

# CE100 Algorithms and Programming II

## Week-3 (Matrix Multiplication/ Quick Sort)

Spring Semester, 2021-2022

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# Matrix Multiplication / Quick Sort

## Outline

- Matrix Multiplication
  - Traditional
  - Recursive
  - Strassen

# Outline

- Quicksort
  - Hoare Partitioning
  - Lomuto Partitioning
  - Recursive Sorting

# Outline

- Quicksort Analysis
  - Randomized Quicksort
  - Randomized Selection
    - Recursive
    - Medians

# Matrix Multiplication

- Input:  $A = [a_{ij}], B = [b_{ij}]$
- Output:  $C = [c_{ij}] = A \cdot B \implies i, j = 1, 2, 3, \dots, n$

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

# Matrix Multiplication

$$\begin{pmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ \boxed{c_{21}} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \boxed{a_{21} \quad a_{22} \quad \cdots \quad a_{2n}} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix} \cdot \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ \boxed{b_{21}} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{nn} \end{pmatrix}$$

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

# Matrix Multiplication: Standard Algorithm

Running Time:  $\Theta(n^3)$

```
for i=1 to n do
  for j=1 to n do
    C[i,j] = 0
    for k=1 to n do
      C[i,j] = C[i,j] + A[i,k] + B[k,j]
    endfor
  endfor
endfor
```

# Matrix Multiplication: Divide & Conquer

**IDEA:** Divide the  $n \times n$  matrix into  $2 \times 2$  matrix of  $(n/2) \times (n/2)$  submatrices.

$$\begin{pmatrix} \boxed{c_{11}} & c_{12} \\ c_{21} & c_{22} \end{pmatrix} = \begin{pmatrix} \boxed{a_{11}} & \boxed{a_{12}} \\ a_{21} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} \boxed{b_{11}} & b_{12} \\ \boxed{b_{21}} & b_{22} \end{pmatrix} \quad \begin{pmatrix} c_{11} & c_{12} \\ \boxed{c_{21}} & c_{22} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ \boxed{a_{21}} & \boxed{a_{22}} \end{pmatrix} \cdot \begin{pmatrix} \boxed{b_{11}} & b_{12} \\ \boxed{b_{21}} & b_{22} \end{pmatrix}$$

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

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

$$\begin{pmatrix} c_{11} & \boxed{c_{12}} \\ c_{21} & c_{22} \end{pmatrix} = \begin{pmatrix} \boxed{a_{11}} & \boxed{a_{12}} \\ a_{21} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} b_{11} & \boxed{b_{12}} \\ b_{21} & \boxed{b_{22}} \end{pmatrix} \quad \begin{pmatrix} c_{11} & c_{12} \\ \boxed{c_{21}} & c_{22} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ \boxed{a_{21}} & \boxed{a_{22}} \end{pmatrix} \cdot \begin{pmatrix} b_{11} & \boxed{b_{12}} \\ b_{21} & \boxed{b_{22}} \end{pmatrix}$$

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

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



## Matrix Multiplication: Divide & Conquer

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \cdot \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$8 \text{ mults and } 4 \text{ adds of } (n/2) \times (n/2) \text{ submatrices} = \begin{cases} c_{11} = a_{11}b_{11} + a_{12}b_{21} \\ c_{21} = a_{21}b_{11} + a_{22}b_{21} \\ c_{12} = a_{11}b_{12} + a_{12}b_{22} \\ c_{22} = a_{21}b_{12} + a_{22}b_{22} \end{cases}$$

# 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[1,1] = MATRIX-MULTIPLY (A[1,1], B[1,1]) +
             MATRIX-MULTIPLY (A[1,2], B[2,1]);

    C[1,2] = MATRIX-MULTIPLY (A[1,1], B[1,2]) +
             MATRIX-MULTIPLY (A[1,2], B[2,2]);

    C[2,1] = MATRIX-MULTIPLY (A[2,1], B[1,1]) +
             MATRIX-MULTIPLY (A[2,2], B[2,1]);

