CSE214 – Analysis of Algorithms

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https://github.com/FurkanGozukara/CSE214 2018

Lecture 5 Quicksort

Based on Kruse's and Ryba's Lecture Notes

Sorting algorithms

- Insertion, selection and bubble sort have quadratic worstcase performance
- The faster comparison based algorithm ?
 O(nlogn)

Mergesort and Quicksort

Quicksort Algorithm

- Fastest known sorting algorithm in practice
 - Caveats: not stable,
 - Vulnerable to certain attacks

Average case complexity $\rightarrow O(N \log N)$

- Worst-case complexity $\rightarrow O(N^2)$
 - Rarely happens, if coded correctly

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:

40	20 1	10 80	60	50	7	30	100
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Pick Pivot Element

There are a number of ways to pick the pivot element. In this example, we will use the first element in the array:

40	20	10	80	60	50	7	30	100
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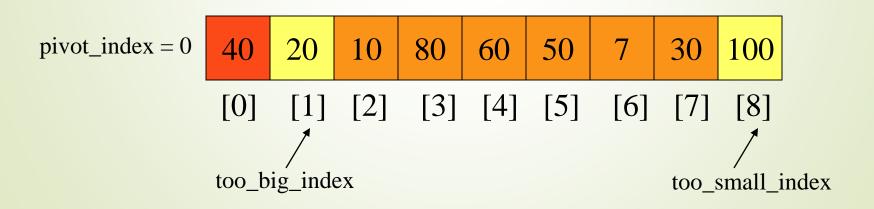
Partitioning Array

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

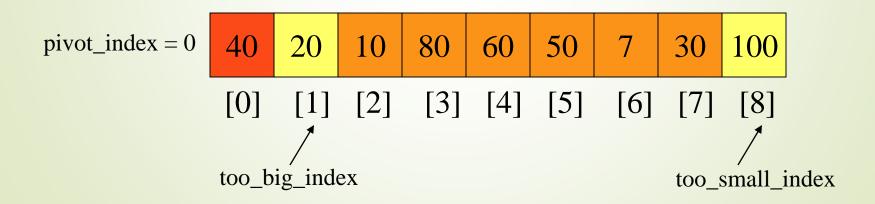
- 1. One sub-array that contains elements >= pivot
- Another sub-array that contains elements < pivot

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

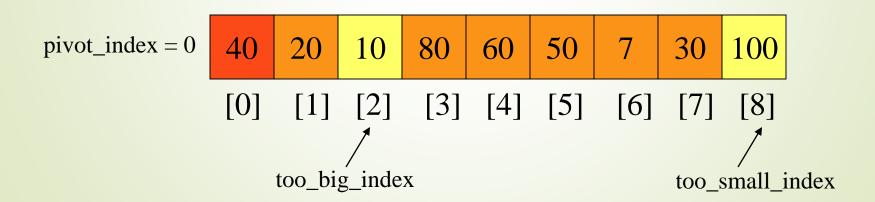
Partitioning loops through, swapping elements below/above pivot.



