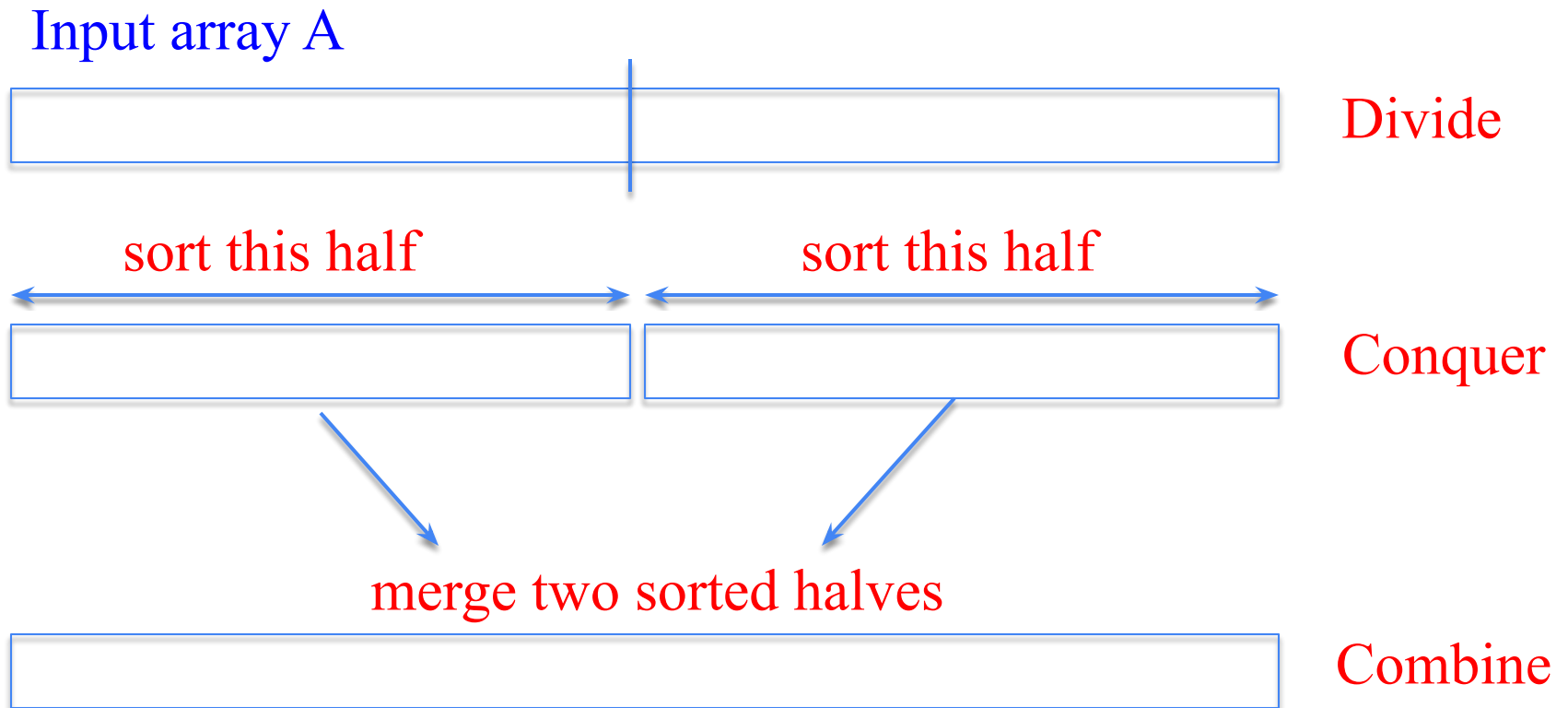


# CS473 - Algorithms I

## Lecture 4

### The Divide-and-Conquer Design Paradigm

# Reminder: Merge Sort



# The Divide-and-Conquer Design Paradigm

1. Divide the problem (instance) into subproblems.
2. Conquer the subproblems by solving them recursively.
3. Combine subproblem solutions.

# Example: Merge Sort

1. Divide: Trivial.
2. Conquer: Recursively sort 2 subarrays.
3. Combine: Linear- time merge.

$$T(n) = 2 T(n/2) + \Theta(n)$$

The diagram illustrates the recurrence relation  $T(n) = 2T(n/2) + \Theta(n)$ . Three arrows point from descriptive text below to parts of the equation above:

- An arrow points from the text "# subproblems" to the coefficient "2".
- An arrow points from the text "subproblem size" to the term "T(n/2)".
- An arrow points from the text "work dividing and combining" to the term " $\Theta(n)$ ".