    C[2,2] = MATRIX-MULTIPLY (A[2,1], B[1,2]) +
             MATRIX-MULTIPLY (A[2,2], B[2,2]);
endif

return C
```

## Matrix Multiplication: Divide & Conquer Analysis

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

- 8 recursive calls  $\implies 8T(\dots)$
- each problem has size  $n/2 \implies \dots T(n/2)$
- Submatrix addition  $\implies \Theta(n^2)$

## Matrix Multiplication: Solving the Recurrence

- $T(n) = 8T(n/2) + \Theta(n^2)$ 
  - $a = 8, b = 2$
  - $f(n) = \Theta(n^2)$
  - $n^{\log_b^a} = n^3$
- Case 1:  $\frac{n^{\log_b^a}}{f(n)} = \Omega(n^\epsilon) \implies T(n) = \Theta(n^{\log_b^a})$

Similar with ordinary (iterative) algorithm.

## Matrix Multiplication: Strassen's Idea

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

In normal case we need 8 as below.

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \cdot \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$8 \text{ mults and } 4 \text{ adds of } (n/2) \times (n/2) \text{ submatrices} = \begin{cases} c_{11} = a_{11}b_{11} + a_{12}b_{21} \\ c_{21} = a_{21}b_{11} + a_{22}b_{21} \\ c_{12} = a_{11}b_{12} + a_{12}b_{22} \\ c_{22} = a_{21}b_{12} + a_{22}b_{22} \end{cases}$$

# Matrix Multiplication: Strassen's Idea

- **Reminder:**
  - Each submatrix is of size  $(n/2) * (n/2)$
  - Each add/sub operation takes  $\Theta(n^2)$  time
- Compute  $P_1 \dots P_7$  using 7 recursive calls to matrix-multiply

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

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

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

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

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

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

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

# Matrix Multiplication: Strassen's Idea

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

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

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

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

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

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

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

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

$$c_{11} = P_5 + P_4 - P_2 + P_6$$

$$c_{12} = P_1 + P_2$$

$$c_{21} = P_3 + P_4$$

$$c_{22} = P_5 + P_1 - P_3 - P_7$$

## Matrix Multiplication: Strassen's Idea

- 7 recursive multiply calls