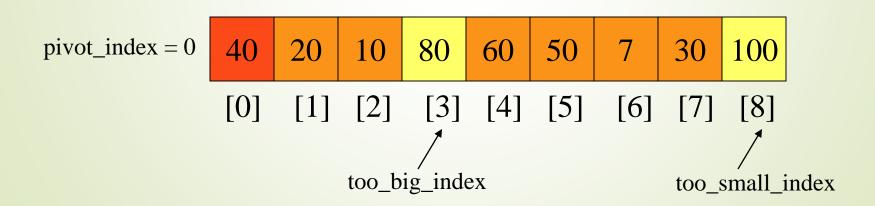
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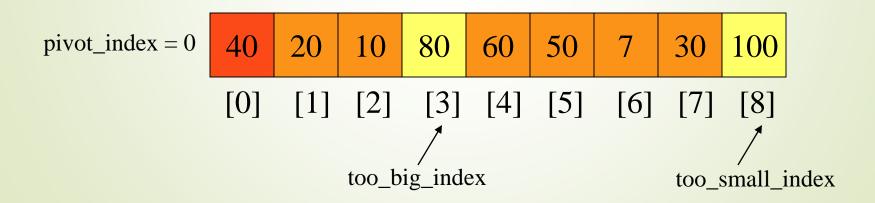
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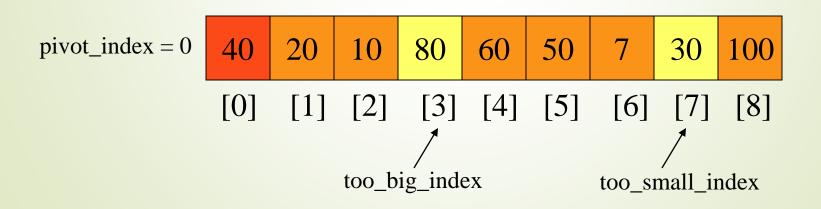
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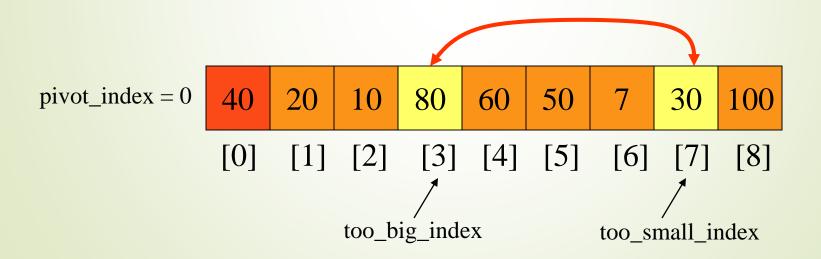
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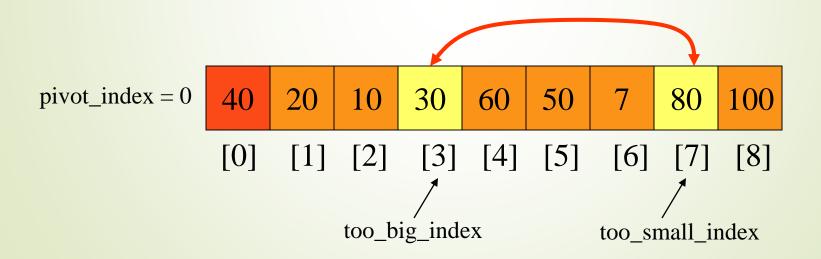
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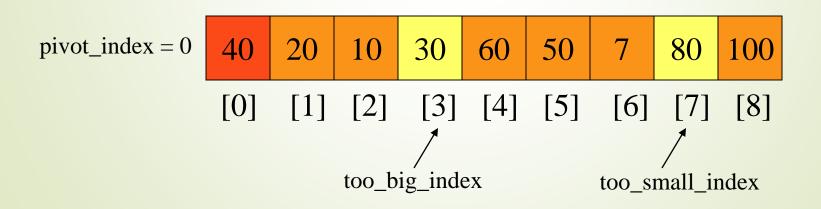
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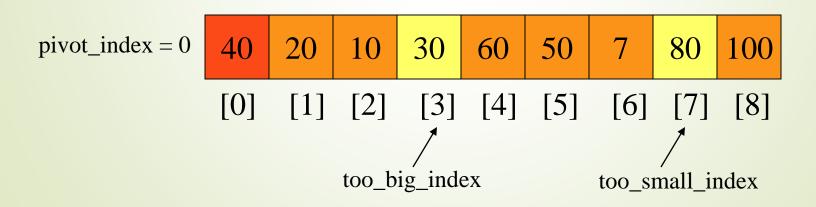
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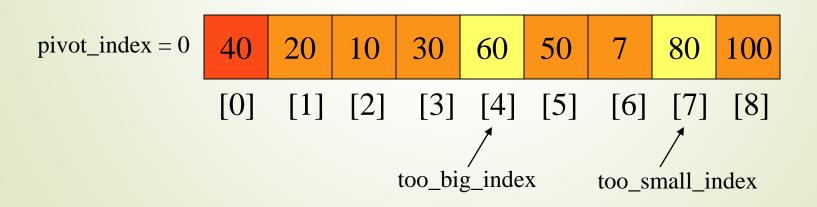
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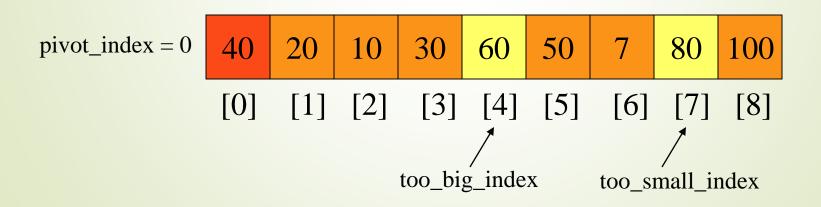
