

# Algorithm Design & Analysis (CS3383)<sup>1</sup>

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## Unit 1 Cont.: Randomized D&C

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January 28, 2019

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<sup>1</sup>Thanks to Dr. Patricia Evans and Dr. David Bremner at UNB, Dr. Erik Demaine at MIT for sharing the teaching stuffs

# Outline<sup>2</sup>

## Even More Divide and Conquer

Quicksort

Randomized Quicksort

Randomized median finding

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### <sup>2</sup>Reading:

- ▶ Main textbook (DPV), Divide and conquer algorithms, Chapter 2 mainly 2.4.
- ▶ Algorithms(Cormen): Chapter 5 (5.2, 5.3, and 5.4), Chapter 7, and Chapter 9.
- ▶ Recursive algorithms from Jeff Ericson's Algorithm page <http://jeffe.cs.illinois.edu/teaching/algorithms/notes/99-recurrences.pdf>

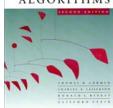
# Contents

Even More Divide and Conquer

Quicksort

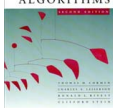
Randomized Quicksort

Randomized median finding



# Quicksort

- Proposed by C.A.R. Hoare in 1962.
- Divide-and-conquer algorithm.
- Sorts “in place” (like insertion sort, but not like merge sort).
- Very practical (with tuning).



# Divide and conquer

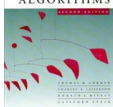
Quicksort an  $n$ -element array:

1. **Divide:** Partition the array into two subarrays around a **pivot**  $x$  such that elements in lower subarray  $\leq x \leq$  elements in upper subarray.



2. **Conquer:** Recursively sort the two subarrays.
3. **Combine:** Trivial.

**Key:** *Linear-time partitioning subroutine.*



# Partitioning subroutine

PARTITION( $A, p, q$ )  $\triangleright A[p \dots q]$

$x \leftarrow A[p]$   $\triangleright$  pivot =  $A[p]$

$i \leftarrow p$

**for**  $j \leftarrow p + 1$  **to**  $q$

**do if**  $A[j] \leq x$

**then**  $i \leftarrow i + 1$

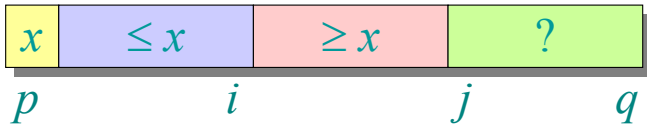
      exchange  $A[i] \leftrightarrow A[j]$

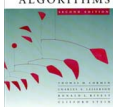
exchange  $A[p] \leftrightarrow A[i]$

**return**  $i$

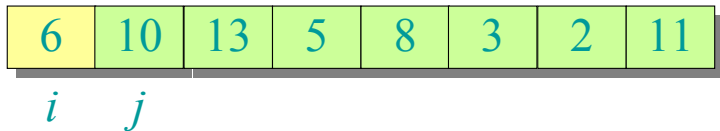
Running time  
=  $O(n)$  for  $n$   
elements.

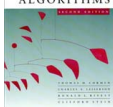
***Invariant:***



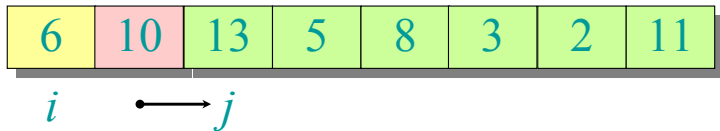


# Example of partitioning

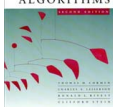




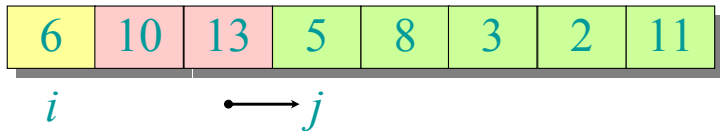
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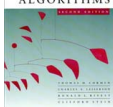