# Master Theorem: Reminder

$$T(n) = aT(n/b) + f(n)$$

Case 1:

$$\frac{n^{\log_b a}}{f(n)} = \Omega(n^\epsilon)$$



$$T(n) = \Theta(n^{\log_b a})$$

Case 2:

$$\frac{f(n)}{n^{\log_b a}} = \Theta(\lg^k n)$$



$$T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$$

Case 3:

$$\frac{f(n)}{n^{\log_b a}} = \Omega(n^\epsilon)$$



$$T(n) = \Theta(f(n))$$

and  $a f(n/b) \leq c f(n)$  for  $c < 1$

# Merge Sort: Solving the Recurrence

$$T(n) = 2 T(n/2) + \Theta(n)$$

→  $a = 2, \quad b = 2, \quad f(n) = \Theta(n), \quad n^{\log_b a} = n$

Case 2:

$$\frac{f(n)}{n^{\log_b a}} = \Theta(\lg^k n)$$



$$T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$$

*holds for  $k = 0$*

→  $T(n) = \Theta(n \lg n)$

# Binary Search

Find an element in a sorted array:

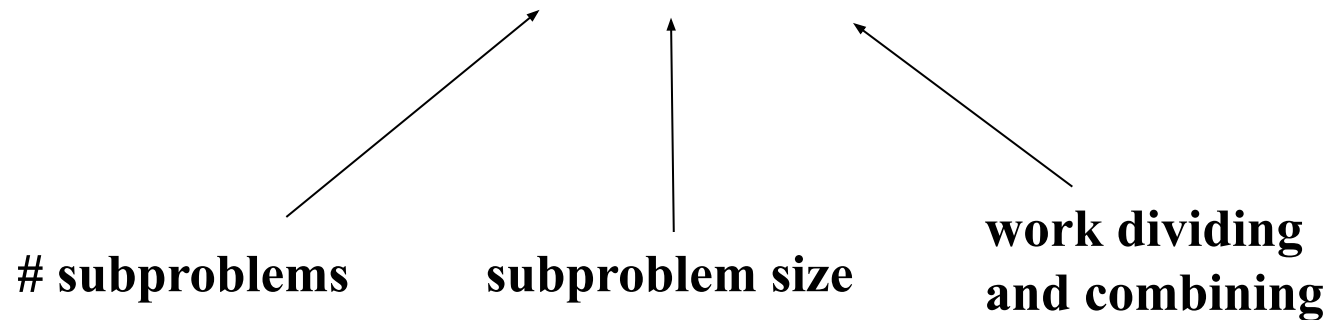
1. Divide: Check middle element.
2. Conquer: Recursively search 1 subarray.
3. Combine: Trivial.

Example: Find 9

3   5   7   8   9   12   15

# Recurrence for Binary Search

$$T(n) = 1 T(n/2) + \Theta(1)$$





# Binary Search: Solving the Recurrence

$$T(n) = T(n/2) + \Theta(1)$$

→  $a = 1, \quad b = 2, \quad f(n) = \Theta(1), \quad n^{\log_b a} = n^0 = 1$

Case 2:

$$\frac{f(n)}{n^{\log_b a}} = \Theta(\lg^k n)$$



$$T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$$

*holds for  $k = 0$*

→  $T(n) = \Theta(\lg n)$

# Powering a Number

- Problem: Compute  $a^n$ , where  $n$  is a natural number

Naive-Power (a, n)

powerVal  $\leftarrow$  1

for i  $\leftarrow$  1 to n

    powerVal  $\leftarrow$  powerVal . a

return powerVal

- What is the complexity?  $T(n) = \Theta(n)$

# Powering a Number: Divide & Conquer

Basic idea:

$$a^n = \begin{cases} a^{n/2} \cdot a^{n/2} & \text{if } n \text{ is } \underline{\text{even}} \\ a^{(n-1)/2} \cdot a^{(n-1)/2} \cdot a & \text{if } n \text{ is } \underline{\text{odd}} \end{cases}$$

# Powering a Number: Divide & Conquer

POWER (a, n)

**if** n = 0 **then return** 1

**else if** n is even **then**

val ← POWER (a, n/2)

**return** val \* val

**else if** n is odd **then**

val ← POWER (a, (n-1)/2)

**return** val \* val \* a

# Powering a Number: Solving the Recurrence

$$T(n) = T(n/2) + \Theta(1)$$

→  $a = 1, \quad b = 2, \quad f(n) = \Theta(1), \quad n^{\log_b a} = n^0 = 1$

Case 2:

$$\frac{f(n)}{n^{\log_b a}} = \Theta(\lg^k n)$$



$$T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$$

*holds for  $k = 0$*

→  $T(n) = \Theta(\lg n)$

# Matrix Multiplication

**Input** :  $A = [a_{ij}]$ ,  $B = [b_{ij}]$   
**Output**:  $C = [c_{ij}] = A \cdot B$  }  $i, j = 1, 2, \dots, n$

$$\begin{pmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \dots & & & \\ c_{n1} & c_{n2} & \dots & c_{nn} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} \cdot \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \dots & & & \\ b_{n1} & b_{n2} & \dots & b_{nn} \end{pmatrix}$$

$$c_{ij} = \sum_{1 \leq k \leq n} a_{ik} \cdot b_{kj}$$

# Standard Algorithm

for  $i \leftarrow 1$  to  $n$

  for  $j \leftarrow 1$  to  $n$

$c_{ij} \leftarrow 0$

    for  $k \leftarrow 1$  to  $n$

$c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}$

Running time =  $\Theta(n^3)$

# Matrix Multiplication: Divide & Conquer

IDEA: Divide the  $n \times n$  matrix into

$2 \times 2$  matrix of  $(n/2) \times (n/2)$  submatrices

$$\begin{array}{c} C \\ \left( \begin{array}{c|c} c_{11} & c_{12} \\ \hline c_{21} & c_{22} \end{array} \right) = \begin{array}{c} A \\ \left( \begin{array}{c|c} a_{11} & a_{12} \\ \hline a_{21} & a_{22} \end{array} \right) \cdot \begin{array}{c} B \\ \left( \begin{array}{c|c} b_{11} & b_{12} \\ \hline b_{21} & b_{22} \end{array} \right) \end{array} \end{array}$$