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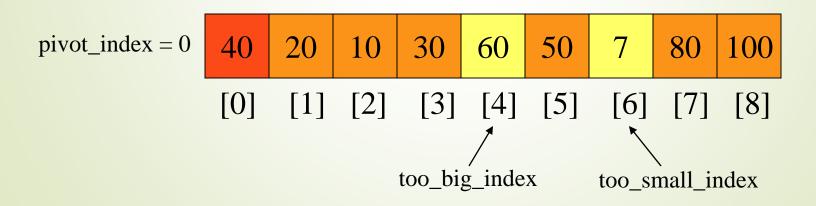
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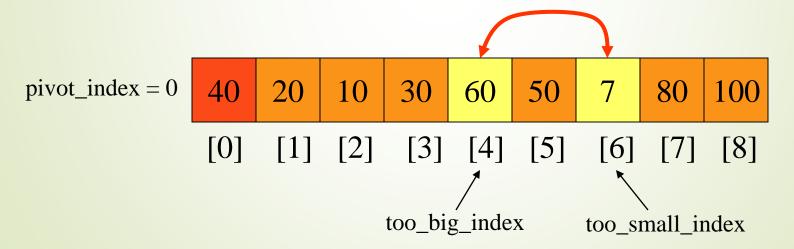
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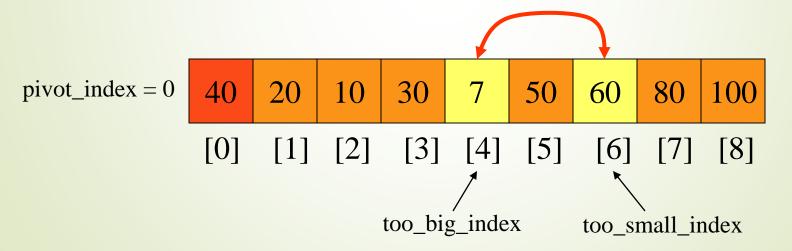
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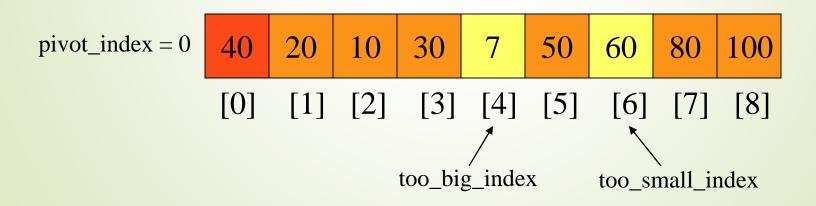
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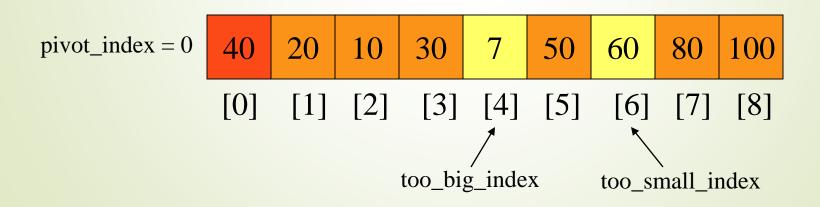
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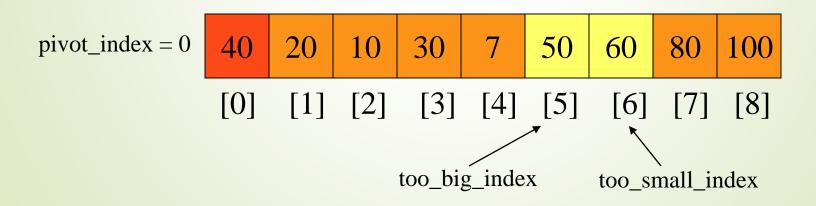
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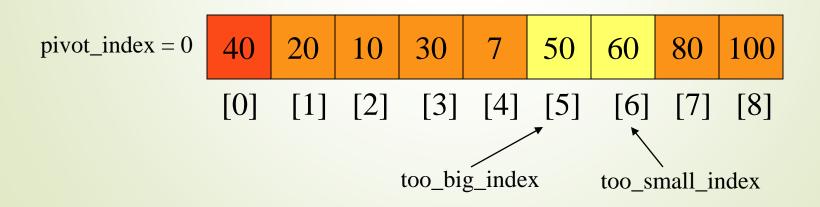
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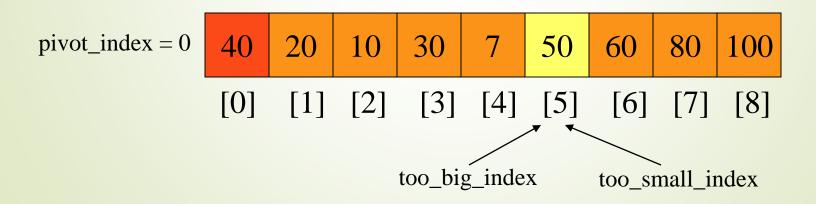