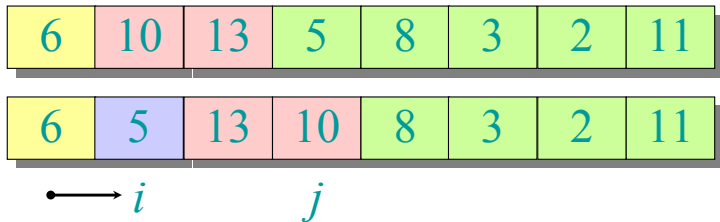


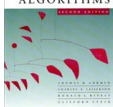
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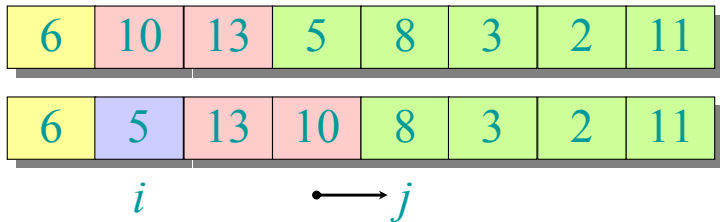


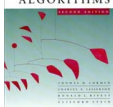
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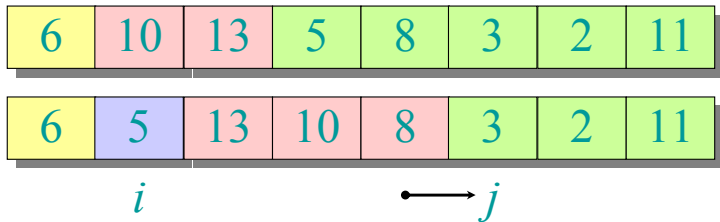


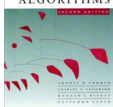
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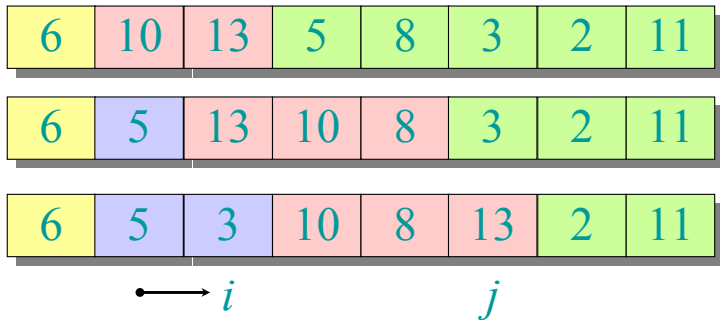


# Example of partitioning



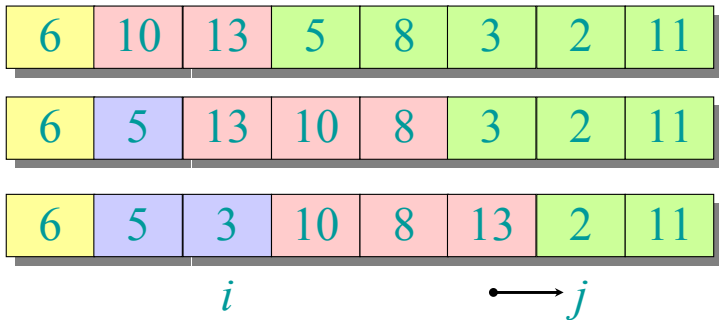


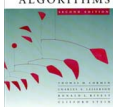
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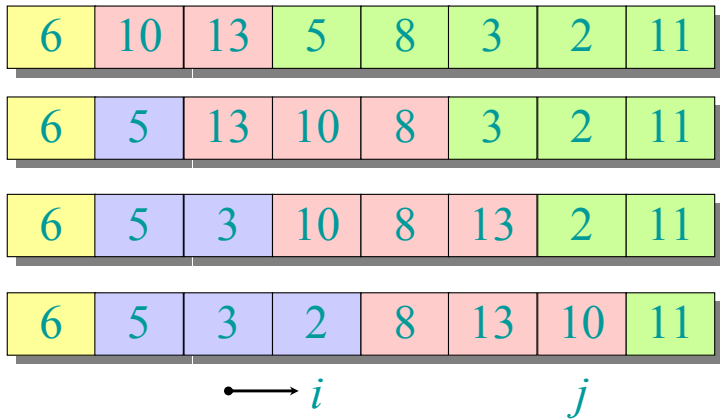


