$$

# A faster method for multiplication of integers

---

- Thus, there is a close correspondence between the integer arithmetic and polynomial arithmetic
- We now describe a divide and conquer based algorithm to multiply two polynomials
- This algorithm is due to Karatsuba and Ofman

# A faster method for multiplication of integers

- 
- First, divide the polynomials  $A(x)$  into two equal parts:

$$A(x) = A_L(x) + x^{n/2} A_H(x)$$

where

$$A_L(x) = \sum_{i=0}^{n/2-1} a_i x^i$$

$$A_H(x) = \sum_{i=0}^{n/2-1} a_i x^i$$

each of which is a polynomial of degree  $(\frac{n}{2} - 1)$

- Similarly,

$$B(x) = B_L(x) + x^{n/2} B_H(x)$$

---

# A faster method for multiplication of integers

---

- As an example, let  $n = 4$ . Then

$$\begin{aligned} A(x) &= a_0 + a_1x + a_2x^2 + a_3x^3 \\ &= (a_0 + a_1x) + x^2(a_2 + a_3x) \\ &= A_L(x) + x^2A_H(x) \end{aligned}$$

where

$$A_L(x) = a_0 + a_1x$$

$$A_H(x) = a_2 + a_3x$$

# A faster method for multiplication of integers

---

- Let  $c = ab$  and let  $C(x)$  be the polynomial corresponding to the product  $c$
- Thus,  $C(x) = A(x) * B(x)$
- Then, substituting for  $A(x)$  and  $B(x)$ :

$$C(x) = (A_L(x) + x^{n/2}A_H(x)) * (B_L(x) + x^{n/2}B_H(x)) \rightarrow \textcircled{1}$$

where  $A_L = A_L(x)$  and so on

# A faster method for multiplication of integers

---

- If we multiply the right hand side, then we get

$$C(x) = A_L B_L + x^{\frac{n}{2}}(A_L B_H + A_H B_H) + x^n A_H B_H$$

- Since  $x=2$ , multiplying by  $2^{\frac{n}{2}}$  and  $2^n$  is equivalent to shifting the binary point to the right by  $\frac{n}{2}$  and  $n$  locations respectively.  
Hence, these multiplications do not take much time.



# A faster method for multiplication of integers

---

- So, when we compute the r.h.s. of ①, we are left with 4 multiplications of polynomials of degree  $\frac{n}{2} - 1$ .

$$A_L B_L, \quad A_L B_H, \quad A_H B_L, \quad \text{and} \quad A_H B_H,$$

and three polynomial additions of polynomials of degree at most  $2n-2$ .

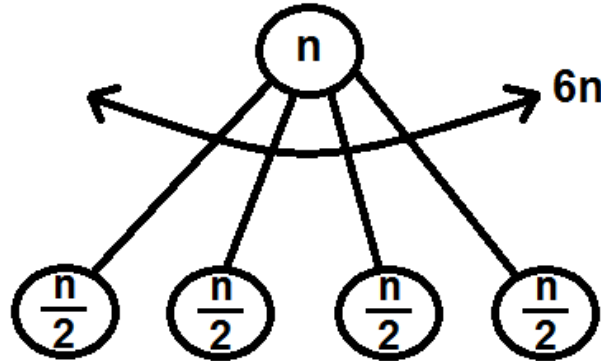
# A faster method for multiplication of integers

---

- Recall, each polynomial addition takes linear time in the degree of the polynomial.
- Hence, the three additions together takes no more than  $3 \times (2n-2) \leq 6n$  operations

# A faster method for multiplication of integers

- If  $T(n)$  is the time to multiply two polynomials of degree  $(n-1)$  each, then we get the following recursive tree that describes the above algorithm.



- $\therefore T(n) = 4T\left(\frac{n}{2}\right) + 6n$ , with  $n=2^k$  and  $T(1) = 1 \rightarrow \textcircled{2}$

# A faster method for multiplication of integers

---

- Solving ②, it follows that

$$T(n) = O(n^{\log_2 4}) = O(n^2) \rightarrow \textcircled{3}$$

- That is, we divided but did not conquer it since ③ is also the time for the grade school method
- The conquering part is one that was introduced by Karatsuba and Ofman

# Karatsuba-Ofman Algorithm

---

- To compute the product  $C(x)$ , first compute three intermediate polynomials
- Let

$$\left. \begin{aligned} C_L &= A_L B_L \\ C_H &= A_H B_H \\ C_M &= (A_L + A_H)(B_L + B_H) \end{aligned} \right\} \textcircled{4}$$

involving a total of 3 multiplications of polynomials of degree  $(\frac{n}{2} - 1)$ , (instead of the 4 such multiplications previously) and two additions.