$$c_{11} = a_{11} b_{11} + a_{12} b_{21}$$



# Matrix Multiplication: Divide & Conquer

IDEA: Divide the  $n \times n$  matrix into

$2 \times 2$  matrix of  $(n/2) \times (n/2)$  submatrices

$$\begin{array}{c} C \\ \left( \begin{array}{c|c} c_{11} & c_{12} \\ \hline c_{21} & c_{22} \end{array} \right) = \begin{array}{c} A \\ \left( \begin{array}{c|c} a_{11} & a_{12} \\ \hline a_{21} & a_{22} \end{array} \right) \cdot \begin{array}{c} B \\ \left( \begin{array}{c|c} b_{11} & b_{12} \\ \hline b_{21} & b_{22} \end{array} \right) \end{array} \end{array}$$

$$c_{12} = a_{11} b_{12} + a_{12} b_{22}$$

# Matrix Multiplication: Divide & Conquer

IDEA: Divide the  $n \times n$  matrix into

$2 \times 2$  matrix of  $(n/2) \times (n/2)$  submatrices

$$\begin{array}{c} C \\ \left( \begin{array}{c|c} c_{11} & c_{12} \\ \hline c_{21} & c_{22} \end{array} \right) = \begin{array}{c} A \\ \left( \begin{array}{c|c} a_{11} & a_{12} \\ \hline a_{21} & a_{22} \end{array} \right) \cdot \begin{array}{c} B \\ \left( \begin{array}{c|c} b_{11} & b_{12} \\ \hline b_{21} & b_{22} \end{array} \right) \end{array} \end{array}$$

$$c_{21} = a_{21} b_{11} + a_{22} b_{21}$$

# Matrix Multiplication: Divide & Conquer

IDEA: Divide the  $n \times n$  matrix into

$2 \times 2$  matrix of  $(n/2) \times (n/2)$  submatrices

$$\begin{array}{c} C \\ \left( \begin{array}{c|c} c_{11} & c_{12} \\ \hline c_{21} & c_{22} \end{array} \right) = \begin{array}{c} A \\ \left( \begin{array}{c|c} a_{11} & a_{12} \\ \hline a_{21} & a_{22} \end{array} \right) \cdot \begin{array}{c} B \\ \left( \begin{array}{c|c} b_{11} & b_{12} \\ \hline b_{21} & b_{22} \end{array} \right) \end{array} \end{array}$$

$$c_{22} = a_{21} b_{12} + a_{22} b_{22}$$

# Matrix Multiplication: Divide & Conquer

$$\begin{array}{c} \mathbf{C} \\ \left( \begin{array}{c|c} c_{11} & c_{12} \\ \hline c_{21} & c_{22} \end{array} \right) \end{array} = \begin{array}{c} \mathbf{A} \\ \left( \begin{array}{c|c} a_{11} & a_{12} \\ \hline a_{21} & a_{22} \end{array} \right) \end{array} \cdot \begin{array}{c} \mathbf{B} \\ \left( \begin{array}{c|c} b_{11} & b_{12} \\ \hline b_{21} & b_{22} \end{array} \right) \end{array}$$

$$c_{11} = a_{11} b_{11} + a_{12} b_{21}$$

$$c_{12} = a_{11} b_{12} + a_{12} b_{22}$$

$$c_{21} = a_{21} b_{11} + a_{22} b_{21}$$

$$c_{22} = a_{21} b_{12} + a_{22} b_{22}$$

8 mults of  $(n/2) \times (n/2)$  submatrices

4 adds of  $(n/2) \times (n/2)$  submatrices

# Matrix Multiplication: Divide & Conquer

## MATRIX-MULTIPLY (A, B)

*// Assuming that both A and B are nxn matrices*

**if**  $n = 1$  **then return**  $A * B$

**else**

partition  $A$ ,  $B$ , and  $C$  as shown before

$$c_{11} = \text{MATRIX-MULTIPLY}(a_{11}, b_{11}) + \text{MATRIX-MULTIPLY}(a_{12}, b_{21})$$

$$c_{12} = \text{MATRIX-MULTIPLY}(a_{11}, b_{12}) + \text{MATRIX-MULTIPLY}(a_{12}, b_{22})$$

$$c_{21} = \text{MATRIX-MULTIPLY}(a_{21}, b_{11}) + \text{MATRIX-MULTIPLY}(a_{22}, b_{21})$$

$$c_{22} = \text{MATRIX-MULTIPLY}(a_{21}, b_{12}) + \text{MATRIX-MULTIPLY}(a_{22}, b_{22})$$

