

# CS483 Design and Analysis of Algorithms

## Chapter 2 Divide and Conquer Algorithms

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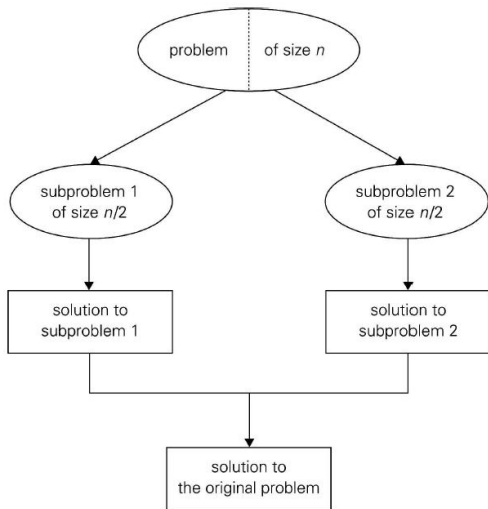
Course web-site:

[http://www.cs.gmu.edu/~lifei/teaching/cs483\\_fall09](http://www.cs.gmu.edu/~lifei/teaching/cs483_fall09)

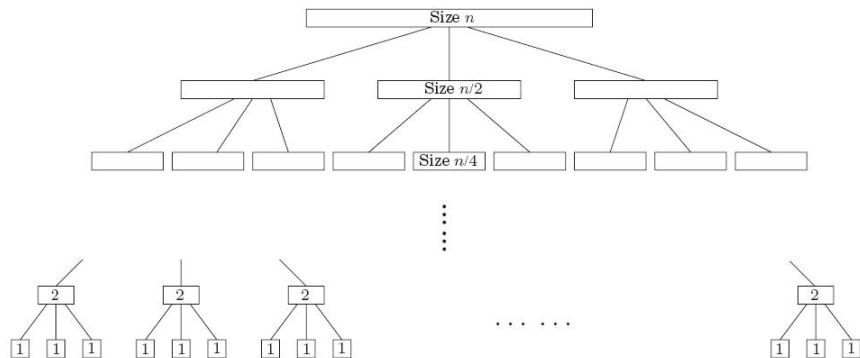
Figures unclaimed are from books “Algorithms” and “Introduction  
to Algorithms”

# Divide and Conquer Algorithms

- 1 Breaking the problem into subproblems of the same type
- 2 Recursively solving these subproblems
- 3 Appropriately combining their answers



# Divide-and-Conquer Recurrence



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$$T(n) = a \cdot T(n/b) + f(n)$$

- **1 The iteration method**

Expand (iterate) the recurrence and express it as a summation of terms depending only on  $n$  and the initial conditions

- **2 The substitution method**

- Guess the form of the solution
- Use mathematical induction to find the constants

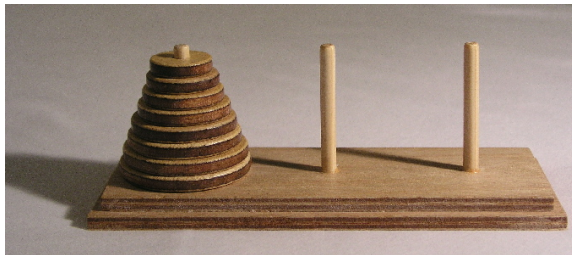
- **3 Master Theorem** ( $T(n) = a \cdot T(n/b) + f(n)$ )

# Iteration Method — Examples

- $n!$

$$T(n) = T(n - 1) + 1$$

- **Tower of Hanoi**



[http://en.wikipedia.org/wiki/Tower\\_of\\_Hanoi](http://en.wikipedia.org/wiki/Tower_of_Hanoi)

$$T(n) = 2 \cdot T(n - 1) + 1$$

## Iteration — Example

- $n!$  ( $T(n) = T(n-1) + 1$ )

$$\begin{aligned}T(n) &= T(n-1) + 1 \\ &= (T(n-2) + 1) + 1 \\ &= T(n-2) + 2 \\ &\dots \quad \dots \\ &= T(n-i) + i \\ &\dots \quad \dots \\ &= T(0) + n = n\end{aligned}$$

- **Tower of Hanoi** ( $T(n) = 2 \cdot T(n-1) + 1$ ) ?



