Unit II

DIVIDE AND CONQUER

General Method

The most-well known algorithm design strategy or technique:

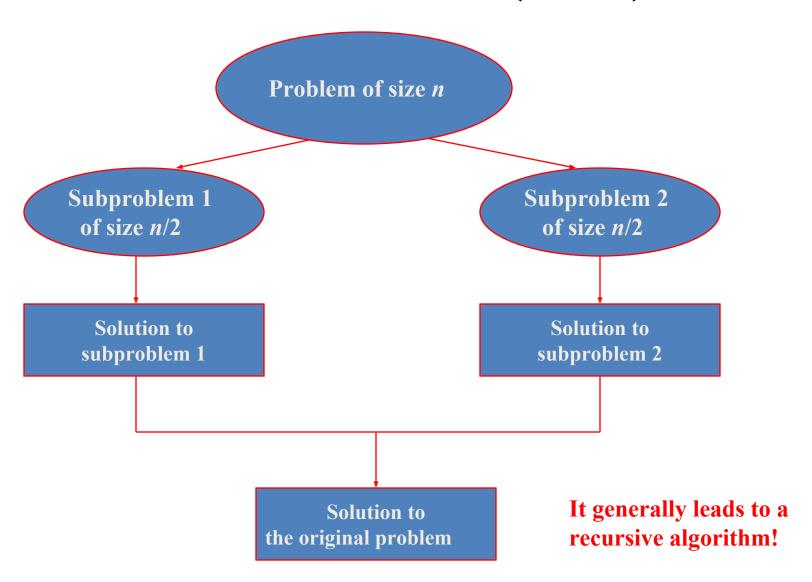
Divide and conquer.

- Break up problem into several parts.
- Solve each part recursively.
- Combine solutions of sub-problems to obtain solution of the original problem

Most common usage.

- Break up problem of size n into two equal parts of size n/2.
- Solve two parts recursively.
- Combine two solutions into overall solution in linear time.

General Method(cont.)



Recurrence Equation for Divide and Conquer

• If the size of problem 'P' is n and the sizes of the 'k' sub problems are n_1 , n_2 n_k , respectively, then the computing time of divide and conquer is described by the recurrence relation

$$T(n) = \begin{cases} g(n) & n \text{ small} \\ T(n_1) + T(n_2) + \dots + T(n_k) + f(n) & \text{otherwise} \end{cases}$$

- Where,
 - T(n) is the time for divide and conquer method on any input of size n and
 - g(n) is the time to compute answer directly for small inputs.
 - $T(n_1), T(n_2)...T(n_k)$ is the time taken to solve sub problems of size $n_1, n_2...n_k$ respectively
 - The function f(n) is the time for dividing the problem 'P' and combining the solutions of subproblems

Recurrence Equation for Divide and Conquer [Contd..]

- ☐ More generally, an instance of size n can be divided into a instances of size n/b, with a of them needing to be solved. (Here, a and b are constants; a>=1 and b > 1.)
- Assuming that size \mathbf{n} is a power of \mathbf{b} (i.e. $n = b^k$), to simplify our analysis, we get the following recurrence for the running time T(n):

$$T(n) = \begin{cases} T(1) & n=1\\ aT(n/b) + f(n) & n>1 \end{cases}$$

Where f(n) is a function that accounts for the time spent on dividing the problem into smaller ones and on combining their solutions.

Recurrence Equation for Divide and Conquer [Contd..]

- **Substitution Method** One of the methods for solving the recurrence relation is called the substitution method. This method repeatedly makes substitution for each occurrence of the function T in the right hand side until all such occurrences disappear.
- **Master Theorem** The efficiency analysis of many divide-and-conquer algorithms is greatly simplified by the master theorem.
 - It states that, in recurrence equation T(n) = aT(n/b) + f(n),

$$T(n) \in \begin{cases} \Theta(n^d) & \text{if } a < b^d, \\ \Theta(n^d \log n) & \text{if } a = b^d, \\ \Theta(n^{\log_b a}) & \text{if } a > b^d. \end{cases}$$

Analogous results hold for the O and Ω notations, too.