**return**  $C$

# Matrix Multiplication: Divide & Conquer Analysis

$$T(n) = 8 T(n/2) + \Theta(n^2)$$

8 recursive calls

each subproblem  
has size  $n/2$

submatrix  
addition

# Matrix Multiplication: Solving the Recurrence

$$T(n) = 8 T(n/2) + \Theta(n^2)$$

→  $a = 8, \quad b = 2, \quad f(n) = \Theta(n^2), \quad n^{\log_b a} = n^3$

Case 1:

$$\frac{n^{\log_b a}}{f(n)} = \Omega(n^\epsilon)$$



$$T(n) = \Theta(n^{\log_b a})$$

→  $T(n) = \Theta(n^3)$

*No better than the ordinary algorithm!*

# Matrix Multiplication: Strassen's Idea

$$\begin{array}{c} \mathbf{C} \\ \left( \begin{array}{c|c} c_{11} & c_{12} \\ \hline c_{21} & c_{22} \end{array} \right) \end{array} = \begin{array}{c} \mathbf{A} \\ \left( \begin{array}{c|c} a_{11} & a_{12} \\ \hline a_{21} & a_{22} \end{array} \right) \end{array} \cdot \begin{array}{c} \mathbf{B} \\ \left( \begin{array}{c|c} b_{11} & b_{12} \\ \hline b_{21} & b_{22} \end{array} \right) \end{array}$$

Compute  $c_{11}$ ,  $c_{12}$ ,  $c_{21}$ , and  $c_{22}$  using **7** recursive multiplications



# Matrix Multiplication: Strassen's Idea

$$P_1 = a_{11} \times (b_{12} - b_{22})$$

$$P_2 = (a_{11} + a_{12}) \times b_{22}$$

$$P_3 = (a_{21} + a_{22}) \times b_{11}$$

$$P_4 = a_{22} \times (b_{21} - b_{11})$$

$$P_5 = (a_{11} + a_{22}) \times (b_{11} + b_{22})$$

$$P_6 = (a_{12} - a_{22}) \times (b_{21} + b_{22})$$

$$P_7 = (a_{11} - a_{21}) \times (b_{11} + b_{12})$$

Reminder: Each submatrix is of size  $(n/2) \times (n/2)$

Each add/sub operation takes  $\Theta(n^2)$  time

Compute  $P_1 \dots P_7$  using **7** recursive calls to matrix-multiply

*How to compute  $c_{ij}$  using  $P_1 \dots P_7$ ?*

# Matrix Multiplication: Strassen's Idea

$$P_1 = a_{11} \times (b_{12} - b_{22})$$

$$P_2 = (a_{11} + a_{12}) \times b_{22}$$

$$P_3 = (a_{21} + a_{22}) \times b_{11}$$

$$P_4 = a_{22} \times (b_{21} - b_{11})$$

$$P_5 = (a_{11} + a_{22}) \times (b_{11} + b_{22})$$

$$P_6 = (a_{12} - a_{22}) \times (b_{21} + b_{22})$$

$$P_7 = (a_{11} - a_{21}) \times (b_{11} + b_{12})$$

$$\begin{aligned}c_{11} &= P_5 + P_4 - P_2 + P_6 \\c_{12} &= P_1 + P_2 \\c_{21} &= P_3 + P_4 \\c_{22} &= P_5 + P_1 - P_3 - P_7\end{aligned}$$

7 recursive multiply calls

18 add/sub operations

*Does not rely on commutativity of multiplication*

# Matrix Multiplication: Strassen's Idea

$$P_1 = a_{11} \times (b_{12} - b_{22})$$

$$P_2 = (a_{11} + a_{12}) \times b_{22}$$

$$P_3 = (a_{21} + a_{22}) \times b_{11}$$

$$P_4 = a_{22} \times (b_{21} - b_{11})$$

$$P_5 = (a_{11} + a_{22}) \times (b_{11} + b_{22})$$

$$P_6 = (a_{12} - a_{22}) \times (b_{21} + b_{22})$$

$$P_7 = (a_{11} - a_{21}) \times (b_{11} + b_{12})$$

e.g. Show that  $c_{12} = P_1 + P_2$

$$\begin{aligned} c_{12} &= P_1 + P_2 \\ &= a_{11}(b_{12} - b_{22}) + (a_{11} + a_{12})b_{22} \\ &= a_{11}b_{12} - a_{11}b_{22} + a_{11}b_{22} + a_{12}b_{22} \\ &= a_{11}b_{12} + a_{12}b_{22} \end{aligned}$$

# Strassen's Algorithm

- 1. Divide:** Partition **A** and **B** into  $(n/2) \times (n/2)$  submatrices. Form terms to be multiplied using  $+$  and  $-$ .
- 2. Conquer:** Perform **7** multiplications of  $(n/2) \times (n/2)$  submatrices recursively.
- 3. Combine:** Form **C** using  $+$  and  $-$  on  $(n/2) \times (n/2)$  submatrices.

**Recurrence:**  $T(n) = 7 T(n/2) + \Theta(n^2)$

# Strassen's Algorithm: Solving the Recurrence

$$T(n) = 7 T(n/2) + \Theta(n^2)$$

→  $a = 7, \quad b = 2, \quad f(n) = \Theta(n^2), \quad n^{\log_b a} = n^{\lg 7}$