- 18 add/sub operations



## Matrix Multiplication: Strassen's Idea

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

$$c_{12} = P_1 + P_2$$

$$= a_{11}(b_{12}-b_{22}) + (a_{11} + a_{12})b_{22}$$

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

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

## Strassen's Algorithm

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

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

# Strassen's Algorithm: Solving the Recurrence

- $T(n) = 7T(n/2) + \Theta(n^2)$ 
  - $a = 7, b = 2$
  - $f(n) = \Theta(n^2)$
  - $n^{\log_b^a} = n^{\lg 7}$
- Case 1:  $\frac{n^{\log_b^a}}{f(n)} = \Omega(n^\epsilon) \implies T(n) = \Theta(n^{\log_b^a})$

$$T(n) = \Theta(n^{\log_2^7})$$

$$2^3 = 8, 2^2 = 4 \text{ so } \implies \log_2^7 \approx 2.81$$

or use <https://www.omnicalculator.com/math/log>

## 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)

# Maximum Subarray Problem

**Input:** An array of values

**Output:** The contiguous subarray that has the largest sum of elements

- Input array:

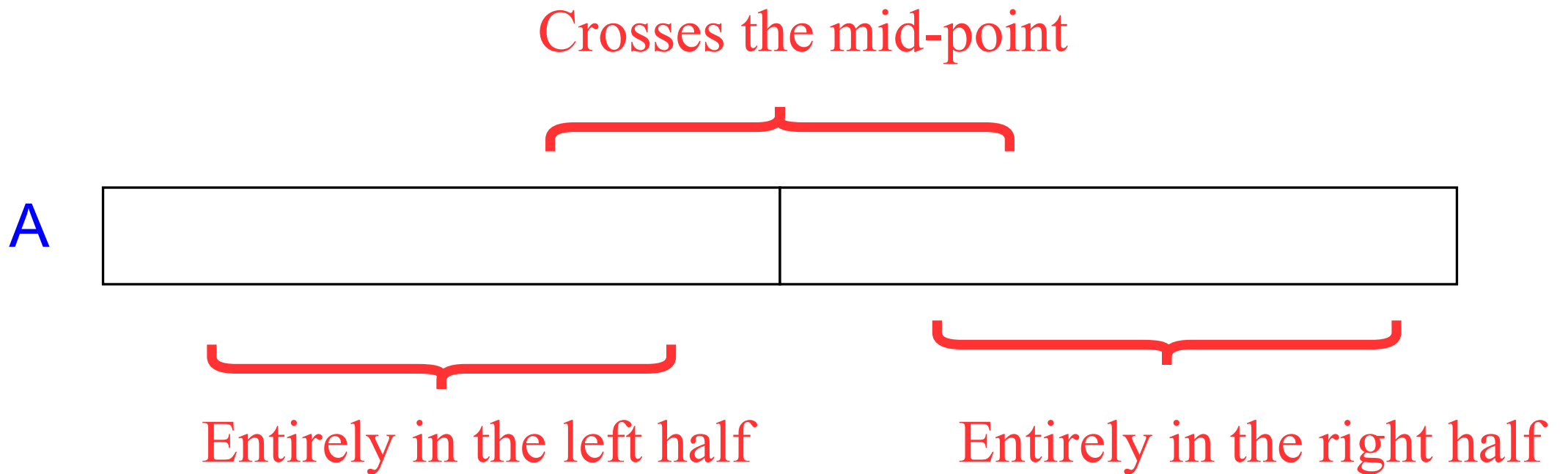
max. contiguous subarray

$[13] [-3] [-25] [20] [-3] [-16] [-23] \quad \overbrace{[18] [20] [-7] [12]} \quad [-22] [-4] [7]$

## Maximum Subarray Problem: Divide & Conquer

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

## Maximum Subarray Problem: Divide & Conquer



## 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:
    - the max subarray of the left-subarray
    - the max subarray of the rightsubarray
    - the max subarray crossing the mid-point

TODO : detailed solution in textbook...



## Conclusion : Divide & Conquer

- 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

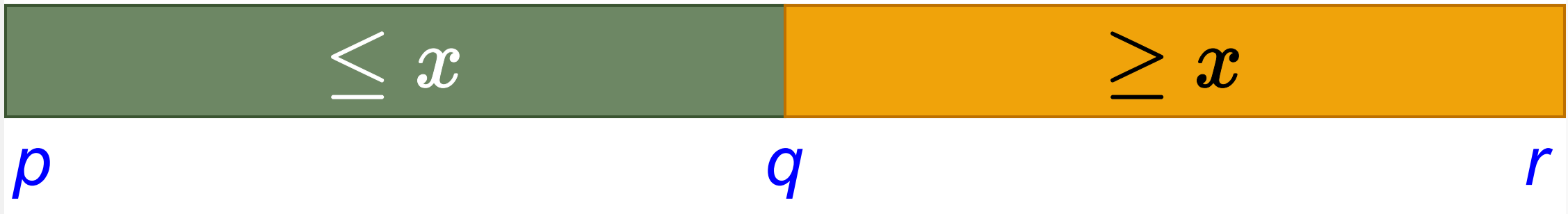
# Quicksort

- One of the most-used algorithms in practice
- Proposed by **C.A.R. Hoare** in 1962.
- Divide-and-conquer algorithm
- In-place algorithm
  - The additional space needed is  $O(1)$
  - The sorted array is returned in the input array
  - *Reminder: Insertion-sort is also an in-place algorithm, but Merge-Sort is not in-place.*
- Very practical

# Quicksort

- **Divide:** Partition the array into 2 subarrays such that elements in the lower part  $\leq$  elements in the higher part
- **Conquer:** Recursively sort 2 subarrays
- **Combine:** Trivial (because in-place)

Key: Linear-time ( $\Theta(n)$ ) partitioning algorithm





# References

TODO