# Quicksort

### Sorting algorithms

- Insertion, selection and bubble sort have quadratic worst-case performance
- The faster comparison based algorithm ?
  O(nlogn)
- · Mergesort and Quicksort

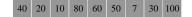
### Quicksort Algorithm

Given an array of n elements (e.g., integers):

- · If array only contains one element, return
- Else
  - pick one element to use as pivot.
  - Partition elements into two sub-arrays:
    - · Elements less than or equal to pivot
    - · Elements greater than pivot
  - Quicksort two sub-arrays
  - Return results

### Example

We are given array of n integers to sort:



#### Pick Pivot Element

There are a number of ways to pick the pivot element. We will use the first element in the array:



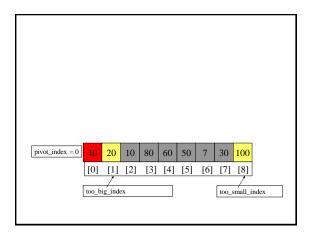
# Partitioning Array

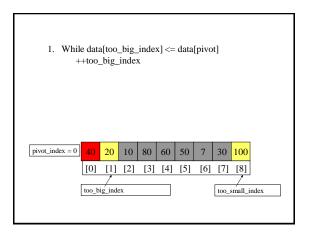
Given a pivot, partition the elements of the array such that the resulting array consists of:

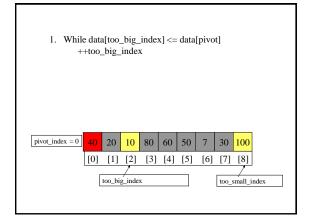
- 1. One sub-array that contains elements >= pivot
- 2. Another sub-array that contains elements < pivot

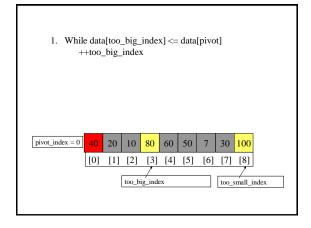
The sub-arrays are stored in the original data array.

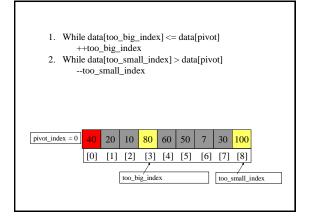
Partitioning loops through, swapping elements below/above pivot.