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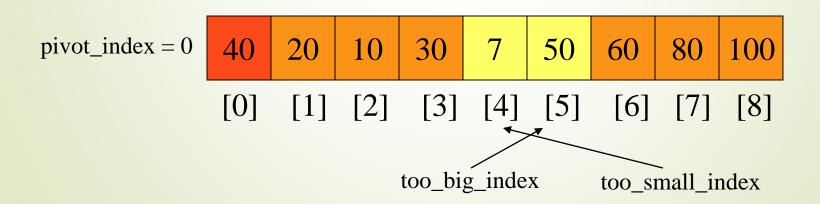
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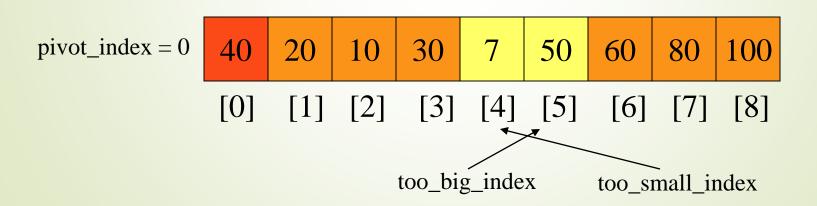
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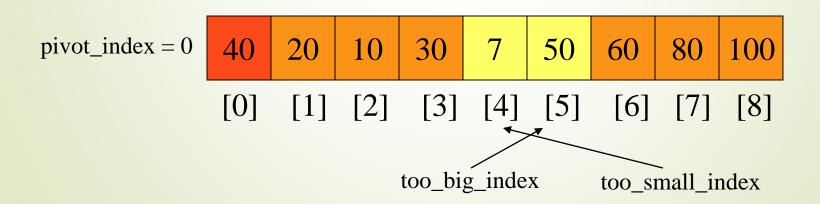
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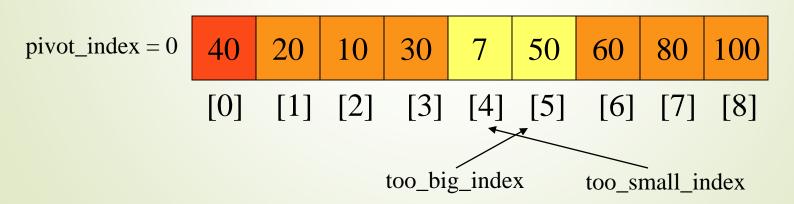
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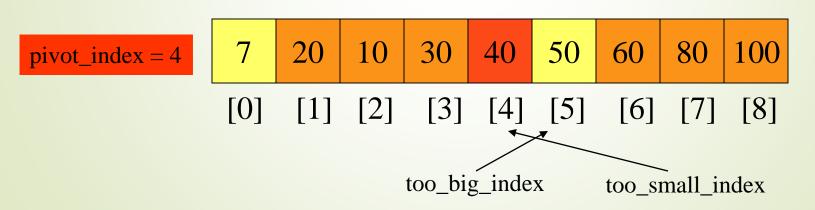
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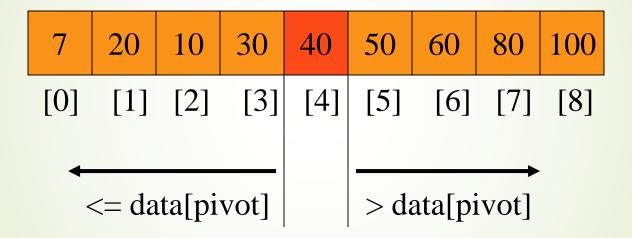
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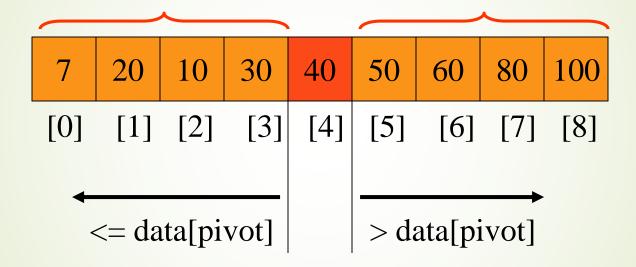
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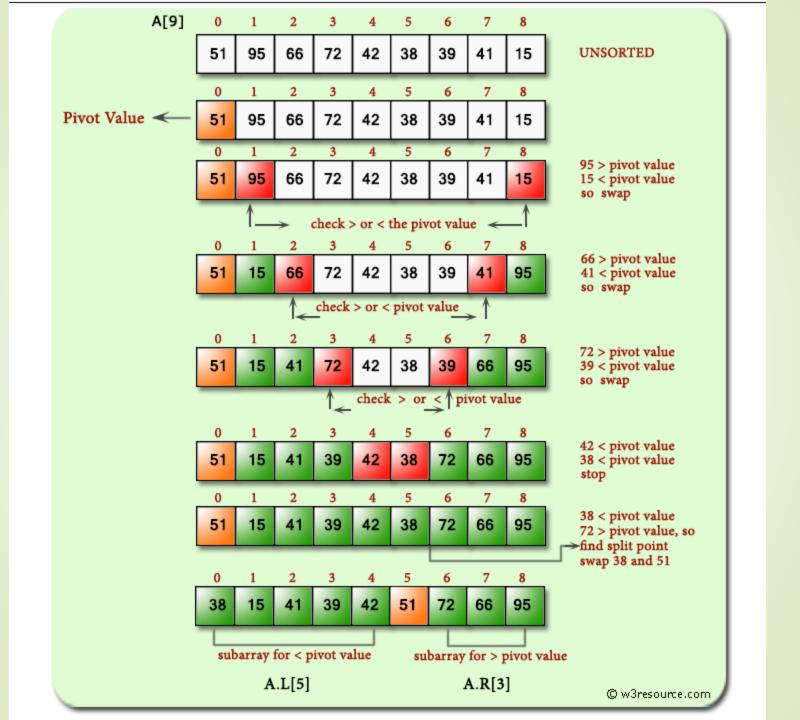


