# Divide-and-conquer

(CLRS 4.2)

It's a powerful technique for solving problems:

### Divide-and-Conquer (Input: Problem P)

To Solve P:

- 1. Divide P into smaller problems  $P_1, P_2$
- 2. Conquer by solving the (smaller) subproblems recursively.
- 3. Combine solutions to  $P_1, P_2$  into solution for P.

## **Matrix Multiplication**

Let X and Y be two  $n \times n$  matrices

$$X = \left\{ \begin{array}{cccc} x_{11} & x_{12} & \cdots & x_{1n} \\ x_{21} & x_{22} & \cdots & x_{1n} \\ x_{31} & x_{32} & \cdots & x_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{nn} \end{array} \right\}$$

We want to compute  $Z = X \cdot Y$ , where  $z_{ij} = \sum_{k=1}^{n} X_{ik} \cdot Y_{kj}$ 

Problem: Given two matrices of size n by n, come up with an algorithm to compute the product.

- The straightfoward method uses  $\Rightarrow n^2 \cdot n = \Theta(n^3)$  operations
- Can we do better? That is, is it possible to multiply two matrices faster than  $\Theta(n^3)$ ?
- This was an open problem for a long time... until Strassen came up with an algorithm in 1969. The idea is to use divide-and-conquer.

## Matrix multiplication with divide-and-conquer

• Let's imagine that n is a power of two. We can view each matrix as consisting of 2x2=4 n/2-by-n/2 matrices.

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$$X = \left\{ \begin{array}{cc} A & B \\ C & D \end{array} \right\}, Y = \left\{ \begin{array}{cc} E & F \\ G & H \end{array} \right\}$$

• Then we see that their product  $X \cdot Y$  can be written as:

$$\left\{ \begin{array}{c} A & B \\ C & D \end{array} \right\} \cdot \left\{ \begin{array}{c} E & F \\ G & H \end{array} \right\} = \left\{ \begin{array}{c} (A \cdot E + B \cdot G) & (A \cdot F + B \cdot H) \\ (C \cdot E + D \cdot G) & (C \cdot F + D \cdot H) \end{array} \right\}$$

- The above naturally leads to divide-and-conquer solution:
  - Divide X and Y into 8 sub-matrices A, B, C, D, E, F, G, H.
  - Compute 8 n/2-by-n/2 matrix multiplications recursively.
  - Combine results (by doing 4 matrix additions) and copy the results into Z.
- ANALYSIS: Running time of algorithm is given by  $T(n) = 8T(n/2) + \Theta(n^2) \Rightarrow T(n) = \Theta(n^3)$
- Cool idea, but not so cool result......since we already discussed a (simpler/naive)  $O(n^3)$  algorithm!
- Can we do better?

#### Strassen's divide-and-conquer

• Strassen's algorithm is based on the following observation:

The recurrence

$$T(n) = 8T(n/2) + \Theta(n^2) \Rightarrow T(n) = \Theta(n^3)$$

while the recurrence

$$T(n) = 7T(n/2) + \Theta(n^2) \Rightarrow T(n) = \Theta(n^{\lg 7})$$

- Strassen foud a way to compute only 7 products of n/2-by-n/2 matrices
- With same notation as before, we define the following 7 n/2-by-n/2 matrices:

$$S_1 = (B - D) \cdot (G + H)$$
  
 $S_2 = (A + D) \cdot (E + H)$   
 $S_3 = (A - C) \cdot (E + F)$   
 $S_4 = (A + B) \cdot H$   
 $S_5 = A \cdot (F - H)$   
 $S_6 = D \cdot (G - E)$   
 $S_7 = (C + D) \cdot E$ 

 $\bullet$  Strassen observed that we can write the product Z as:

$$Z = \left\{ \begin{array}{cc} A & B \\ C & D \end{array} \right\} \cdot \left\{ \begin{array}{cc} E & F \\ G & H \end{array} \right\} = \left\{ \begin{array}{cc} (S_1 + S_2 - S_4 + S_6) & (S_4 + S_5) \\ (S_6 + S_7) & (S_2 + S_3 + S_5 - S_7) \end{array} \right\}$$

• For e.g. let's test that  $S_6 + S_7$  is really  $C \cdot E + D \cdot G$ 

$$S_6 + S_7 = D \cdot (G - E) + (C + D) \cdot E$$
$$= DG - DE + CE + DE$$
$$= DG + CE$$

- This leads to a divide-and-conquer algorithm:
  - Divide X and Y into 8 sub-matrices A, B, C, D, E, F, G, H.
  - Compute  $S_1, S_2, S_3, ..., S_7$ . This involves 10 matrix additions and 7 multiplications recursively.
  - Compute  $S_1 + S_2 S_4 + S_6$ , ... and copy them in Z. This step involves only additions/subtractions of n/2-by-n/2 matrices.
- ANALYSIS:  $T(n) = 7T(n/2) + \Theta(n^2)$ , with solution  $O(n^{\lg 7})$ .
- Lets solve the recurrence using the iteration method

$$\begin{split} T(n) &=& 7T(n/2) + n^2 \\ &= n^2 + 7(7T(\frac{n}{2^2}) + (\frac{n}{2})^2) \\ &= n^2 + (\frac{7}{2^2})n^2 + 7^2T(\frac{n}{2^2}) \\ &= n^2 + (\frac{7}{2^2})n^2 + 7^2(7T(\frac{n}{2^3}) + (\frac{n}{2^2})^2) \\ &= n^2 + (\frac{7}{2^2})n^2 + (\frac{7}{2^2})^2 \cdot n^2 + 7^3T(\frac{n}{2^3}) \\ &= n^2 + (\frac{7}{2^2})n^2 + (\frac{7}{2^2})^2n^2 + (\frac{7}{2^2})^3n^2.... + (\frac{7}{2^2})^{\log n - 1}n^2 + 7^{\log n} \\ &= \sum_{i=0}^{\log n - 1} (\frac{7}{2^2})^i n^2 + 7^{\log n} \\ &= n^2 \cdot \Theta((\frac{7}{2^2})^{\log n - 1}) + 7^{\log n} \\ &= n^2 \cdot \Theta(\frac{7^{\log n}}{(2^2)^{\log n}}) + 7^{\log n} \\ &= n^2 \cdot \Theta(\frac{7^{\log n}}{n^2}) + 7^{\log n} \\ &= \Theta(7^{\log n}) \end{split}$$

- Now we have the following:

$$7^{\log n} = 7^{\frac{\log_7 n}{\log_7 2}}$$
$$= (7^{\log_7 n})^{(1/\log_7 2)}$$

$$= n^{(1/\log_7 2)}$$

$$= n^{\frac{\log_2 7}{\log_2 2}}$$

$$= n^{\log 7}$$

So the solution is  $T(n) = \Theta(n^{\lg 7}) = \Theta(n^{2.81...})$ 

#### • Note:

- We are 'hiding' a much bigger constant in  $\Theta()$  than before.
- Currently best known bound is  $O(n^{2.376..})$  (Coppersmith and Winograd'78).
- Lower bound is (trivially)  $\Omega(n^2)$ .
- Big open problem!!
- Strassen's algorithm has been found to be efficient in practice once n is large enough. For small values of n the straightforward cubic algorithm is used instead. The crossover point where Strassen becomes more efficient depends from system to system.