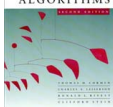
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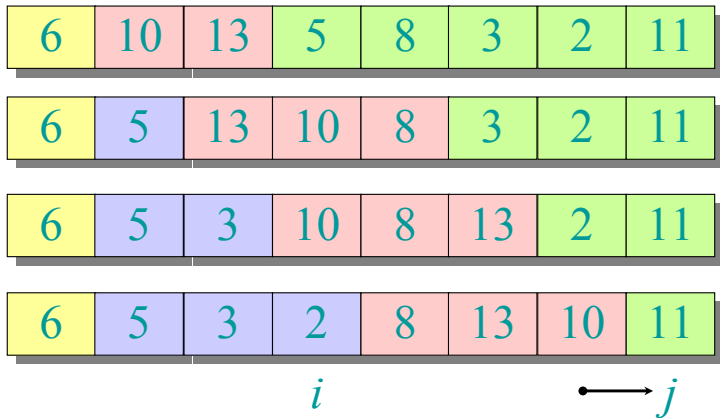


# Example of partitioning

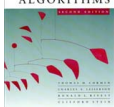




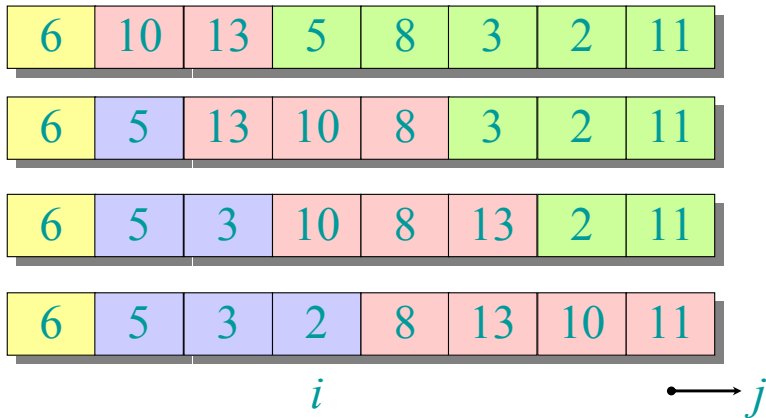
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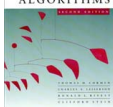




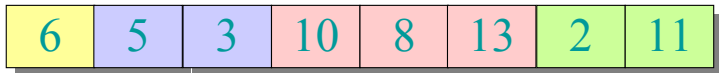
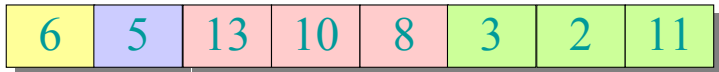


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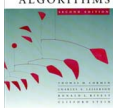




# Example of partitioning



$i$



# Pseudocode for quicksort

QUICKSORT( $A, p, r$ )

  if  $p < r$

    then  $q \leftarrow$  PARTITION( $A, p, r$ )

      QUICKSORT( $A, p, q-1$ )

      QUICKSORT( $A, q+1, r$ )

**Initial call:** QUICKSORT( $A, 1, n$ )

# Analysis of quicksort

- ▶ Quicksort is  $\Theta(n^2)$  in the worst case. What kind of input is bad? **Sorted ( $\uparrow$  /  $\downarrow$ ), Array of Same Elements? Why?**
- ▶ Quicksort is supposed to be fast "in practice".
- ▶ We can choose a better pivot in  $O(n)$  time, but we'll see it's a bit complicated.
- ▶ What if we choose a random element as pivot?

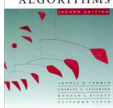
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Randomized Quicksort

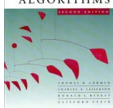
Randomized median finding



# Randomized quicksort

**IDEA:** Partition around a *random* element.

- Running time is independent of the input order.
- No assumptions need to be made about the input distribution.
- No specific input elicits the worst-case behavior.
- The worst case is determined only by the output of a random-number generator.