## Iteration — Example

**Tower of Hanoi** ( $T(n) = 2 \cdot T(n-1) + 1$ )

$$\begin{aligned}T(n) &= 2 \cdot T(n-1) + 1 \\ &= 2 \cdot (2 \cdot T(n-2) + 1) + 1 \\ &= 2^2 \cdot T(n-2) + 2 + 1 \\ &\dots \quad \dots \\ &= 2^i \cdot T(n-i) + 2^{i-1} + \dots + 1 \\ &\dots \quad \dots \\ &= 2^{n-1} \cdot T(1) + 2^{n-1} + 2^{n-1} + \dots + 1 \\ &= 2^{n-1} \cdot T(1) + \sum_{i=0}^{n-2} 2^i \\ &= 2^{n-1} + 2^{n-1} - 1 \\ &= 2^n - 1\end{aligned}$$

## Substitution Method — Count Number of Bits

- **Count number of bits** ( $T(n) = T(\lfloor n/2 \rfloor) + 1$ )

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- Guess  $T(n) \leq \log n$ .

$$\begin{aligned}T(n) &= T(\lfloor n/2 \rfloor) + 1 \\ &\leq \log(\lfloor n/2 \rfloor) + 1 \\ &\leq \log(n/2) + 1 \\ &\leq (\log n - \log 2) + 1 \\ &\leq \log n - 1 + 1 \\ &= \log n\end{aligned}$$

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- Guess  $T(n) \leq 2^n - 1$

$$\begin{aligned}T(n) &= 2 \cdot T(n-1) + 1 \\ &\leq 2 \cdot (2^{n-1} - 1) + 1 \\ &= 2^n - 2 + 1 \\ &= 2^n - 1, \quad \text{correct!}\end{aligned}$$

## Substitution Method — Extension $F_n$

- **Fibonacci Numbers** ( $F_0 = 0, F_1 = 1, F_n = F_{n-1} + F_{n-2}$ )

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Guess  $F_n = c \cdot \phi^n, 1 < \phi < 2$

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$$\phi^2 = \phi + 1$$

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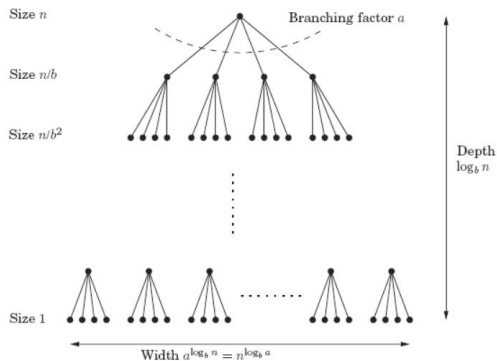
- **General solution:**  $F_n = c_1 \cdot \phi_1^n + c_2 \cdot \phi_2^n$   
 $F_1 = 0, F_2 = 1$

$$F_n = \frac{1}{\sqrt{5}} \left( \frac{1 + \sqrt{5}}{2} \right)^n - \frac{1}{\sqrt{5}} \left( \frac{1 - \sqrt{5}}{2} \right)^n$$

# Master Theorem

$$T(n) = a \cdot T(n/b) + f(n)$$

**Figure 2.3** Each problem of size  $n$  is divided into  $a$  subproblems of size  $n/b$ .



# Master Theorem

$$T(n) = a \cdot T(n/b) + f(n), \quad a \geq 1, b > 1 \text{ be constants.}$$

## Theorem

We interpret  $n/b$  to mean either  $\lceil n/b \rceil$  or  $\lfloor n/b \rfloor$ .

If  $f(n) \in \Theta(n^d)$ , where  $d \geq 0$ , then

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

$$T(n) = \begin{cases} \Theta(f(n)) & \text{if } f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ and if } a \cdot f(n/b) \leq c \cdot f(n) \\ & \text{for some constant } c < 1 \text{ and all sufficiently large } n \\ \Theta(n^{\log_b a} \cdot \log n) & \text{if } f(n) = \Theta(n^{\log_b a}) \\ \Theta(n^{\log_b a}) & \text{if } f(n) = O(n^{\log_b a - \epsilon}) \text{ for some constant } \epsilon > 0 \end{cases}$$

- 1  $T(n) = 4 \cdot T(n/2) + n = ?$
- 2  $T(n) = 4 \cdot T(n/2) + n^2 = ?$
- 3  $T(n) = 4 \cdot T(n/2) + n^3 = ?$

# Chapter 2 of DPV — Divide and Conquer Algorithms

- 1 Mergesort
- 2 Medians
- 3 Matrix Multiplication

# Mergesort

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- `function mergesort(A[n])`

```
if (n = 1)
    return A;
else
    B = A[1, ..., [  $\frac{n}{2}$  ]];
    C = A[[  $\frac{n}{2}$  ] + 1, ..., n];

    mergesort(B);
    mergesort(C);
    merge(B, C, A);
```

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- Is this algorithm **complete**?