Merge Sort: Design and Implementation

MergeSort

- ☐ Merge sort is a perfect example of a successful application of the divide-and conquer technique for sorting purpose.
- ☐ It sorts a given array A[1],...,A[n]
 - By dividing it into two halves

$$A[1],...,A[n/2]$$
 and $A[n/2+1],...,A[n],$

- ✓ Sorting each of them recursively,
- ✓ And then merging the two smaller sorted arrays into a single sorted one.

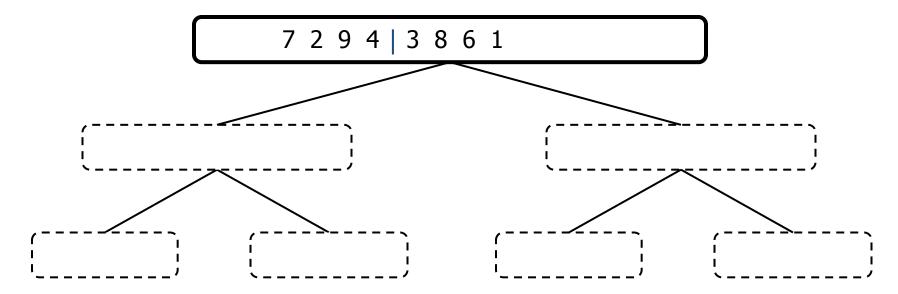
Algorithm for MergeSort

Algorithm Mergesort (L, low, high)

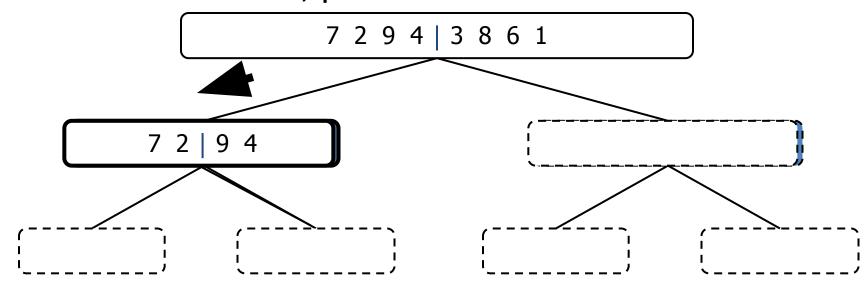
```
// L is list of elements to be sorted, low -> first element, high -> last element
    if the list has two elements then
         Compare and sort it
    else
        // Divide L into two sub problems
        mid:=(low+high)/2;
         // Solve the sub problems recursively
          Mergesort (L,low, mid); // sorts the first half sub-array L[low:mid]
         Mergesort (L,mid+1,high); //sorts the second half sub-array L[mid+1:high]
          // combine solutions
          Merge (L,low, mid, high);
```