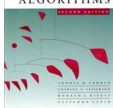
# Randomized quicksort analysis

Let  $T(n)$  = the random variable for the running time of randomized quicksort on an input of size  $n$ , assuming random numbers are independent.

For  $k = 0, 1, \dots, n-1$ , define the *indicator random variable*

$$X_k = \begin{cases} 1 & \text{if PARTITION generates a } k : n-k-1 \text{ split,} \\ 0 & \text{otherwise.} \end{cases}$$

$E[X_k] = \Pr\{X_k = 1\} = 1/n$ , since all splits are equally likely, assuming elements are distinct.

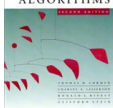


# Analysis (continued)

$$T(n) = \begin{cases} T(0) + T(n-1) + \Theta(n) & \text{if } 0 : n-1 \text{ split,} \\ T(1) + T(n-2) + \Theta(n) & \text{if } 1 : n-2 \text{ split,} \\ \vdots & \\ T(n-1) + T(0) + \Theta(n) & \text{if } n-1 : 0 \text{ split,} \end{cases}$$

$$= \sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))$$

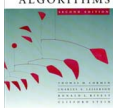




# Calculating expectation

$$E[T(n)] = E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right]$$

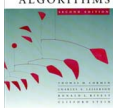
Take expectations of both sides.



# Calculating expectation

$$\begin{aligned} E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))] \end{aligned}$$

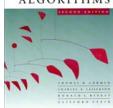
Linearity of expectation.



# Calculating expectation

$$\begin{aligned} E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))] \\ &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(k) + T(n-k-1) + \Theta(n)] \end{aligned}$$

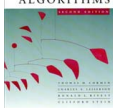
Independence of  $X_k$  from other random choices.



# Calculating expectation

$$\begin{aligned} E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))] \\ &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(k) + T(n-k-1) + \Theta(n)] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} E[T(k)] + \frac{1}{n} \sum_{k=0}^{n-1} E[T(n-k-1)] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n) \end{aligned}$$

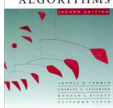
Linearity of expectation;  $E[X_k] = 1/n$ .



# Calculating expectation

$$\begin{aligned} E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))] \\ &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(k) + T(n-k-1) + \Theta(n)] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} E[T(k)] + \frac{1}{n} \sum_{k=0}^{n-1} E[T(n-k-1)] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n) \\ &= \frac{2}{n} \sum_{k=1}^{n-1} E[T(k)] + \Theta(n) \end{aligned}$$

Summations have identical terms.



# Hairy recurrence

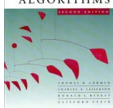
$$E[T(n)] = \frac{2}{n} \sum_{k=2}^{n-1} E[T(k)] + \Theta(n)$$

(The  $k = 0, 1$  terms can be absorbed in the  $\Theta(n)$ .)

**Prove:**  $E[T(n)] \leq an \lg n$  for constant  $a > 0$ .

- Choose  $a$  large enough so that  $an \lg n$  dominates  $E[T(n)]$  for sufficiently small  $n \geq 2$ .

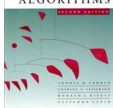
**Use fact:**  $\sum_{k=2}^{n-1} k \lg k \leq \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2$  (exercise).



# Substitution method

$$\begin{aligned} E[T(n)] &\leq \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n) \\ &\leq \frac{2a}{n} \left( \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + \Theta(n) \end{aligned}$$

Use fact.

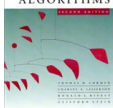


# Substitution method

$$\begin{aligned}E[T(n)] &\leq \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n) \\&\leq \frac{2a}{n} \left( \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + \Theta(n) \\&= an \lg n - \left( \frac{an}{4} - \Theta(n) \right)\end{aligned}$$

Express as *desired* – *residual*.





