

# Divide and Conquer methods (and Branch and Bound) in a nutshell

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August 3, 2011

# Introduction

Divide and conquer works by:

1. Breaking it into subproblems, that are themselves smaller instances of the same type of problem
2. Recursively solving these subproblems
3. Appropriately combining their answers

(from Dasgupta, Papadimitriou and Vazirani)

# Analyzing Divide and Conquer Methods

- ▶ Analyzing Divide and Conquer methods relies on analyzing recurrence relations, e.g.

$$T(n) = 3T(n/2) + O(n) \quad (1)$$

- ▶ The resulting complexity can be described using the following theorem:

## Theorem

If  $T(n) = aT(\lceil n/b \rceil) + O(n^d)$  for some constants  $a > 0$ ,  $b > 1$ , and  $d \geq 0$ , then

$$T(n) = \begin{cases} O(n^d) & \text{if } d > \log_b a \\ O(n^d \log n) & \text{if } d = \log_b a \\ O(n^{\log_b a}) & \text{if } d < \log_b a \end{cases} \quad (2)$$

# Merge Sort

- ▶ Merge sort works by
  1. Dividing the list in two
  2. Calling merge sort on each half
  3. Merging the sorted lists
- ▶ This has the recursion relation:

$$T(n) = 2T(\lceil n/2 \rceil) + O(n) \quad (3)$$

because there are two calls to merge-sort,  $2T(\lceil n/2 \rceil)$ , and the merge process takes  $O(n)$

- ▶ From the Theorem, this gives a complexity of  $O(n \log(n))$  which is very good.

# Matrix Multiplication

- ▶ Matrix multiplication was generally believed to be  $O(n^3)$ ! For  $Z = XY$

$$Z_{ik} = \sum_j X_{ij} Y_{jk} \quad (4)$$

- ▶ Strassen discovered a more efficient divide and conquer approach. Instead of:

$$XY = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} E & F \\ G & H \end{bmatrix} = \begin{bmatrix} AE + BG & AF + BH \\ CE + DG & CF + DH \end{bmatrix} \quad (5)$$

- ▶ Which has recurrence relation  $T(n) = 8T(n/2) + O(n)^2$  and complexity  $O(n^3)$

## Matrix Multiplication, cont'd

- ▶ The multiplication was expressed as:

$$XY = \begin{bmatrix} P_5 + P_4 - P_2 + P_6 & P_1 + P_2 \\ P_3 + P_4 & P_1 + P_5 - P_3 - P_7 \end{bmatrix} \quad (6)$$

where:

$$\begin{aligned} P_1 &= A(F - H) & P_2 &= (A + B)H & P_3 &= (C + D)E \\ P_4 &= D(G - E) & P_5 &= (A + D)(E + H) & P_6 &= (B - D)(G + H) \\ & & & & P_7 &= (A - C)(E + F) \end{aligned} \quad (7)$$

- ▶ The complexity is  $T(n) = 7T(n/2) + O(n^2)$ , which is

$$O(n^{\log_2 7}) \approx O(n^{2.81}) \quad (8)$$

## Other examples

- ▶ Fast fourier transform
- ▶ Nearest Neighbor search

# Branch and Bound

- ▶ Method to solve integer programming problems
- ▶ Simple example:

$$\begin{array}{ll} \text{minimize} & f(x, y) \\ \text{subject to} & g(x, y) = 0 \\ & x \in \{0, 1\} \\ & y \in \{0, 1\} \end{array} \quad (9)$$

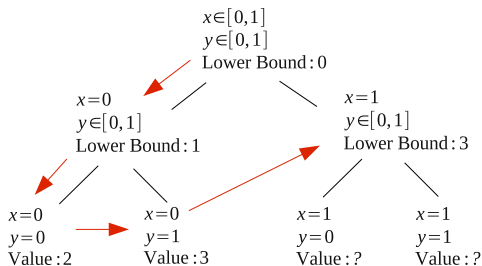
- ▶ Note that the optimal solution to

$$\begin{array}{ll} \text{minimize} & f(x, y) \\ \text{subject to} & g(x, y) = 0 \\ & x \in [0, 1] \\ & y \in \{0, 1\} \end{array} \quad (10)$$

is a lower bound on the original ( $x$  has more choices).



## Branch and Bound, cont'd



- ▶ The lower bounds can be compared with a given feasible solution.
- ▶ This can reduce the number of solutions that need to be checked.