# Karatsuba-Ofman Algorithm

---

- This is similar in idea to the Strassen's method, which trades the expensive multiplication for the cheap additions.
- Then, assemble the product as

$$C(x) = C_L + [C_M - C_L - C_H]x^{\frac{n}{2}} + C_Hx^n \rightarrow \textcircled{5}$$

- In  $\textcircled{5}$ , there are only 4 additions and there are no multiplications except those by  $x^n$  and  $x^{\frac{n}{2}}$  which are very inexpensive.

# Karatsuba-Ofman Algorithm

---

- Combining,

$$T(n) = 3T\left(\frac{n}{2}\right) + \alpha n \rightarrow \textcircled{6}$$

where  $\alpha n$  is the time to perform 6 additions of polynomials of degree less than or equal to  $2n-2$ . Hence, the constant in  $\textcircled{6}$  is  $\alpha \leq 12$ .

# Karatsuba-Ofman Algorithm

---

- Solving ⑥, it follows that

$$\begin{aligned}T(n) &= O(n^{\log_2 3}) \\ &= O(n^{1.58})\end{aligned}$$

which is much faster than the grade school method



# Homework

---

1. Multiplying the r.h.s. of (5), verify that you get the same answer as in (1).
2. Solve the recurrences in (2) and in (5) exactly and verify the answers

# Faster Multiplication

---

- The question now comes: can we multiply integers ever faster than  $O(n^{1.58})$ .
- The answer is indeed Yes! The fastest known algorithm takes only  $O(\log n)$  time and is due to Schonhage and Strassen (1971)

# Faster Multiplication

---

- A detailed analysis of this parallel algorithm is given in Chapter 3 of the following book:
- S. Lakshmivarahan and S.H. Dhall (1990). Analysis and Design of Parallel Algorithms: Arithmetic and Matrix Problems, McGraw Hill, New York

# A Fast Parallel Algorithm for Integer Addition

---

- Let  $s = a + b$  where

$$a = a_{n-1} a_{n-2} \dots a_2 a_1 a_0$$

$$b = b_{n-1} b_{n-2} \dots b_2 b_1 b_0$$

- And

$$s = s_n s_{n-1} \dots s_2 s_1 s_0$$

# A Fast Parallel Algorithm for Integer Addition

- Recall that

$$s_i = a_i \oplus b_i \oplus c_i, 0 \leq i \leq n-1$$

$$s_n = c_n, c_0 = 0 \text{ and for } 1 \leq i \leq n$$

$$c_i = (a_i \wedge b_i) \vee (a_i \wedge c_{i-1}) \vee (b_i \wedge c_{i-1})$$

①

where

$\wedge$  is the Boolean AND

$\vee$  is the Boolean OR

$\oplus$  is the exclusive OR

$\wedge$	0	1
0	0	0
1	0	1

$\vee$	0	1
0	0	1
1	1	1

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

# A Fast Parallel Algorithm for Integer Addition

---

- The expression for  $c_i$  can be rewritten as

$$c_i = g_i \vee (p_i \wedge c_{i-1}) \rightarrow \textcircled{2}$$

where

$g_i = a_i \wedge b_i$ , the generate bit

$p_i = a_i \vee b_i$ , the propagate bit

- That is,  $c_i$  in  $\textcircled{2}$  is given by the first order linear Boolean recurrence

# A Fast Parallel Algorithm for Integer Addition

---

- We can write ② as

$$c_i = f_i(c_{i-1}) \rightarrow \textcircled{3}$$

where the Boolean function depends on two Boolean parameters  $g_i$  and  $p_i$ ,  $1 \leq i \leq n$

# A Fast Parallel Algorithm for Integer Addition

---

- Iterating ③, we readily obtain

$$c_1 = f_1(c_0)$$

$$c_2 = f_2(c_1) = f_2(f_1(c_0)) = f_1 \circ f_2(c_0)$$

$$c_3 = f_3(c_2) = f_1 \circ f_2 \circ f_3(c_0) \quad \text{④}$$

...