# Substitution method

$$\begin{aligned}E[T(n)] &\leq \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n) \\&= \frac{2a}{n} \left( \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + \Theta(n) \\&= an \lg n - \left( \frac{an}{4} - \Theta(n) \right) \\&\leq an \lg n ,\end{aligned}$$

if  $a$  is chosen large enough so that  $an/4$  dominates the  $\Theta(n)$ .

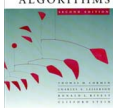
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# Order statistics

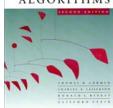
Select the  $i$ th smallest of  $n$  elements (the element with *rank*  $i$ ).

- $i = 1$ : *minimum*;
- $i = n$ : *maximum*;
- $i = \lfloor (n+1)/2 \rfloor$  or  $\lceil (n+1)/2 \rceil$ : *median*.

*Naive algorithm*: Sort and index  $i$ th element.

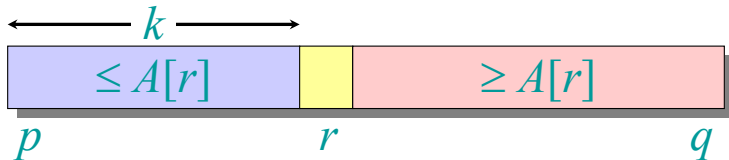
Worst-case running time =  $\Theta(n \lg n) + \Theta(1)$   
=  $\Theta(n \lg n)$ ,

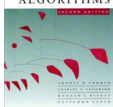
using merge sort or heapsort (*not* quicksort).



# Randomized divide-and-conquer algorithm

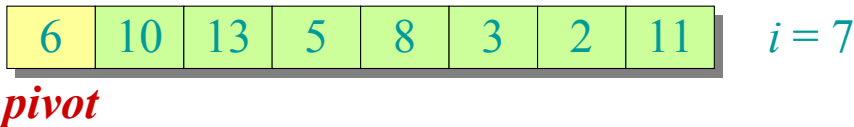
**RAND-SELECT**( $A, p, q, i$ )     $\triangleright$   $i$ th smallest of  $A[p..q]$   
  **if**  $p = q$  **then return**  $A[p]$   
   $r \leftarrow$  **RAND-PARTITION**( $A, p, q$ )  
   $k \leftarrow r - p + 1$      $\triangleright k = \text{rank}(A[r])$   
  **if**  $i = k$  **then return**  $A[r]$   
  **if**  $i < k$   
    **then return** **RAND-SELECT**( $A, p, r - 1, i$ )  
  **else return** **RAND-SELECT**( $A, r + 1, q, i - k$ )



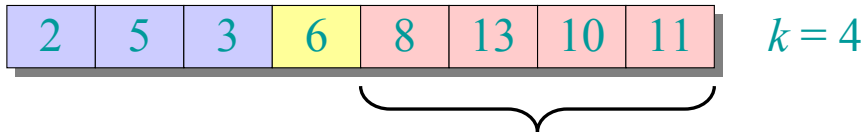


# Example

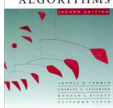
Select the  $i = 7$ th smallest:



Partition:



Select the  $7 - 4 = 3$ rd smallest recursively.



# Intuition for analysis

(All our analyses today assume that all elements are distinct.)

**Lucky:**

$$\begin{aligned}T(n) &= T(9n/10) + \Theta(n) \\ &= \Theta(n)\end{aligned}$$

$$n^{\log_{10/9} 1} = n^0 = 1$$

CASE 3

**Unlucky:**

$$\begin{aligned}T(n) &= T(n-1) + \Theta(n) \\ &= \Theta(n^2)\end{aligned}$$

arithmetic series

***Worse than sorting!***

## Randomized median finding

```
Select2(A, p, q, i)
  n ← q - p + 1
  do {
    r ← RandPartition(A, p, q)
    k ← r - p + 1
    if i = k then return A[r]
  } while ((k < n/4) or (k > 3n/4));
  if i < k
    then return Select2(A, p, r - 1, i)
    else return Select2(A, r + 1, q, i - k)
```

Exercise: Analyze the randomized median finding.<sup>3</sup>

<sup>3</sup>There is a bonus to do this exercise before I post the solution on D2L (by next lecture)