Execution Example

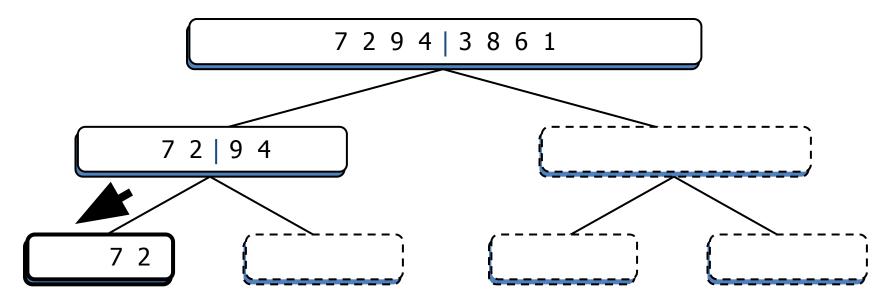
• Initial Call, Partition



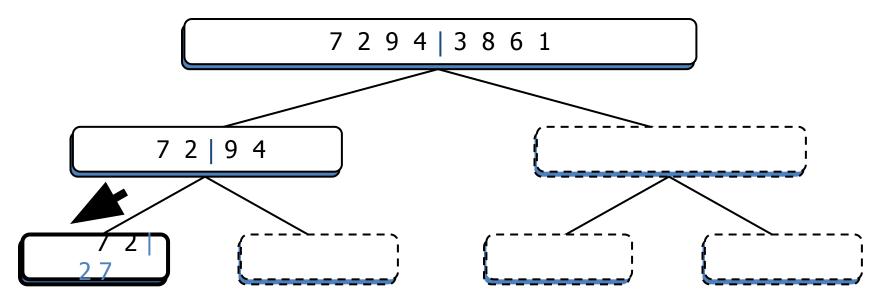
Recursive call, partition



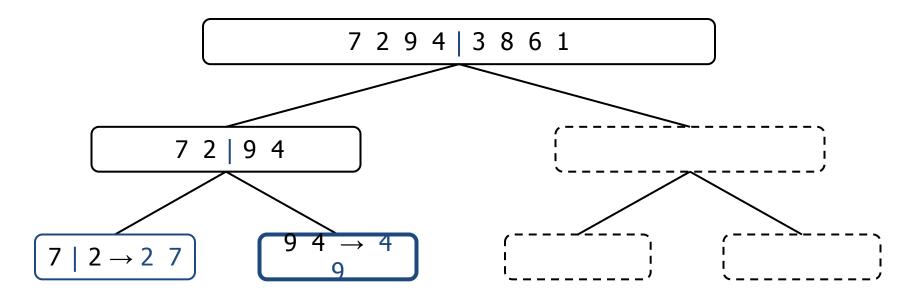
Recursive call, partition



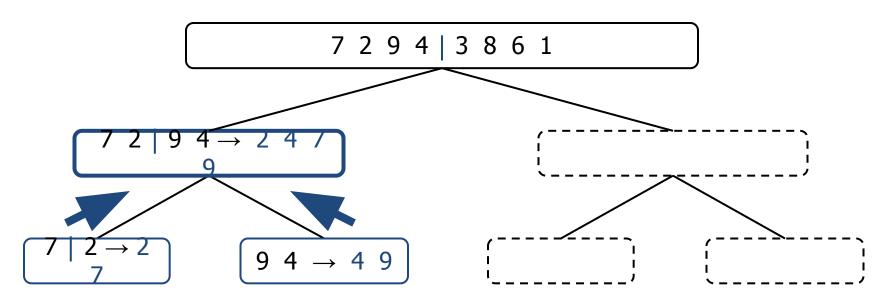
Recursive call, base case



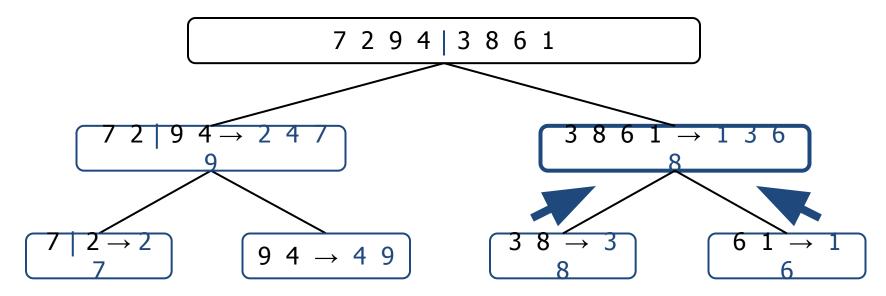
Recursive call, base case



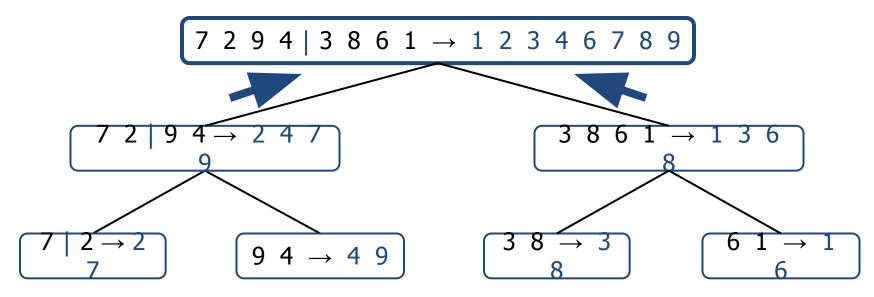
Merge



• Recursive call, ..., merge



Merge



Start with two sorted sets of values

```
a: 3 7 8 19 24 25
```

b: 2 5 6 10

c:

```
a: 3 7 8 19 24 25
b: _ 5 6 10
c: 2
```

```
a: _ 7 8 19 24 25
b: _ 5 6 10
```

```
a: _ 7 8 19 24 25
b: _ 6 10
```

```
a: _ 7 8 19 24 25
b: _ _ _ 10
c: 2 3 5 6
```

```
a: _ _ 8 19 24 25
b: _ _ _ 10
c: 2 3 5 6 7
```

```
a: _ _ _ 19 24 25
b: _ _ _ 10
c: 2 3 5 6 7 8
```

• Merge

```
a: _ _ _ 19 24 25
b: _ _ _ _ _ _
```

Second sub-array exhausted

```
a: _ _ _ _ 24 25
b: _ _ _ _ _ _
c: 2 3 5 6 7 8 10 19
```

```
a: _ _ _ _ Copied a-values to c
b: _ _ _ _ _
c: 2 3 5 6 7 8 10 19 24 25
```

Algorithm for Merge

```
Algorithm Merge (a,low, mid, high)
// a (low: high) is an array containing two sorted subsets in a (low: mid) //and in
a(mid+1, high). The goal is to merge these two subsets in to a single set residing in
a(low: high)
// b[] is a temporary global array.
    h:=low, i:=low; j=mid+1;
    while ((h \le mid) and (j \le high)) do