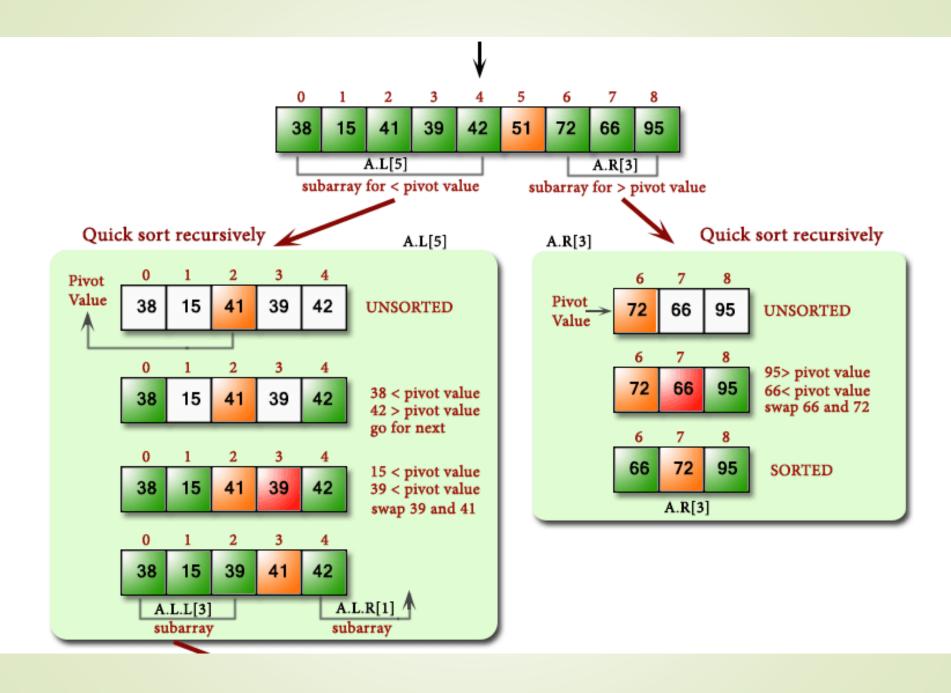
Partition Result

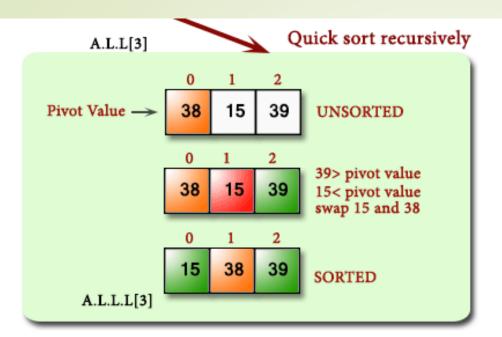


Recursion: Quicksort Sub-arrays

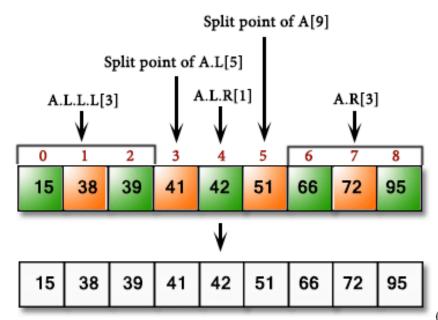


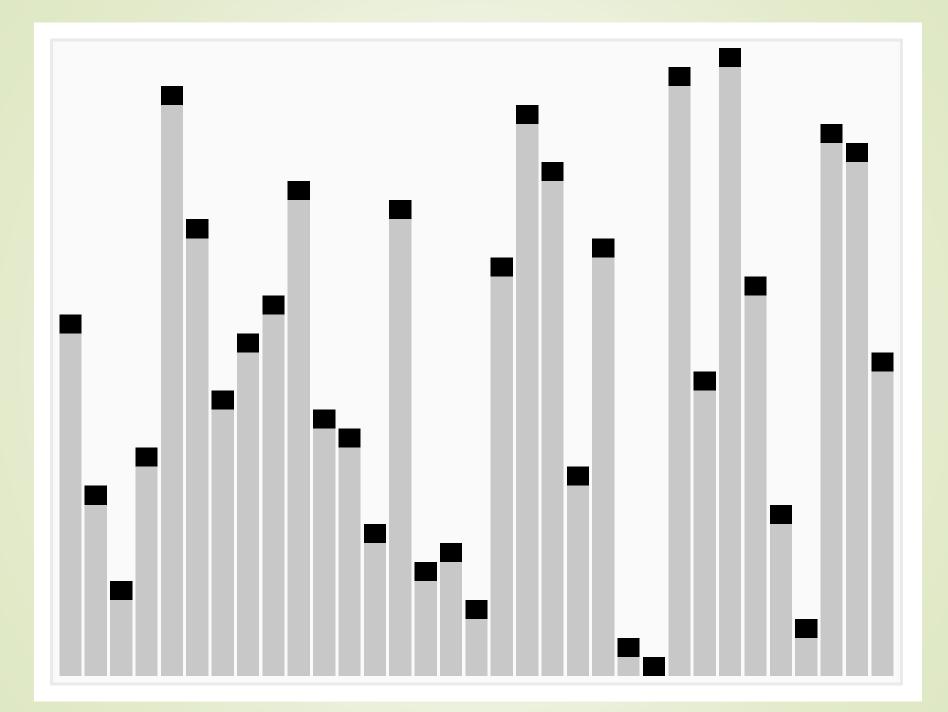


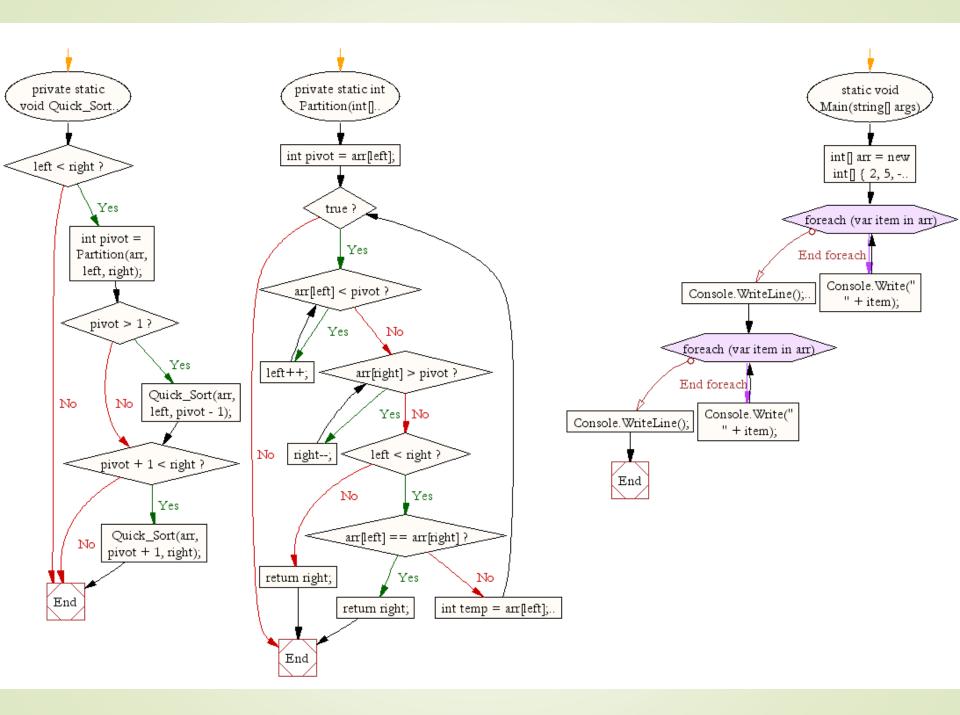




FINAL SORTING







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- What is best case running time?

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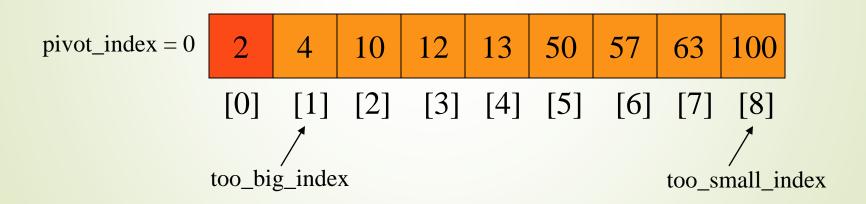
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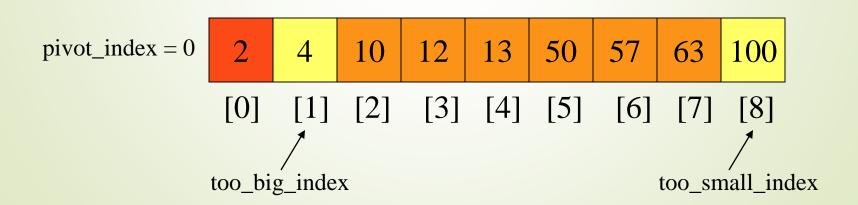
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Quicksort: Worst Case

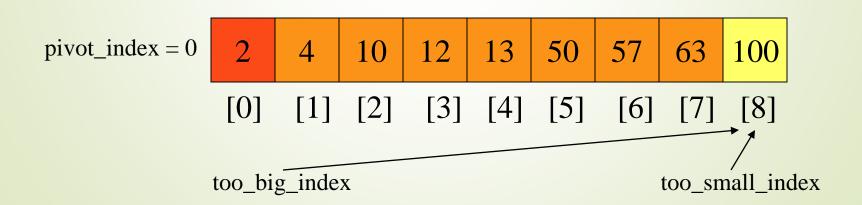
- Assume first element is chosen as pivot.
- Assume we get array that is already in order:



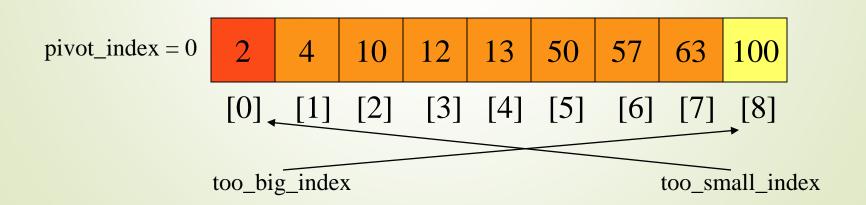
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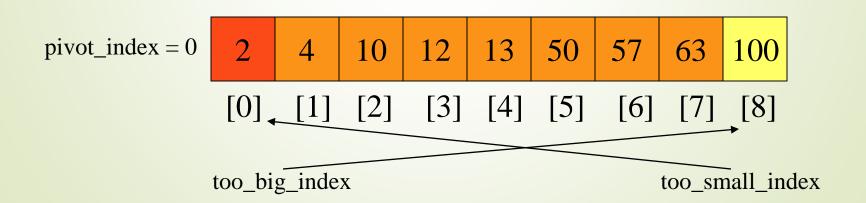
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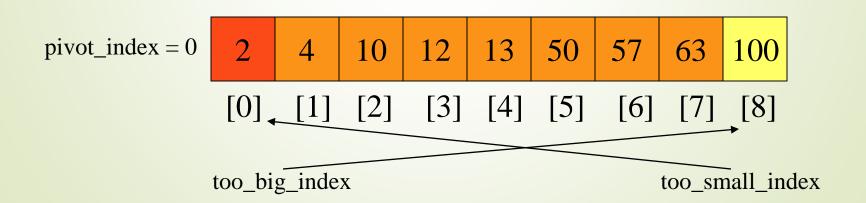
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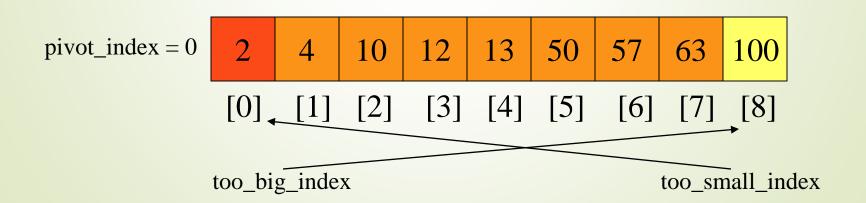
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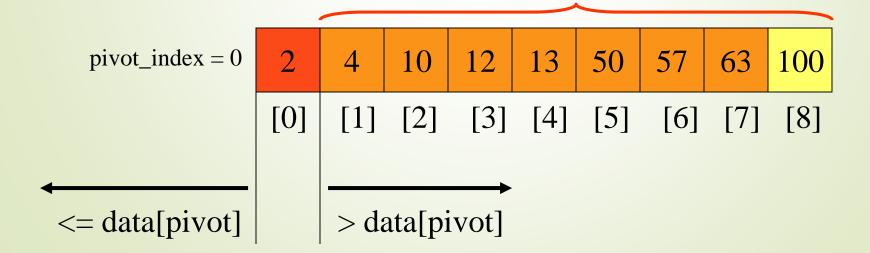
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 - one sub-array of size 0
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- What can we do to avoid worst case?

Improved Pivot Selection

Pick median value of three elements from data array:

data[0], data[n/2], and data[n-1].

- Use this median of the array
 - Partitioning always cuts the array into roughly half
 - An optimal quicksort (O(N log N))
 - However, hard to find the exact median

Improving Performance of Quicksort

- ■Improved selection of pivot.
- For sub-arrays of size 3 or less, apply brute force search:
 - Sub-array of size 1: trivial
 - Sub-array of size 2:
 - if(data[first] > data[second]) swap them
 - Sub-array of size 3?

Pivot: median of three

- We will use median of three
 - Compare just three elements: the leftmost, rightmost and center
 - Swap these elements if necessary so that

```
    A[left] = Smallest
    A[right] = Largest
    A[center] = Median of three
```

- Pick A[center] as the pivot
- ► Swap A[center] and A[right 1] so that pivot is at second last position

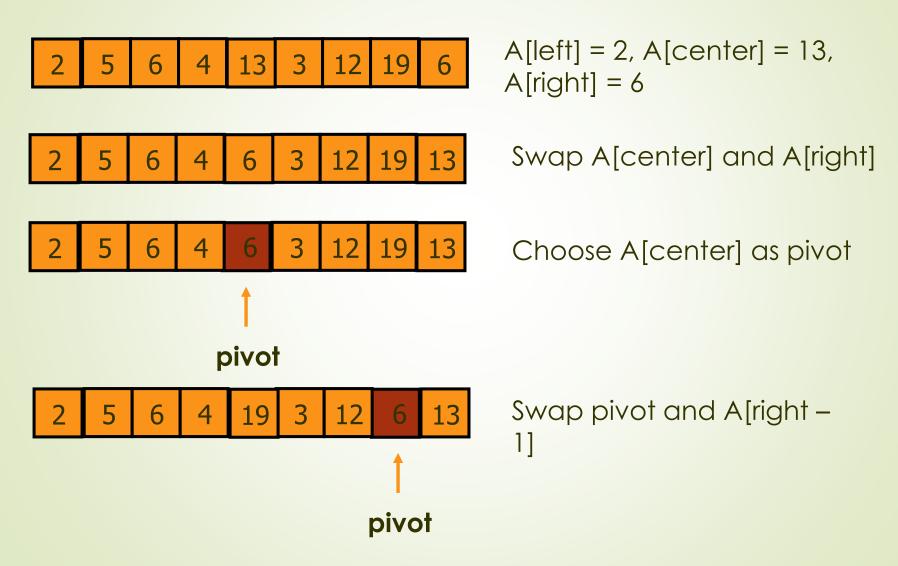
Pivot: median of three

Code for partitioning with median of three pivot:

```
int center = ( left + right ) / 2;
if( a[ center ] < a[ left ] )
        swap( a[ left ], a[ center ] );
if( a[ right ] < a[ left ] )
        swap( a[ left ], a[ right ] );
if( a[ right ] < a[ center ] )
        swap( a[ center ], a[ right ] );

        // Place pivot at position right - 1
swap( a[ center ], a[ right - 1 ] );</pre>
```

Pivot: median of three



Note we only need to partition A[left + 1, ..., right – 2]

Implementation of partitioning step

- Works only if pivot is picked as median-of-three.
 - A[left] <= pivot and
 A[right] >= pivot
 - Thus, only need to partition
 A[left + 1, ..., right 2]
- j will not run past the end
 - because a[left] <= pivot</p>
- i will not run past the end
 - because a[right-1] = pivot

```
int i = left, j = right - 1;
for(;;)
{
    while(a[++i] < pivot) { }
    while(pivot < a[--j]) { }
    if(i < j)
        swap(a[i],a[j]);
    else
        break;
}</pre>
```

Main Quicksort Routine

```
if( left + 10 <= right )
    Comparable pivot = median3( a, left, right );
                                                                 Choose pivot
       // Begin partitioning
    int i = left, j = right - 1;
    for(;;)
       while( a[ ++i ] < pivot ) { }
       while( pivot < a[ --j ] ) { }
       if(i < i)
                                                                 Partitioning
           swap( a[ i ], a[ j ] );
       e1se
           break;
    swap( a[ i ], a[ right - 1 ] ); // Restore pivot
    quicksort( a, left, i - 1 ); // Sort small elements
                                                                Recursion
   quicksort( a, i + 1, right ); // Sort large elements
else // Do an insertion sort on the subarray
                                                                For small arrays
   insertionSort( a, left, right );
```

Quicksort 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.
 - Mergesort involves a large number of data movements.
 - Quicksort is done in-place.

```
int i = left, j = right - 1;
for(;;)
{
    while(a[++i] < pivot) { }
    while(pivot < a[--j]) { }
    if(i < j)
        swap(a[i], a[j]);
    else
        break;
}</pre>
```

Performance of quicksort

- Worst-case: takes O(n2) time.
- Average-case: takes O(n log n) time.