$$c_n = f_1 \circ f_2 \circ f_3 \circ \dots \circ f_n(c_0)$$

where  $f_1 \circ f_2(c_0) = f_2(f_1(c_0))$



# Property of the Composition

---

- $f_1 \circ f_2 (c_0)$  is the binary operation of the composition of  $f_1$  and  $f_2$  (from left to write)
- What is  $f_1 \circ f_2 (c_0)$ ?
- Recall,  $f_i \circ f_j (x) = f_j(f_i(x))$   
$$= f_j(g_i \vee (p_i \wedge x))$$
$$= g_j \vee (p_j \wedge [g_i \vee (p_i \wedge x)])$$
$$= [g_j \vee (p_j \wedge g_i)] \vee [(p_j \wedge p_i) \wedge x]$$

# A Fast Parallel Algorithm for Integer Addition

---

- Thus,  $f_i \circ f_j (x) = G \vee (P \wedge x) \rightarrow \textcircled{5}$
- Where  $G = g_i \vee (p_i \wedge g_j)$  is the new generate bit and  $P = (p_i \wedge p_j)$  is the new propagate bit
- Thus, the functions  $f_i$  and  $f_j$  are composed, and the resulting function is of the same type as  $f_i$  and  $f_j$
- This is called the reproducibility property under composition.

# Homework

---

3. Verify that

$$\begin{aligned} f_1 \circ f_2 \circ f_3(x) &= (f_1 \circ f_2) \circ f_3(x) \\ &= f_1 \circ (f_2 \circ f_3)(x) \end{aligned}$$

4. Thus, the composition of functions is an associative, binary operation

# Parallel Algorithm for Integer Addition

---

- Our goal is to compute the parameters of the series of  $n$  functions in ④ repeated here for convenience:

$$f_1$$

$$f_1 \circ f_2$$

$$f_1 \circ f_2 \circ f_3 \quad \text{⑥}$$

...

$$f_1 \circ f_2 \circ f_3 \circ \dots \circ f_n$$

# Parallel Algorithm for Integer Addition

---

- Since each function in ⑥ is of the type ⑤, we can evaluate them all at  $c_0$  to get all the carry bits  $c_1$  to  $c_n$ .
- Computing these composite functions where the composition of functions is an associative binary operation is known as the prefix problem.

# Parallel Algorithm for Integer Addition

---

- There are circuit based efficient parallel algorithms to compute these composite functions in  $O(\log n)$  time.
- To enable this, we now digress to describe the prefix problem and some of the ideas to compute them in parallel.

# The Prefix Problem

---

- Let  $\{x_1, x_2, \dots, x_n\}$  be a set of  $n$  numbers, and let  $+$  denote an associative binary operation- such as addition or multiplication of numbers.
- Our goal is to compute  $\textcircled{6}$ .
- This is called the prefix problem, all partial sum problem, etc.

# The Prefix Problem

---

- To compute these serially, let  $(y_1, y_2, \dots, y_n)$  be an array. Then,

$$y_1 = x_1$$

DO  $i = 2$  to  $n$

$$y_i = y_{i-1} + x$$

END DO

is the serial algorithm to compute the  $n$  prefixes in ⑥



# The Prefix Problem

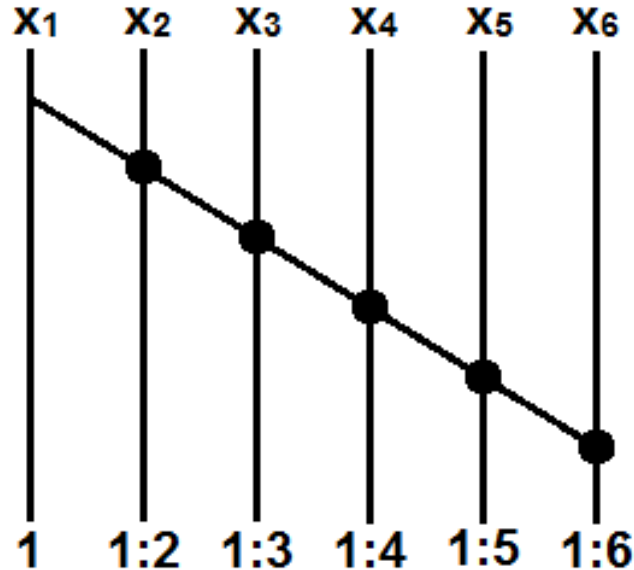
---

- Clearly, the serial time  $T_s(n) = n-1$ .
- The question now is: can we compute ⑥ in parallel in a time faster than  $O(n)$ ?
- To answer this, we first represent the serial algorithm as an  $n$ -input,  $n$ -output layered circuit as shown below.