          if (a[h] \le a[j]) then
              b[i] = a[h];
              h := h+1;
         else
              b[i]:=a[j];
              j:=j+1;
         i = i + 1;
```

Algorithm for Merge

```
if (h>mid) then // ???
for k=j to high do
     b[i] := a[k];
      i: = i+1;
else // ???
for k:=h to mid do
  b[i]:=a[k];
   i:=i+1;
for k: = low to high do
     a[k] := b[k];
```

Analysis

- Input Size: n- elements in the input array
- Here the basic operation is the **comparisons made**.
- No best, worst and average case.
- Let T(n) denote the number of comparisons made to sort n elements, Assuming n is a even number of elements
- Hence,

$$T(n)=T(n/2)+T(n/2)+C_{merge}(n)$$
 for $n>2$

$$T(2) = 1$$
 for $n=2$, $T(1)=0$ for $n=1$

Where T(n/2)= Time taken to sort n/2 subproblem and $C_{merge}(n)$ =No of comparisons made to merge 2 sorted Subproblems.

Best Case Occurs When all the elements in a subproblem is smaller than the other subproblem

10 22 33

56 87 92 97

If

- ■Size of I subproblem is m=3
- ■Size of II problem is n=4
- ■No of comparisons made =3 =m

34 45 56 67

10 22 33

If

- ■Size of I subproblem is m=4
- ■Size of II problem is n=3
- ■No of comparisons made =3 =n

Worst Case Occurs When smaller elements may come from the alternating subproblems

34 45 56 67

If

- ■Size of I subproblem is m=4
- ■Size of II problem is n=3
- ■No of comparisons made =6 =m+n-1