On typical inputs, quicksort runs faster than other algorithms.

Further Analysis of Quicksort

The analysis is quite tricky.

- Assume all the input elements are distinct
 - no duplicate values makes this code faster!
 - there are better partitioning algorithms when duplicate input elements exist (e.g. Hoare's original code)

Let T(n) = worst-case running time on an array of n elements.

Worst-case of quicksort

- QUICKSORT runs very slowly when its input array is already sorted (or is reverse sorted).
 - o almost sorted data is quite common in the real-world

• This is caused by the partition using the min (or max) element which means that one side of the partition will have has no elements. Therefore:

$$T(n) = T(0) + T(n-1) + \Theta(n)$$
 $= \Theta(1) + T(n-1) + \Theta(n)$
 $= T(n-1) + \Theta(n)$
 $= \Theta(n^2)$ (arithmetic series)

no elements

 $= \Theta(n^2)$ no elements

$$T(n) = T(0) + T(n-1) + cn$$

$$T(n) = T(0) + T(n-1) + cn$$

$$T(n)$$

$$T(n) = T(0) + T(n-1) + cn$$

$$Cn$$

$$T(0)$$

$$T(n-1)$$

$$T(n) = T(0) + T(n-1) + cn$$

$$T(0)$$

$$C(n-1)$$

$$T(0)$$

$$T(n-2)$$

$$T(n) = T(0) + T(n-1) + cn$$

$$T(0) \qquad c(n-1)$$

$$T(0) \qquad T(n-2)$$

$$T(0)$$

$$\Theta(1)$$

$$T(n) = T(0) + T(n-1) + cn$$

$$\begin{array}{c|c}
 & Cn \\
\Theta(1) & C(n-1) \\
\Theta(1) & C(n-2)
\end{array}$$

$$\begin{array}{c}
\Theta\left(\sum_{k=1}^{n} k\right) = \Theta(n^2) \\
T(n) = \Theta(n) + \Theta(n^2) \\
= \Theta(n^2)
\end{array}$$

$$\begin{array}{c}
\Theta(1)
\end{array}$$

Quicksort isn't Quick?

■ In the worst case, quicksort isn't any quicker than insertion sort.

- ■So why bother with quicksort?
- ■It's average case running time is very good, as we'll see.

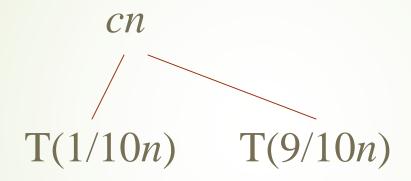
Best-case Analysis

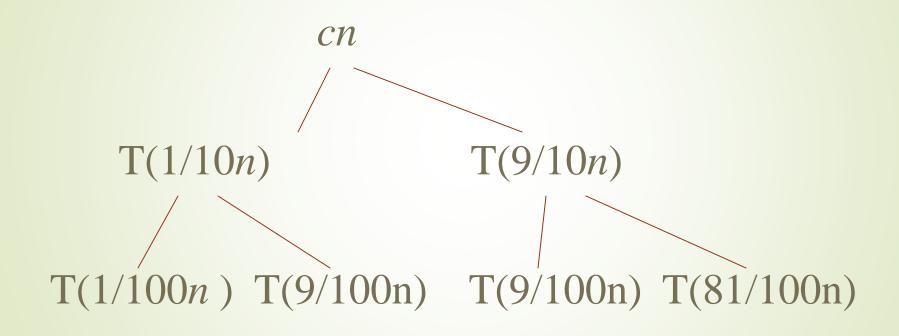
- If we're lucky, PARTITION splits the array evenly:
 Case 2 of the Master Method
- $T(n) = 2T(n/2) + \Theta(n)$
- $= \Theta(n \log n)$ (same as merge sort)