# The Prefix Problem

## 1. Serial Circuit

- Let  $n=6$ :



- denotes the operation of addition of two inputs into it

$$i:j = x_i + x_{i+1} + \dots + x_j \quad \text{for } i \leq j$$

# The Prefix Problem: Serial Circuit

---

- There are two parameters that describe the performance of this circuit- size and depth
- The size ( $n$ ) = total number of operation nodes.
- For the serial circuit with  $n=6$ ,  $\text{size}(6) = 5$ . In general, the size of the serial circuit denotes the total number of operations, and  $s(n) = n-1$

# The Prefix Problem: Serial Circuit

---

- The depth,  $d(n)$  of the serial circuit is the length of the longest path- it denotes the time.
- For the illustration with  $n=6$ ,  $d(6) = 5$ . In general,  $d(n) = n-1$

# The Prefix Problem: Serial Circuit

---

- Another quantity of interest is called the size+depth
- In general for the serial circuits,

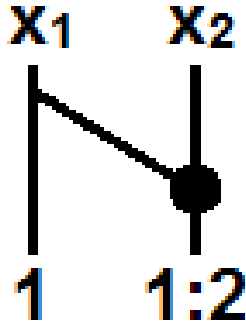
$$s(n) + d(n) = 2n - 2$$

- We move to describing parallel circuits to compute the prefixes with depth of order of  $\log n$ .