42 54 65

In MergeSort, $C_{merge}(n)=n/2+n/2-1=n-1$

Analysis

• Hence, $T(n) \leq T(n/2) + T(n/2) + C_{merge}(n) \text{ for } n > 2$ $T(n) \leq 2T(n/2) + (n-1) \text{ for } n > 2$ $T(n) \leq 2T(n/2) + O(n) \text{ for } n > 2$ $T(n) \leq 2T(n/2) + cn \text{ for } n > 2$ and $T(2) \leq c \text{ for } n = 2$

Solving the recurrence equation using master theorem:

- Here a = 2, b = 2, f(n) = n, d = 1.
- Therefore $2 = 2^1$, case 2 holds in the master theorem
- $T(n) \subseteq \Theta$ (n^d log n)= Θ (n¹ log n) = Θ (n log n)
- Therefore $T(n) \subseteq \Theta$ (n log n)

Analysis

Two ways of Analyzing Recursion:

- Unrolling the Recursion method
- Substituting a solution method

Refer to class notes/ Text book1 page no-212 and 213

Comparison between Sorting Algorithms

	Worst Case	Average Case	Best Case
Bubble Sort	$\mathcal{O}(n^2)$	$\mathcal{O}(n^2)$	$\mathcal{O}(n)$
Insertion Sort	$\mathcal{O}(n^2)$	$\mathcal{O}(n^2)$	$\mathcal{O}(n)$
Merge Sort	$\mathcal{O}(n \log n)$	$\mathcal{O}(n \log n)$	$O(n \log n)$
Heap Sort	$\mathcal{O}(n \log n)$	$\mathcal{O}(n \log n)$	$O(n \log n)$
Quick Sort	$O(n^2)$	$O(n \log n)$	$O(n \log n)$

QuickSort - Introduction

- Quicksort is the other important sorting algorithm that is based on the divide and conquer approach.
- Unlike mergesort, which divides its input elements according to their position in the array, quicksort divides them according to their value.
- A partition is an arrangement of the array's elements so that all the elements to the left of some element A[s] are less than or equal to A[s], and all the elements to the right of A[s] are greater than or equal to it

$$\underbrace{A[0]\dots A[s-1]}_{\text{all are } \leq A[s]} A[s] \underbrace{A[s+1]\dots A[n-1]}_{\text{all are } \geq A[s]}$$

QuickSort-Introduction

$$\underbrace{A[0]...A[s-1]}_{\text{all are } \leq A[s]} A[s] \underbrace{A[s+1]...A[n-1]}_{\text{all are } \geq A[s]}$$

- Obviously, after a partition is achieved, A[s] will be in its final position in the sorted array, and we can continue sorting the two subarrays to the left and to the right of A[s] independently (e.g., by the same method).
- Note the difference with mergesort: there, the division of the problem into two subproblems is immediate and the entire work happens in combining their solutions; here, the entire work happens in the division stage, with no work required to combine the solutions to the subproblems.

Algorithm for QuickSort

```
ALGORITHM Quicksort(A[l..r])
    //Sorts a subarray by quicksort
    //Input: Subarray of array A[0..n-1], defined by its left and right
            indices l and r
    //Output: Subarray A[l..r] sorted in nondecreasing order
    if l < r
        s \leftarrow Partition(A[l..r]) //s is a split position
        Quicksort(A[l..s-1])
        Quicksort(A[s+1..r])
```

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:

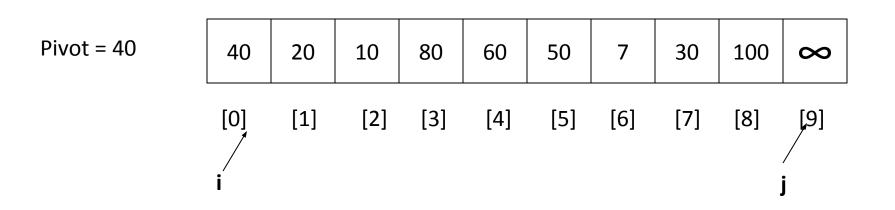
40 20 10 80 60 50 7 30 10

Partitioning Array

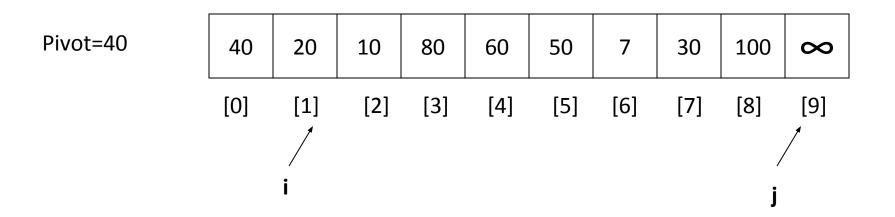
Given a pivot, partition the elements of the array such that the resulting array consists of:

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

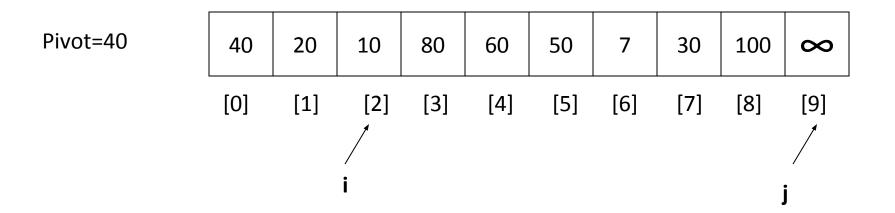
The sub-arrays are stored in the original data array.