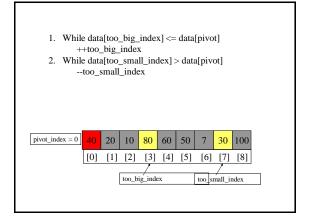


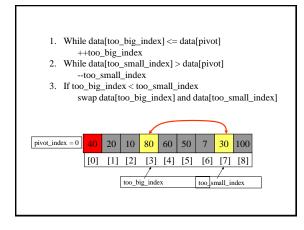


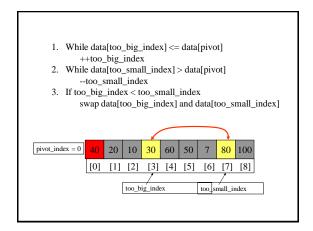


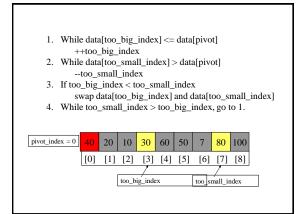


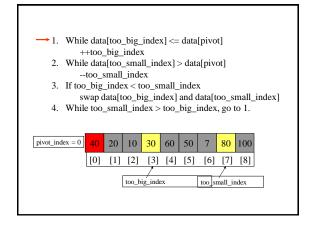


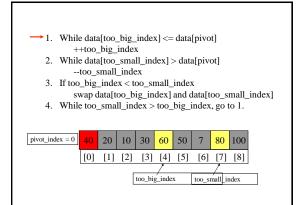


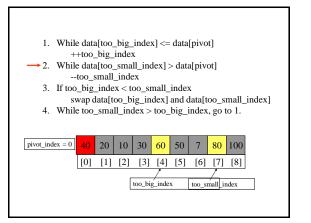


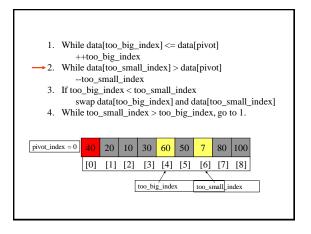


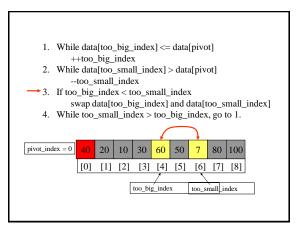


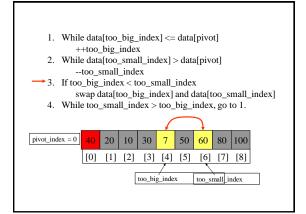


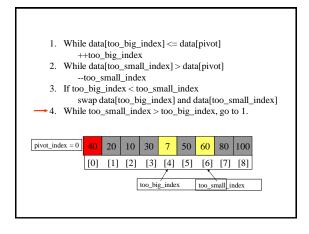


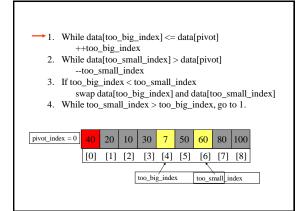


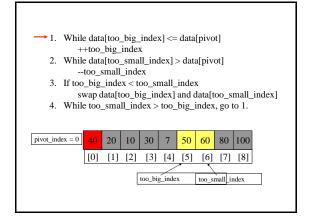


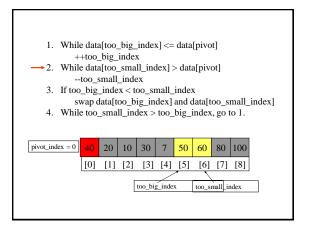


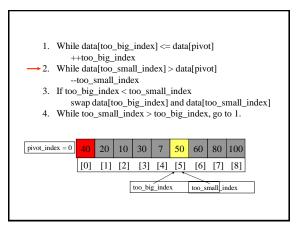


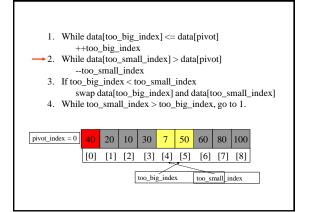


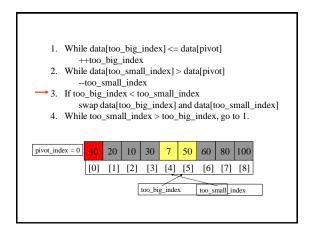


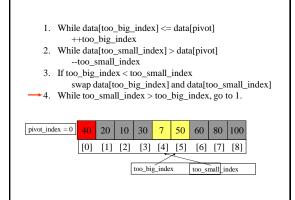


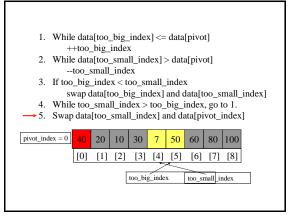


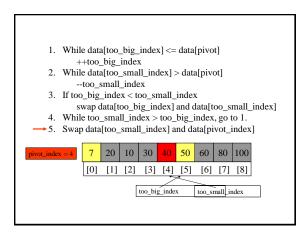


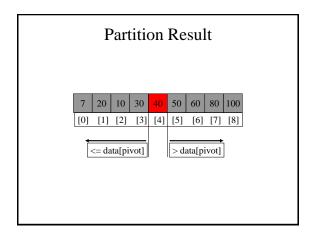


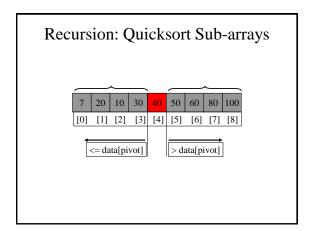


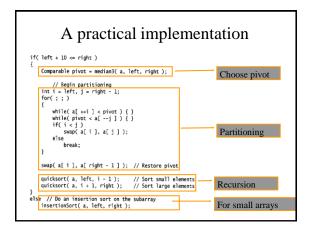












# Small arrays

- For very small arrays, quicksort does not perform as well as insertion sort
  - how small depends on many factors, such as the time spent making a recursive call, the compiler, etc
- Do not use quicksort recursively for small arrays
  - Instead, use a sorting algorithm that is efficient for small arrays, such as insertion sort

# **Quicksort Analysis**

- · Assume that keys are random, uniformly distributed.
- · What is best case running time?

### **Quicksort Analysis**

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- What is best case running time?
  - Recursion:
    - 1. Partition splits array in two sub-arrays of size n/2
    - 2. Quicksort each sub-array

### **Quicksort Analysis**

- Assumption:
  - A random pivot
- · Running time
  - pivot selection: constant time, i.e. O(1)
  - partitioning: linear time, i.e. O(N)
  - running time of the two recursive calls
- T(N) = O(1) + O(N) + T(i) + T(N-i-1)
- T(N)=cN+T(i)+T(N-(i+1))
  - where c is a constant
  - i is number of elements in array 1

#### Best-case Analysis

- · What will be the best case?
  - Partition is perfectly balanced.
  - Pivot is always in the middle (median of the array)
  - T(N) = cN + 2T(N/2)
  - $T(N) = O(N \log N)$