# The Prefix Problem: DC Circuit

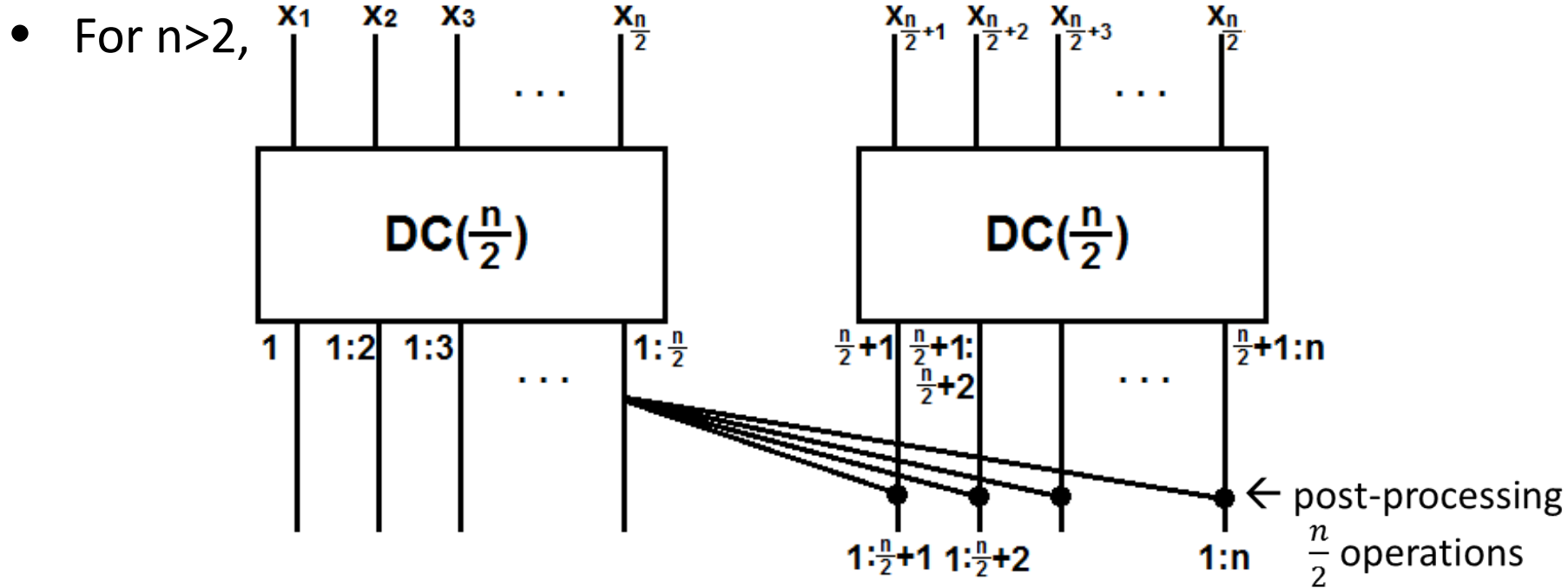
---

1. Divide and Conquer Circuit: (DC-Circuit)
  - Let  $n=2^k$ . For  $n=2$ , the DC circuit  $DC(2)$  is given by



with  $s(n)=1$ ,  $d(n)=1$

# The Prefix Problem: DC Circuit



# The Prefix Problem: DC Circuit

---

- The structure is quite explanatory. The last output of the first  $DC(\frac{n}{2})$  box is combined with each out of the second  $DC(\frac{n}{2})$  box to get the overall correct results.



# The Prefix Problem: DC Circuit

---

- Clearly,