Case 1:

$$\frac{n^{\log_b a}}{f(n)} = \Omega(n^\epsilon)$$



$$T(n) = \Theta(n^{\log_b a})$$

→  $T(n) = \Theta(n^{\lg 7})$

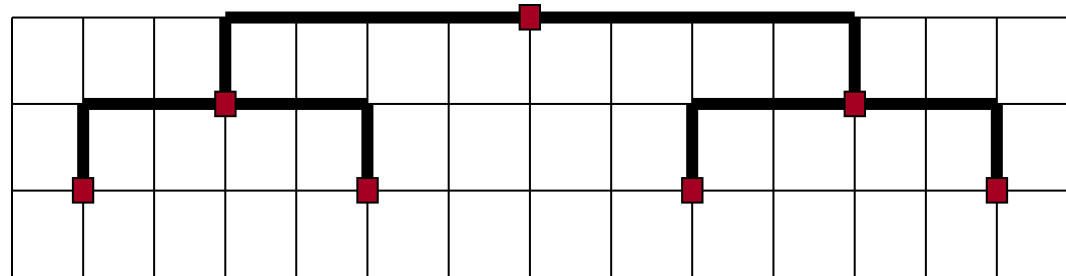
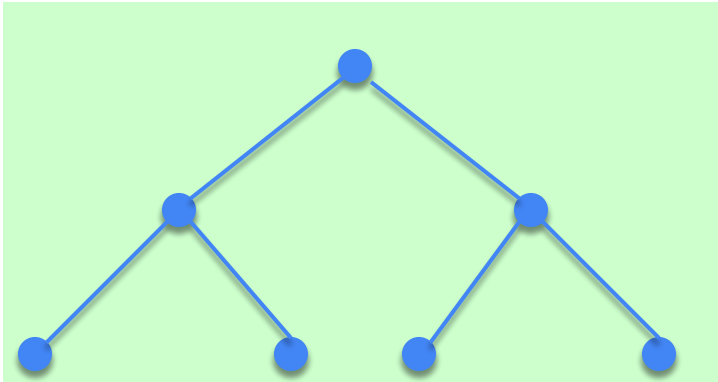
Note:  $\lg 7 \approx 2.81$

# Strassen's Algorithm

- The number **2.81** may not seem much smaller than **3**
- But, it is significant because the difference is in the exponent.
- Strassen's algorithm **beats** the ordinary algorithm on today's machines for  **$n \geq 30$**  or so.
- **Best to date:  $\Theta(n^{2.376\dots})$**  (*of theoretical interest only*)

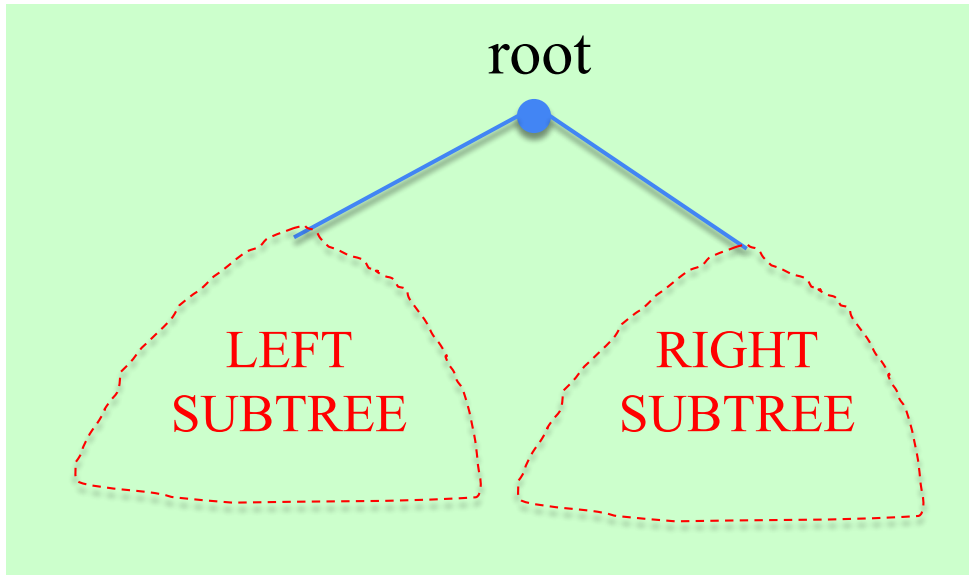
# VLSI Layout: Binary Tree Embedding

- **Problem:** Embed a complete binary tree with  $n$  leaves into a 2D grid with minimum area.
- **Example:**