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\begin{array}{ll} T(N) &=& 2T(N/2) + cN \\ \frac{T(N)}{N} &=& \frac{T(N/2)}{N/2} + c \\ \frac{T(N/2)}{N/2} &=& \frac{T(N/4)}{N/4} + c \\ \frac{T(N/4)}{N/4} &=& \frac{T(N/8)}{N/8} + c \\ \vdots &\vdots &\vdots &\vdots &\vdots \\ \frac{T(2)}{2} &=& \frac{T(1)}{1} + c \\ \frac{T(N)}{N} &=& \frac{T(1)}{1} + c \log N \\ \frac{T(N)}{N} &=& \frac{T(1)}{1} + c \log N \\ \frac{T(N)}{N} &=& \frac{cN \log N + N = O(N \log N)}{T(N)} \end{array}
```

#### Average-Case Analysis

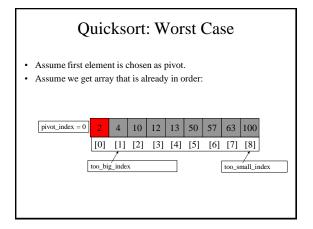
- Assume
  - Each of the sizes for S1 is equally likely
- This assumption is valid for our pivoting (median-of-three) strategy
- On average, the running time is  $O(N \ log \ N)$

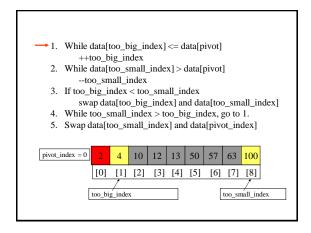
# **Quicksort Analysis**

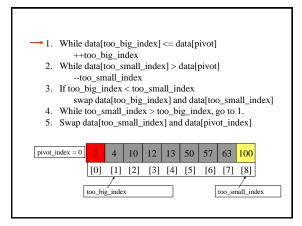
- · Assume that keys are random, uniformly distributed.
- Best case running time: O(n log<sub>2</sub>n)
- Worst case running time?

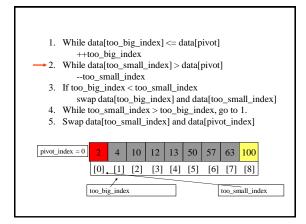
## **Quicksort Analysis**

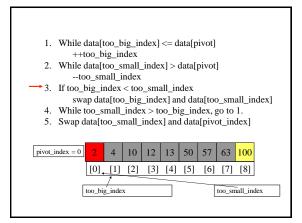
- Assume that keys are random, uniformly distributed.
- Best case running time: O(n log<sub>2</sub>n)
- · Worst case running time?
  - Recursion
    - 1. Partition splits array in two sub-arrays:
      - · one sub-array of size 0
      - the other sub-array of size n-1
    - 2. Quicksort each sub-array

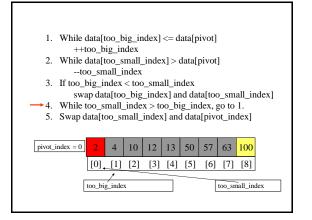




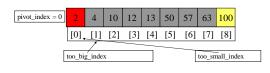


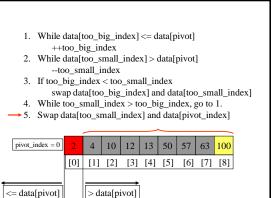






- $1. \begin{tabular}{ll} While $$ data[too\_big\_index] <= data[pivot] \\ ++too\_big\_index \end{tabular}$
- 2. While data[too\_small\_index] > data[pivot] --too\_small\_index
- 3. If too\_big\_index < too\_small\_index swap data[too\_big\_index] and data[too\_small\_index]
- 4. While too\_small\_index > too\_big\_index, go to 1.
- → 5. Swap data[too\_small\_index] and data[pivot\_index]





#### Worst-Case Analysis

- What will be the worst case?
  - The pivot is the smallest element, all the time
  - Partition is always unbalanced
  - i = 0

$$\begin{array}{rcl} T(N) & = & T(N-1) + cN \\ T(N-1) & = & T(N-2) + c(N-1) \\ T(N-2) & = & T(N-3) + c(N-2) \\ & \vdots \\ T(2) & = & T(1) + c(2) \\ T(N) & = & T(1) + c \sum_{i=1}^{N} i = O(N^2) \end{array}$$

### Quicksort is 'faster' than Mergesort

- · Both quicksort and mergesort take O(N log N) in the average case.
- · Why is quicksort faster than mergesort?
  - The inner loop consists of an increment/decrement (by 1, which is fast), a test and a jump.
  - There is no extra organising as in mergesort.

## **Quicksort Analysis**

- · Assume that keys are random, uniformly distributed.
- Best case running time: O(n log<sub>2</sub>n)
- Worst case running time: O(n2)

### **Quicksort Analysis**

- Assume that keys are random, uniformly distributed.
- Best case running time: O(n log<sub>2</sub>n)
- Worst case running time: O(n²)
- What can we do to avoid worst case?

# Improved Pivot Selection

Pick median value of three elements from data array:  $data[0],\, data[n/2],\, and\, data[n-1].$ 

Use this median value as pivot.