$$\left. \begin{aligned} s(n) &= 2s\left(\frac{n}{2}\right) + \frac{n}{2}, \quad s(2) = 1 \\ d(n) &= d\left(\frac{n}{2}\right) + 1, \quad d(2) = 1 \end{aligned} \right\} \textcircled{8}$$

- Solving, we get

$$\left. \begin{aligned} s(n) &= \frac{n}{2} \log n \\ d(n) &= \log n \end{aligned} \right\} \textcircled{9}$$

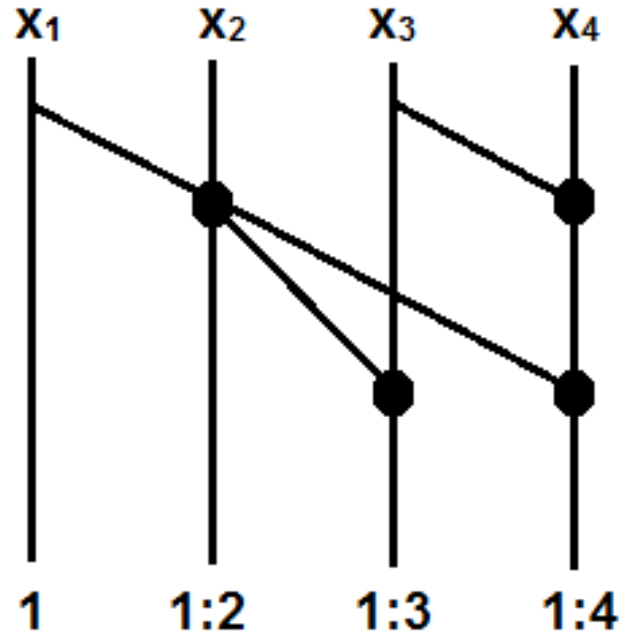
# The Prefix Problem

---

- **Homework**: Solve ⑧ and verify the answers in ⑨.
- While we have achieved our goal of  $\log n$ , it is achieved at the cost of performing a whole lot of operations of the order  $n \log n$ .

# DC Circuit Examples

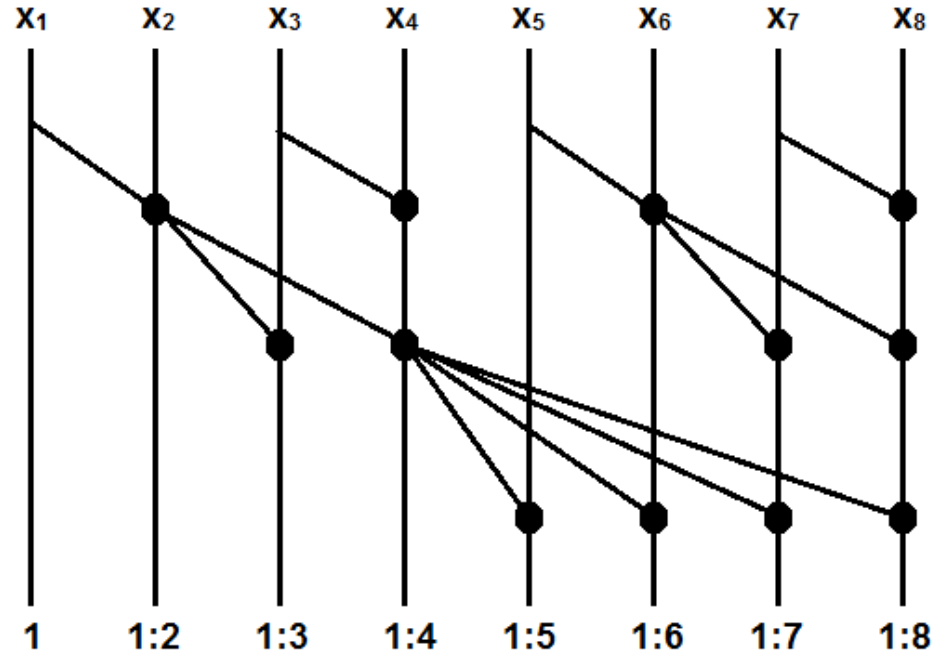
- DC(4):