# Mergesort

Merge two sorted arrays,  $B$  and  $C$  and put the result in  $A$

```
function merge(B[1, .. p], C[1, .. q], A[1, .. p + q])
```

```
    i = 1; j = 1;
```

```
    for (k = 1, 2, .. p + 1)
```

```
        if (B[i] < C[j])
```

```
            A[k] = B[i];
```

```
            i = i + 1;
```

```
        else
```

```
            A[k] = C[j];
```

```
            j = j + 1;
```

24, 11, 91, 10, 22, 32, 22, 3, 7, 99

$$T(n) = 2 \cdot T(n/2) + O(n) = O(n \cdot \log n)$$

# Medians

The *median* is a single representative value of a list of numbers: half of them are larger and half of them are smaller; less sensitive to outliers

$$S := \{2, 36, 5, 21, 8, 13, 11, 20, 5, 4, 1\}$$

**Selection problem:**

- **Input:** A list of number  $S$ ; an integer  $k$ .
- **Output:** The  $k$ -th smallest element of  $S$ .

$$\text{selection}(S, 8) = ?$$

# Medians

The *median* is a single representative value of a list of numbers: half of them are larger and half of them are smaller; less sensitive to outliers

$$S := \{2, 36, 5, 21, 8, 13, 11, 20, 5, 4, 1\}$$

Let us split at  $v = 5$

$$S_L = \{2, 4, 1\}$$

$$S_v = \{5, 5\}$$

$$S_R = \{36, 21, 8, 13, 11, 20\}$$

$$\text{selection}(S, 8) = \text{selection}(S_R, 3).$$

$$\text{selection}(S, k) = \begin{cases} \text{selection}(S_L, k) & \text{if } k \leq |S_L| \\ v & \text{if } |S_L| < k \leq |S_L| + |S_v| \\ \text{selection}(S_R, k - |S_L| - |S_v|) & \text{if } k > |S_L| + |S_v|. \end{cases}$$

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If  $|S_L| \approx |S_R|$  (i.e., pick up  $v$  to be the median),

$$T(n) = T(n/2) + O(n)$$

**Pick up  $v$  randomly from  $S$**

# Medians

$v$  is *good* if it lies within 25% and 75% of the array it is chosen

## Lemma

*On average a fair coin needs to be tossed two times before a "heads" is seen*

## Proof.

$$E = 1 + \frac{1}{2} \cdot E$$



## Remark

$v$  has 50% chance of being *in-between* [25%, 75%]. We need to pick  $v$  twice to make it good

## Theorem

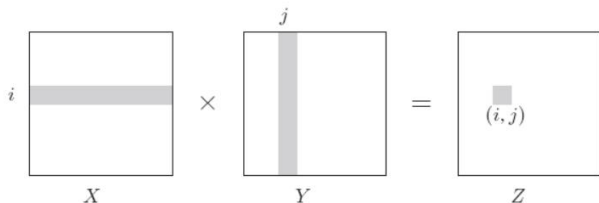
$$T(n) \leq T((3/4) \cdot n) + O(n) = O(n)$$

# Matrix Multiplication

The product of two  $n \times n$  matrices  $X$  and  $Y$  is a third  $n \times n$  matrix  $Z = XY$ , with  $(i, j)$ th entry

$$Z_{ij} = \sum_{k=1}^n X_{ik} Y_{kj}.$$

To make it more visual,  $Z_{ij}$  is the dot product of the  $i$ th row of  $X$  with the  $j$ th column of  $Y$ :





# Matrix Multiplication

1 Matrix Multiplication (by definition):

$$\begin{aligned} \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} &= \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \\ &= \begin{bmatrix} A_{11} \cdot B_{11} + A_{12} \cdot B_{21} & A_{11} \cdot B_{12} + A_{12} \cdot B_{22} \\ A_{21} \cdot B_{11} + A_{22} \cdot B_{21} & A_{21} \cdot B_{12} + A_{22} \cdot B_{22} \end{bmatrix} \end{aligned}$$

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2

$$T(n) = 8 \cdot T\left(\frac{n}{2}\right) + O(n) = O(n^3)$$

Time complexity of the brute-force algorithm is  $O(n^3)$

# Matrix Multiplication

## 1 Strassen's Matrix Multiplication:

$$\begin{aligned} \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} &= \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \\ &= \begin{bmatrix} m_1 + m_4 - m_5 + m_7 & m_3 + m_5 \\ m_2 + m_4 & m_1 + m_3 - m_2 + m_6 \end{bmatrix} \end{aligned}$$

- $m_1 = (A_{11} + A_{22}) \cdot (B_{11} + B_{22})$
- $m_2 = (A_{21} + A_{22}) \cdot B_{11}$
- $m_3 = A_{11} \cdot (B_{12} - B_{22})$
- $m_4 = A_{22} \cdot (B_{21} - B_{11})$
- $m_5 = (A_{11} + A_{12}) \cdot B_{22}$
- $m_6 = (A_{21} - A_{11}) \cdot (B_{11} + B_{12})$
- $m_7 = (A_{12} - A_{22}) \cdot (B_{21} + B_{22})$

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- $m_2 = (A_{21} + A_{22}) \cdot B_{11}$
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- $m_4 = A_{22} \cdot (B_{21} - B_{11})$
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- $m_7 = (A_{12} - A_{22}) \cdot (B_{21} + B_{22})$

## 2

$$T(n) = 7 \cdot T\left(\frac{n}{2}\right) + O(n) = O(\log n^{\log_2 7}) \approx O(n^{2.81})$$

Time complexity of the brute-force algorithm is  $O(n^3)$