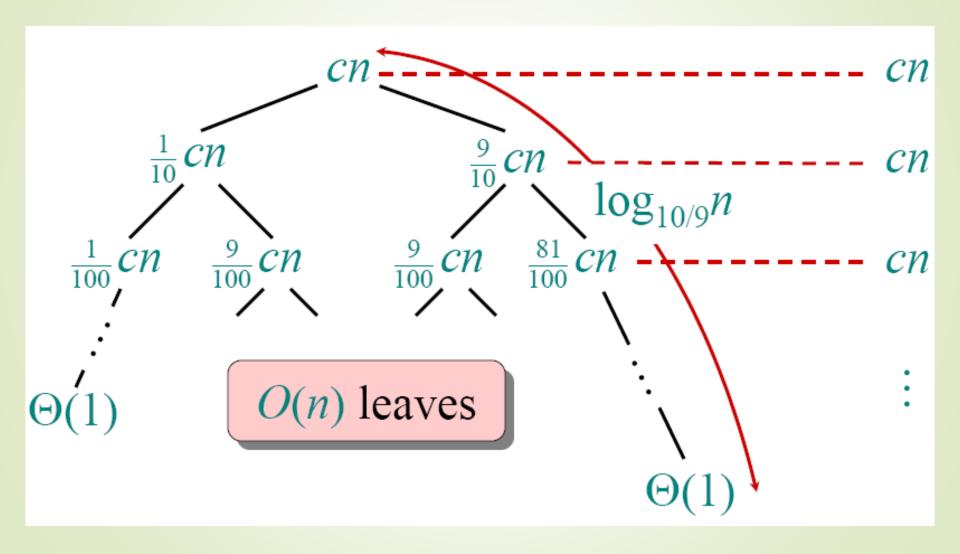
Almost Best-case

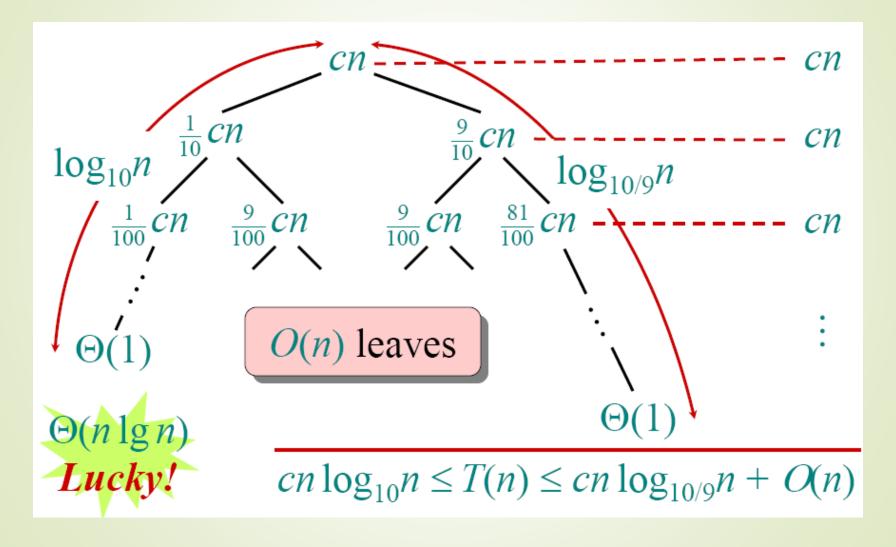
- What if the split is always 1/10: 9/10?
- T(n) = T(1/10n) + T(9/10n) + Θ(n)

T(n)









Short and Long Path Heights

- Short path node value: $n \rightarrow (1/10)n \rightarrow (1/10)^2n \rightarrow ... \rightarrow 1$
- \therefore n(1/10)^{sp} = 1
- \therefore n = 10^{sp} // take logs
- $\log_{10} n = sp$
- Long path node value: $n \rightarrow (9/10)n \rightarrow (9/10)^2n \rightarrow ... \rightarrow 1$
- : $n(9/10)^{lp} = 1$
- : $n = (10/9)^{lp}$ // take logs
- $\log_{10/9} n = lp$

Quicksort in Practice

- Quicksort is a great general-purpose sorting algorithm.
 - especially with a randomized pivot
 - Quicksort can benefit substantially from code tuning
 - Quicksort can be over twice as fast as merge sort

Quicksort behaves well even with caching and virtual memory.

Timing Comparisons

Lesson 1. Good algorithms are better than supercomputers.
Lesson 2. Great algorithms are better than good ones.

- Running time estimates:
- ► Home PC executes 10⁸ compares/second.
- Supercomputer executes 10¹² compares/second

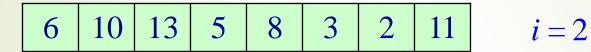
	insertion sort (N²)			mergesort (N log N)			quicksort (N log N)		
computer	thousand	million	billion	thousand	million	billion	thousand	million	billion
home	instant	2.8 hours	317 years	instant	1 second	18 min	instant	0.6 sec	12 min
super	instant	1 second	1 week	instant	instant	instant	instant	instant	instant

Quickselect

- Quickselect algorithm is used to find the i-th smallest element in a given unordered array
- Randomized algorithm using divide and conquer
- □ Similar to randomized quicksort
 - □ *Like quicksort*: Partitions input array recursively
 - □ *Unlike quicksort*: Makes a single recursive call
 - Reminder: Quicksort makes two recursive calls
- \square Expected runtime: $\Theta(n)$
 - Reminder: Expected runtime of quicksort: $\Theta(nlgn)$

Selection in Expected Linear Time: Example 1

Select the 2nd smallest element:



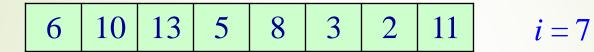
Partition the input array:



make a recursive call to select the 2nd smallest element in left subarray

Selection in Expected Linear Time: Example 2

Select the 7th smallest element:



Partition the input array:



make a recursive call to select the 4th smallest element in right subarray

Selection in Expected Linear Time

```
R-SELECT(A,p,r,i)

if p = r then

return A[p]

q \leftarrow \text{R-PARTITION}(A, p, r)

k \leftarrow q - p + 1

if i \leq k then

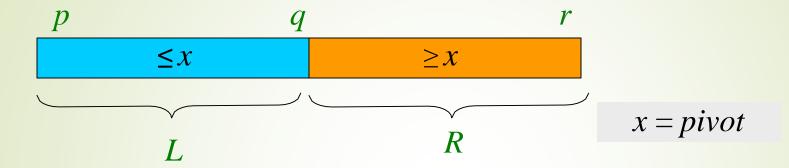
return R-SELECT(A, p, q, i)

else

return R-SELECT(A, q+1, r, i-k)
```

x = pivot

Selection in Expected Linear Time



- All elements in $L \le all$ elements in R
- L contains |L| = q-p+1 = k smallest elements of A[p...r] if $i \le |L| = k$ then

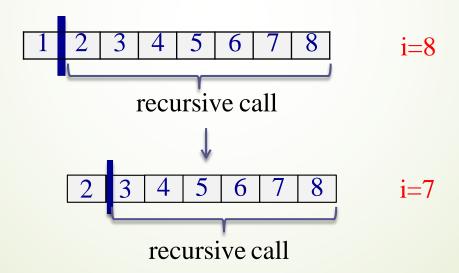
search L recursively for its *i*-th smallest element else

search R recursively for its (i-k)-th smallest element

Runtime Analysis

□ Worst case:

Imbalanced partitioning at every level and the recursive call always to the larger partition



Runtime Analysis

□ *Worst case*:

$$T(n) = T(n-1) + \Theta(n)$$

$$\rightarrow$$
 T(n) = Θ (n²)

Worse than the naïve method (based on sorting)

□ **Best case**: Balanced partitioning at every recursive level

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

$$\rightarrow$$
 T(n) = Θ (n)

□ Avg case: Expected runtime – need analysis