$$s(4) = 4$$
$$d(n) = 2$$

# DC Circuit Examples

- DC(8):



$$s(8) = 12$$
$$d(8) = 3$$

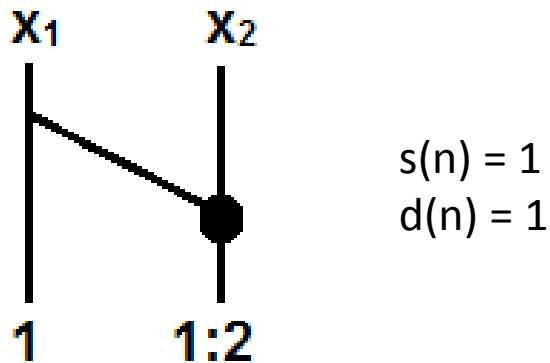
# The Prefix Problem

---

- **Homework**: Draw  $DC(16)$ . Compute  $s(16)$ ,  $d(16)$ .
- We now provide a third class of circuits that, while it keeps the depth to be  $O(\log n)$ , it reduced the size to  $O(n)$  instead of  $O(n \log n)$ .

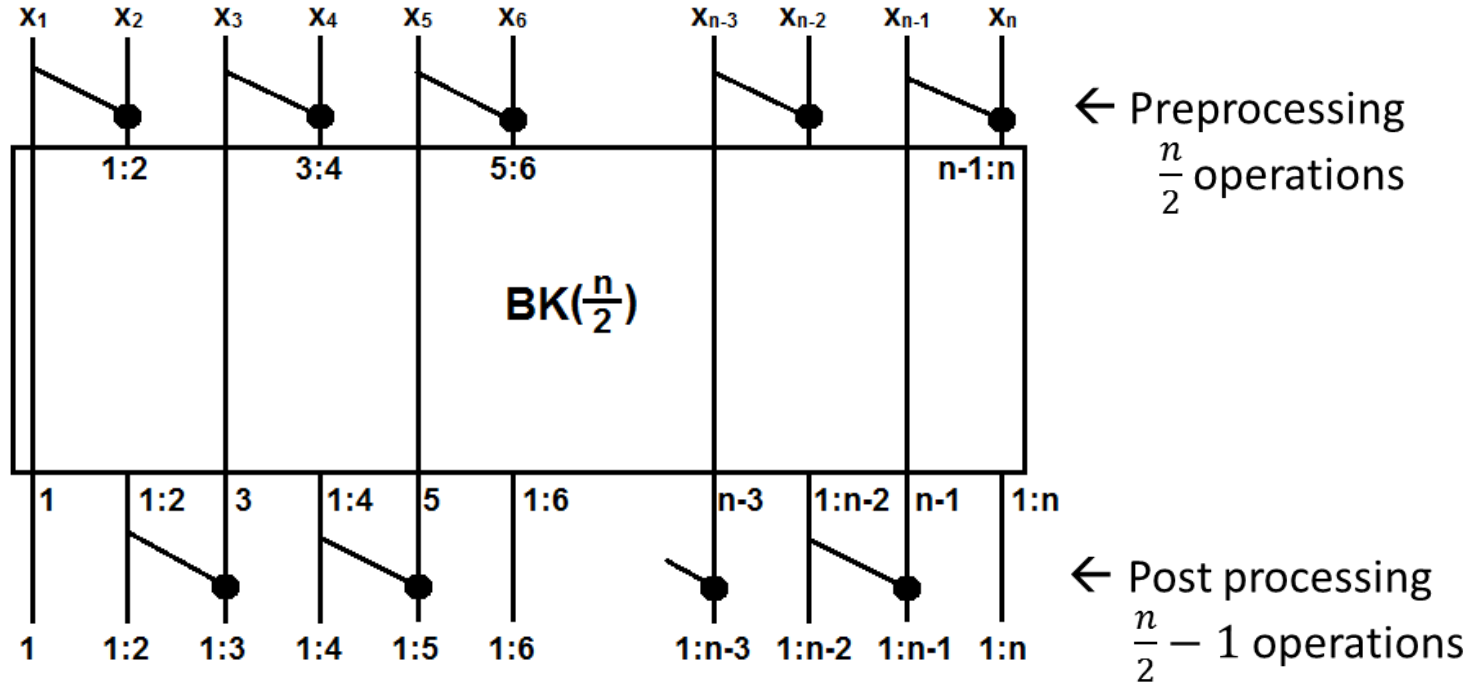
# Prefix Problem: Brent-Kung Circuit

- Brent-kung prefix circuit: BK(N)
- Let  $n=2^k$
- BK(2):



# Prefix Problem: Brent-Kung Circuit

- $BK(n)$



# Prefix Problem: Brent-Kung Circuit

---

- Now

$$\left. \begin{aligned} s(n) &= s\left(\frac{n}{2}\right) + (n - 1) \\ d(n) &= d\left(\frac{n}{2}\right) + 2 \end{aligned} \right\} \textcircled{10}$$

- Hence

$$\left. \begin{aligned} s(n) &= 2n - \log n - 2 \\ d(n) &= 2 \log n - 2 \end{aligned} \right\} \textcircled{11}$$

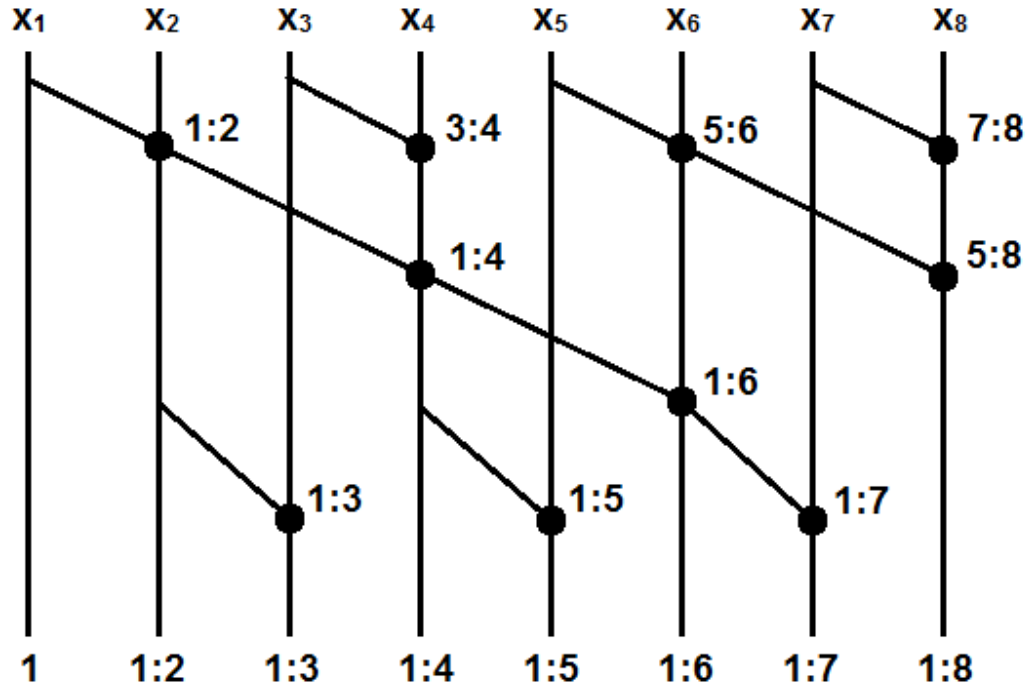


# Prefix Problem: Brent-Kung Circuit

---

- Homework: Solve ⑩
- Notice that while the depth is still  $O(\log n)$ , the size has reduced considerably to linear in  $n$ .