# Binary Tree Embedding

- Use divide and conquer

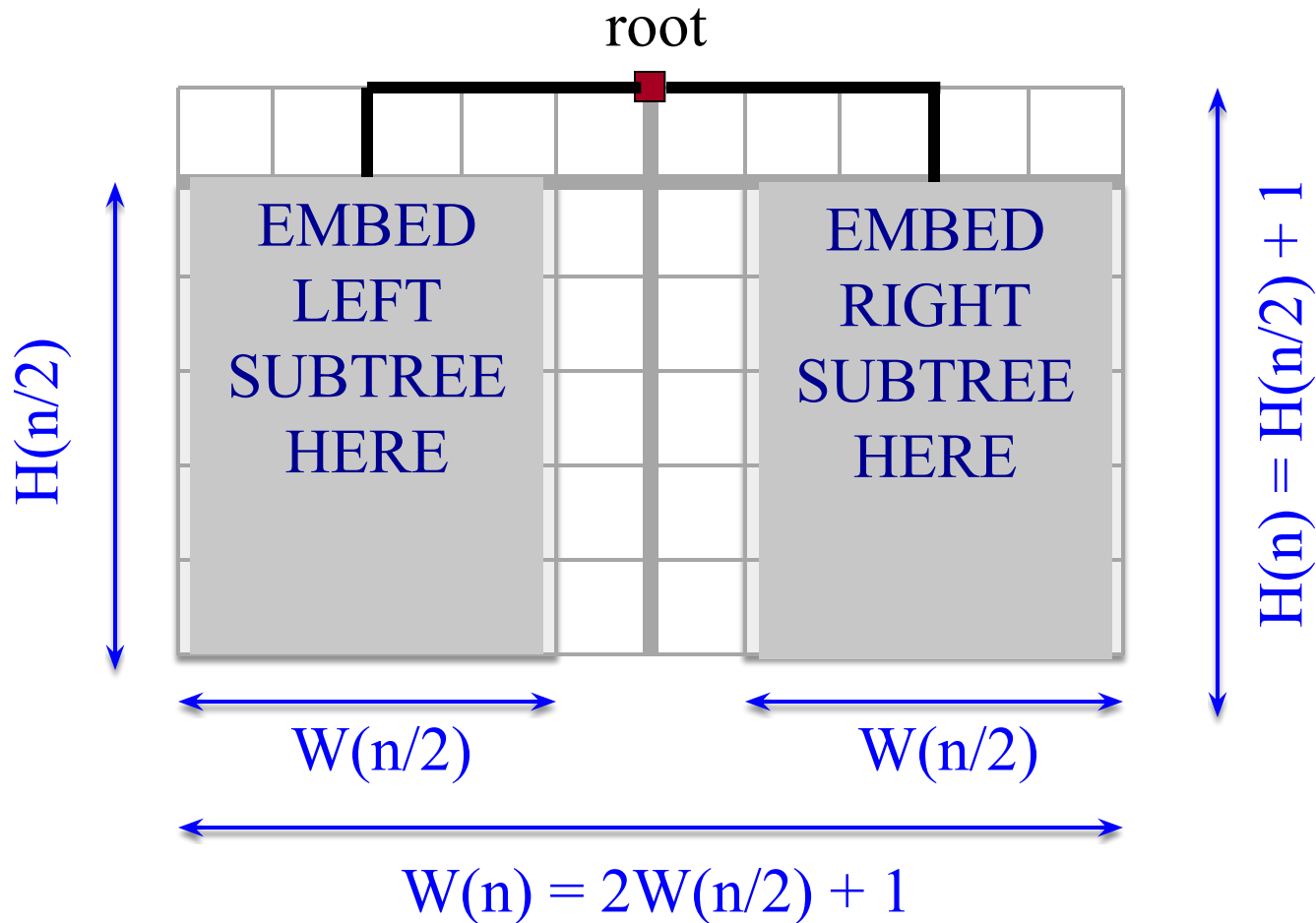


1. Embed the root node
2. Embed the left subtree
3. Embed the right subtree

What is the min-area required for  $n$  leaves?



# Binary Tree Embedding



# Binary Tree Embedding

- Solve the recurrences:

$$W(n) = 2W(n/2) + 1$$

$$H(n) = H(n/2) + 1$$

- $W(n) = \Theta(n)$

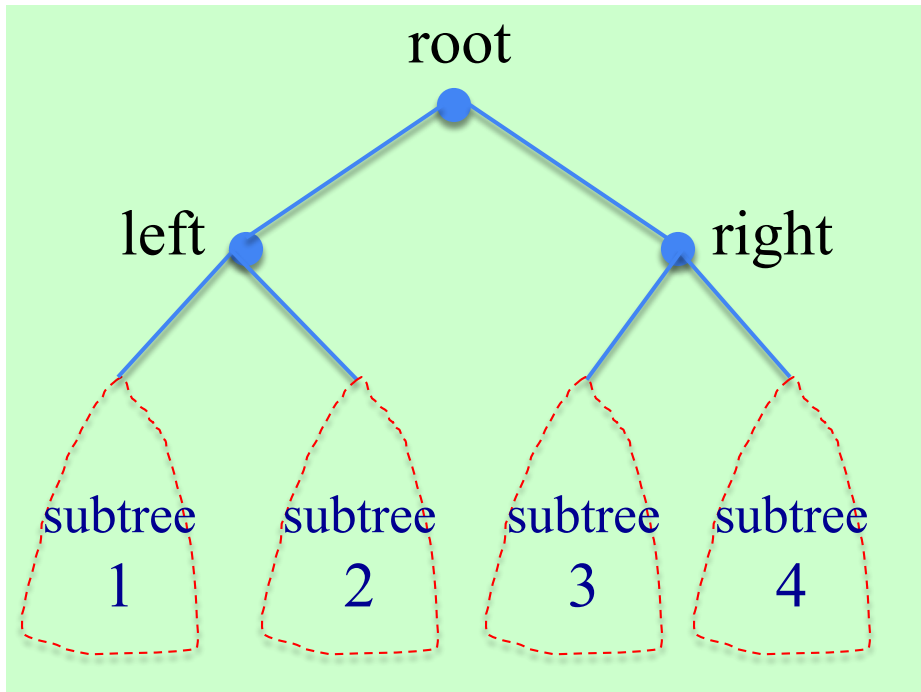
- $H(n) = \Theta(\lg n)$

- $\text{Area}(n) = \Theta(n \lg n)$



# Binary Tree Embedding: H-Tree

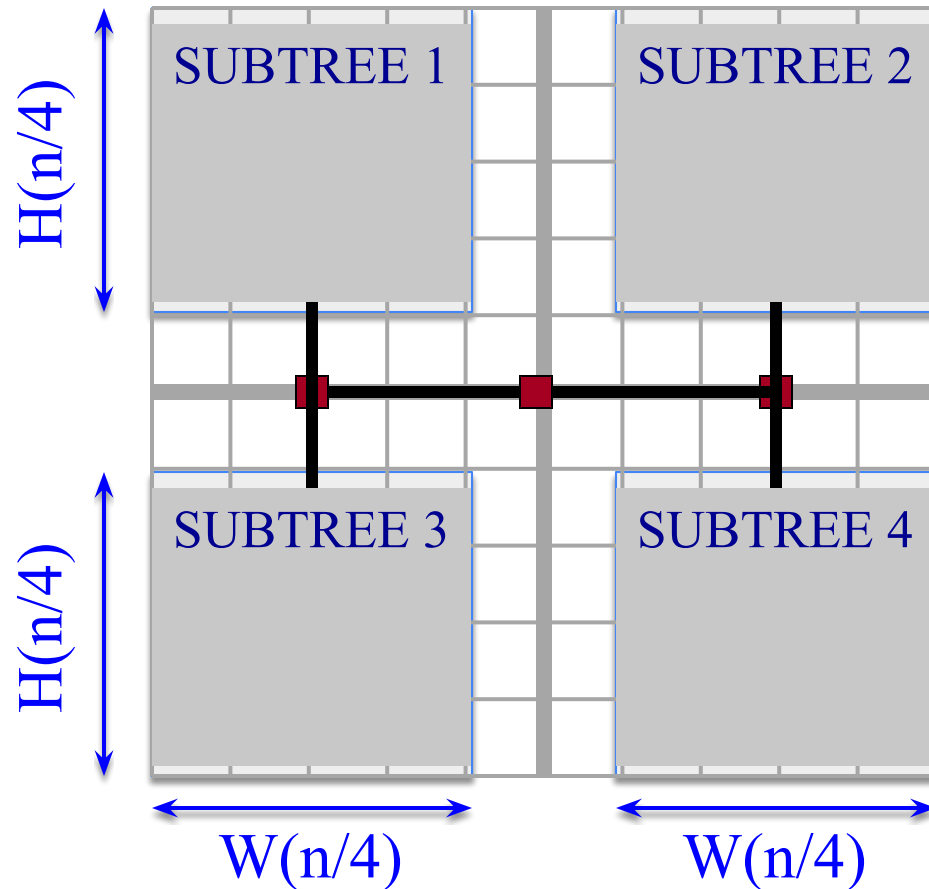
- Use a different divide and conquer method



1. Embed root, left, right nodes
2. Embed subtree 1
3. Embed subtree 2
4. Embed subtree 3
5. Embed subtree 4

*What is the min-area required for  $n$  leaves?*

# Binary Tree Embedding: H-Tree



$$W(n) = 2W(n/4) + 1$$

$$H(n) = 2H(n/4) + 1$$

# Binary Tree Embedding: H-Tree

- Solve the recurrences:

$$W(n) = 2W(n/4) + 1$$

$$H(n) = 2H(n/4) + 1$$

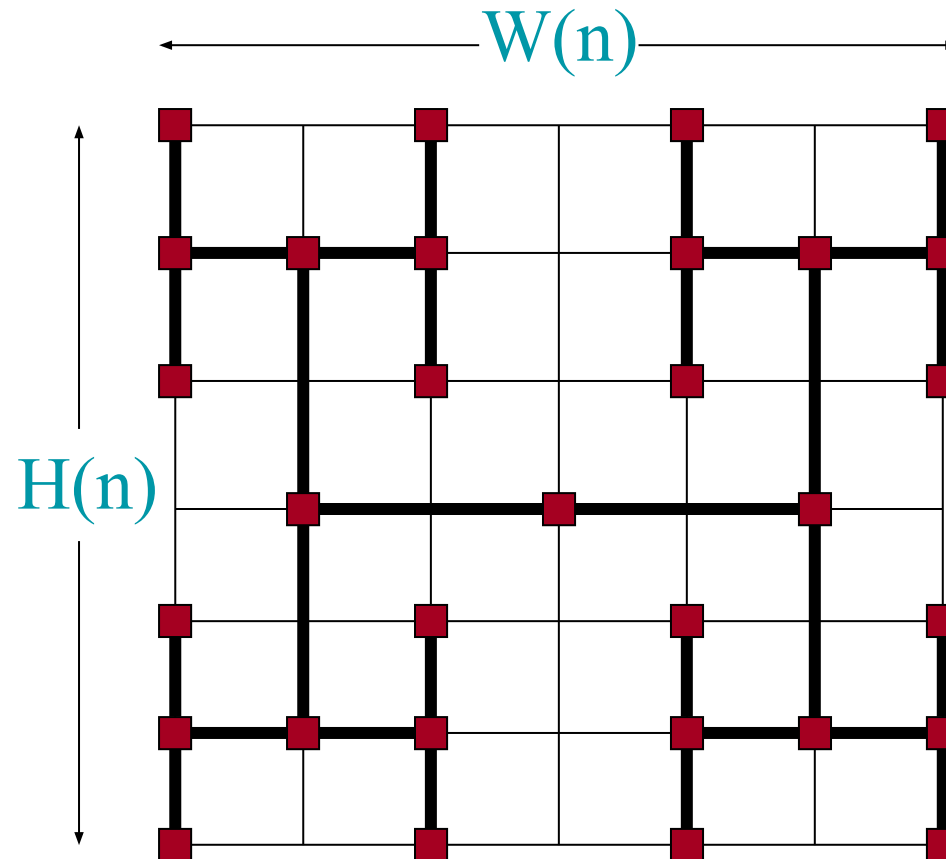
- $W(n) = \Theta(\sqrt{n})$

- $H(n) = \Theta(\sqrt{n})$

- $Area(n) = \Theta(n)$

# Binary Tree Embedding: H-Tree

Example:



# Correctness Proofs

- ***Proof by induction*** commonly used for D&C algorithms
- **Base case**: Show that the algorithm is correct when the recursion bottoms out (i.e., for sufficiently small  $n$ )
- **Inductive hypothesis**: Assume the alg. is correct for any recursive call on any smaller subproblem of size  $k$  ( $k < n$ )
- **General case**: Based on the inductive hypothesis, prove that the alg. is correct for any input of size  $n$



# Example Correctness Proof: Powering a Number

POWER (a, n)

**if** n = 0 **then return** 1

**else if** n is even **then**

val ← POWER (a, n/2)

**return** val \* val

**else if** n is odd **then**

val ← POWER (a, (n-1)/2)

**return** val \* val \* a

# Example Correctness Proof: Powering a Number

- Base case:  $\text{POWER}(a, 0)$  is correct, because it returns 1
- Ind. hyp: Assume  $\text{POWER}(a, k)$  is correct for any  $k < n$
- General case:

In  $\text{POWER}(a, n)$  function:

If  $n$  is even:

$$\text{val} = a^{n/2} \text{ (due to ind. hyp.)}$$

$$\text{it returns } \text{val} \cdot \text{val} = a^n$$

If  $n$  is odd:

$$\text{val} = a^{(n-1)/2} \text{ (due to ind. hyp.)}$$

$$\text{it returns } \text{val} \cdot \text{val} \cdot a = a^n$$

□ *The correctness proof is complete*

# Maximum Subarray Problem

- Input: An array of values
- Output: The contiguous subarray that has the largest sum of elements

Input array:

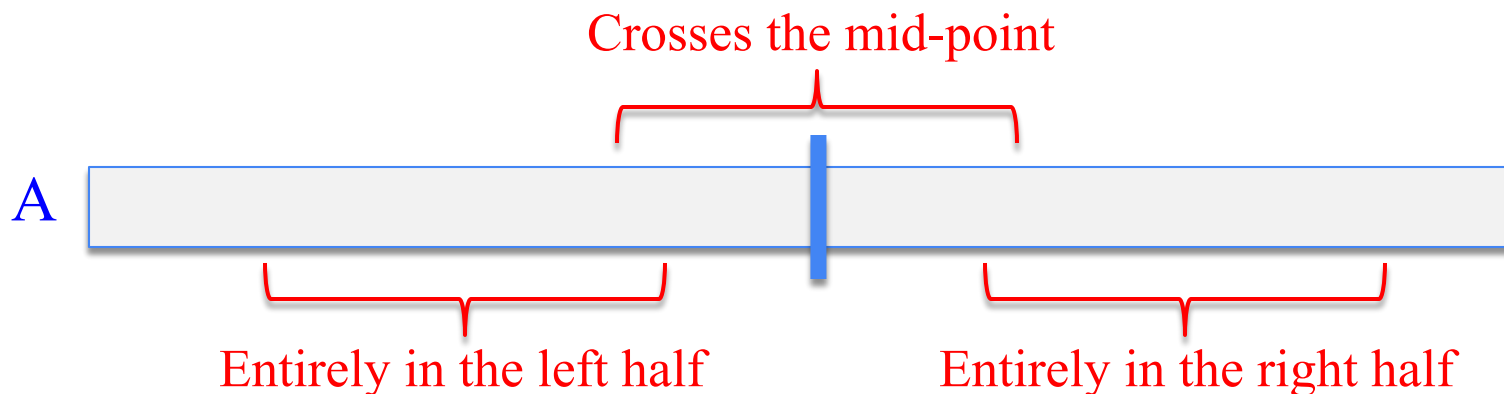
13	-3	-25	20	-3	-16	-23	18	20	-7	12	-22	-4	7
----	----	-----	----	----	-----	-----	----	----	----	----	-----	----	---



the maximum contiguous subarray

# Maximum Subarray Problem: Divide & Conquer

- Basic idea:
  - **Divide** the input array into 2 from the middle
  - Pick the **best** solution among the following:
    1. The max subarray of the **left half**
    2. The max subarray of the **right half**
    3. The max subarray **crossing the mid-point**



# Maximum Subarray Problem: Divide & Conquer

- Divide: Trivial (divide the array from the middle)
- Conquer: Recursively compute the max subarrays of the **left** and **right** halves
- Combine: Compute the max-subarray crossing the mid-point (*can be done in  $\Theta(n)$  time*). Return the max among the following:
  1. the max subarray of the **left subarray**
  2. the max subarray of the **right subarray**
  3. the max subarray **crossing the mid-point**

See textbook for the detailed solution.

# Conclusion

- Divide and conquer is just one of several powerful techniques for algorithm design.
- Divide-and-conquer algorithms can be analyzed using recurrences and the master method (so practice this math).
- Can lead to more efficient algorithms