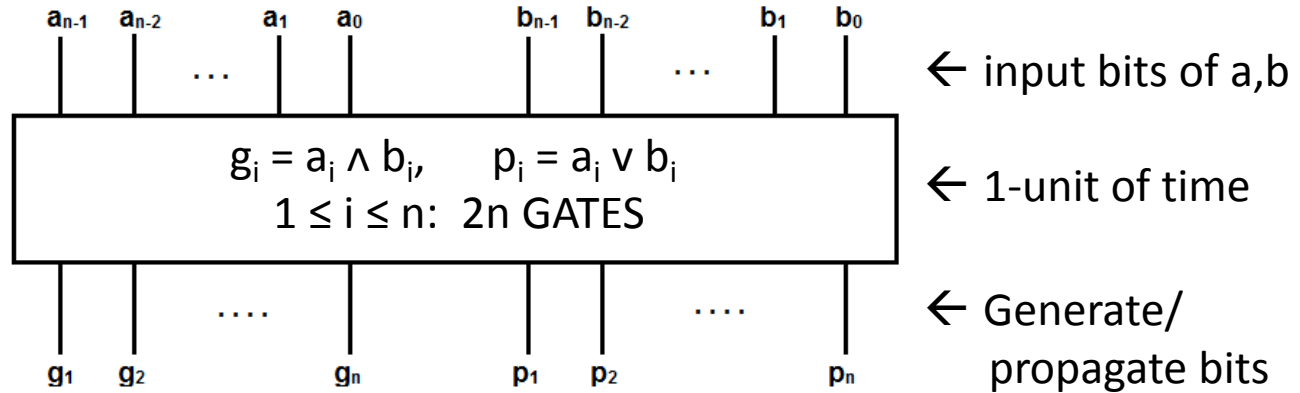
# Example of BK(8)



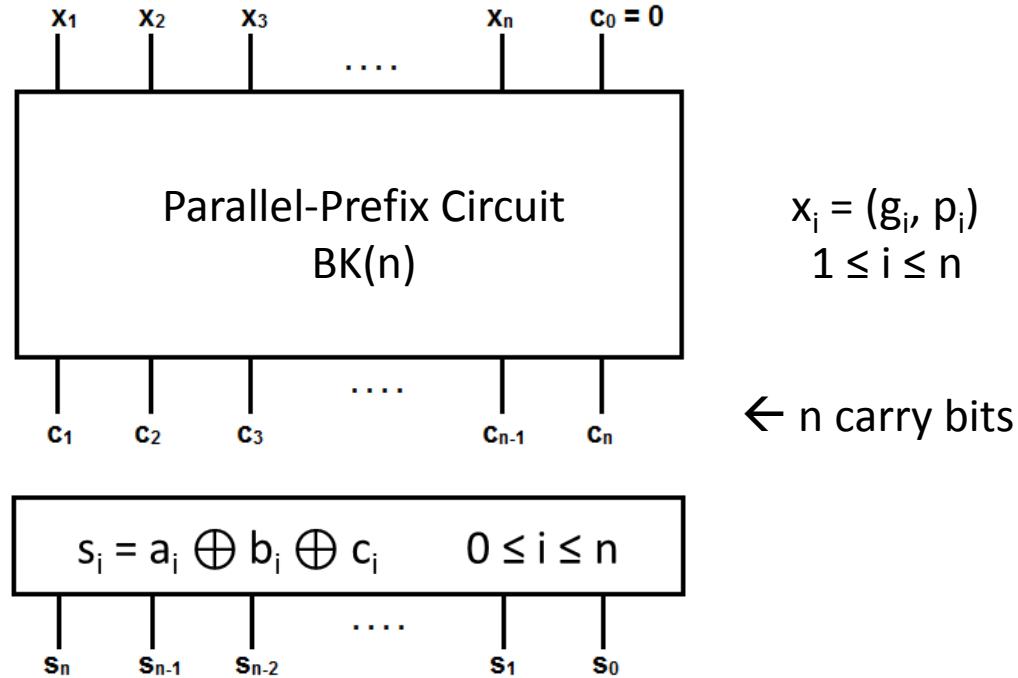
$$s(n) = 11$$
$$d(n) = 4$$

# Back to Parallel Adder Circuit

- We can now use BK(n) to compute the prefixes in in  $O(\log n)$  time.
- We now provide a layout of the parallel adder.



# Parallel Adder Circuit



# Parallel Adder Circuit

---

- Hence, overall depth is of the form  $\alpha + \beta \log n$  for some constants  $\alpha$  and  $\beta$ .
- This is the fastest known algorithm to add two  $n$ -bit integers.
- Reference: S. Lakshmivarahan and S. K. Dhall (1994). Parallel Computation using the Prefix Problem, Oxford University Press.