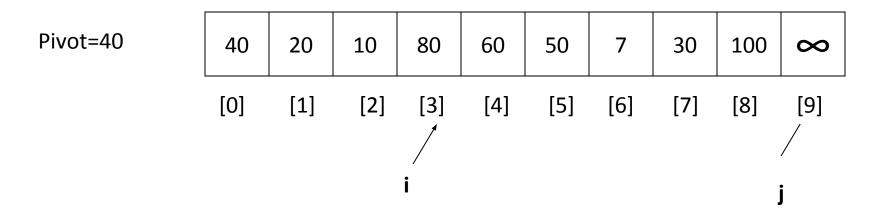
1. Repeat i++ until a[i]>=pivot



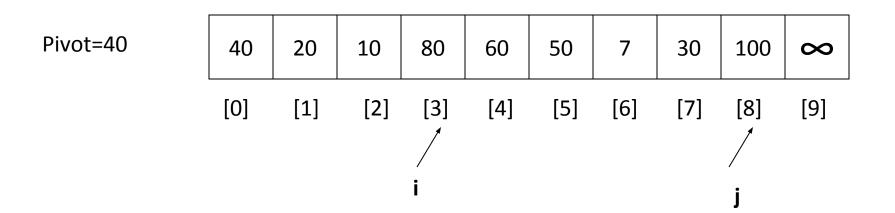
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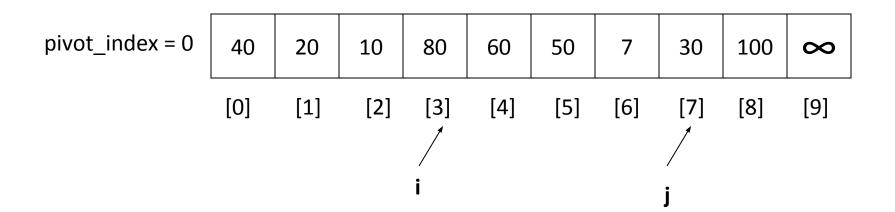
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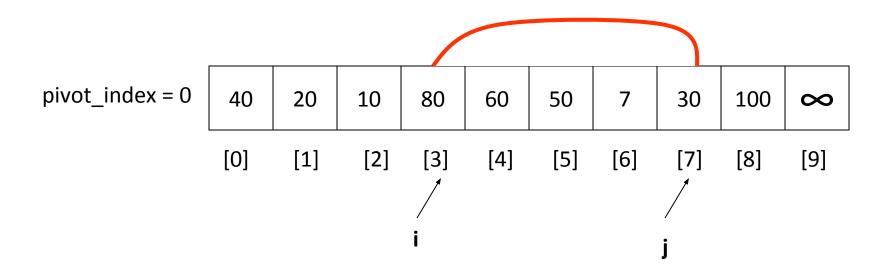
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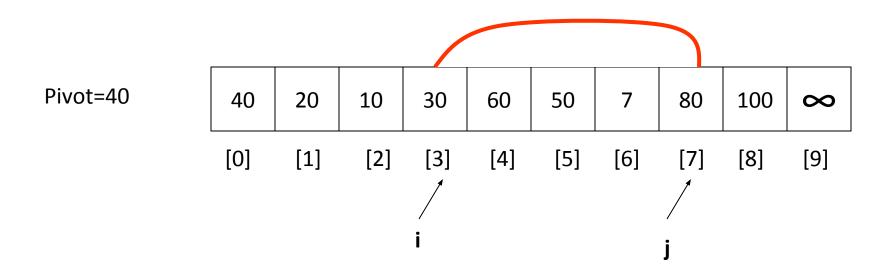
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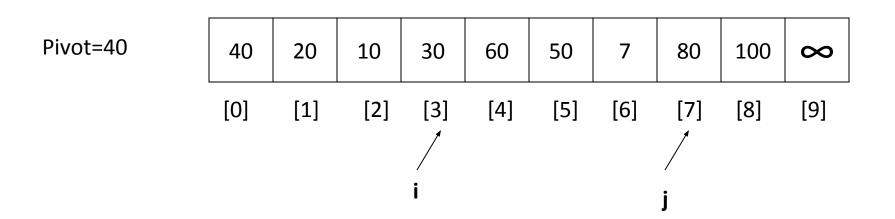
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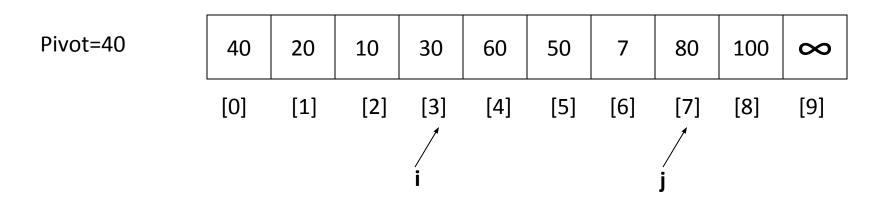
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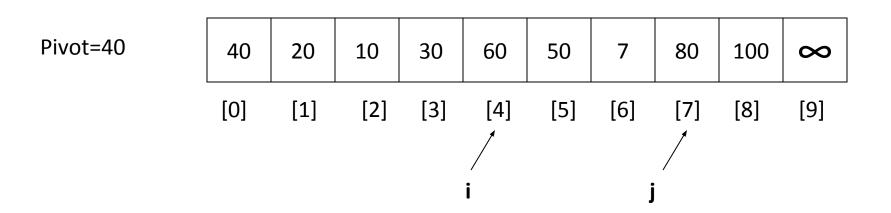
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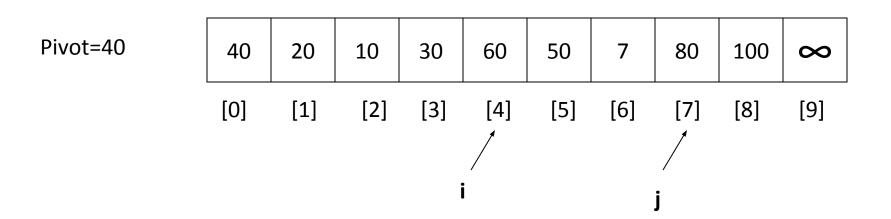
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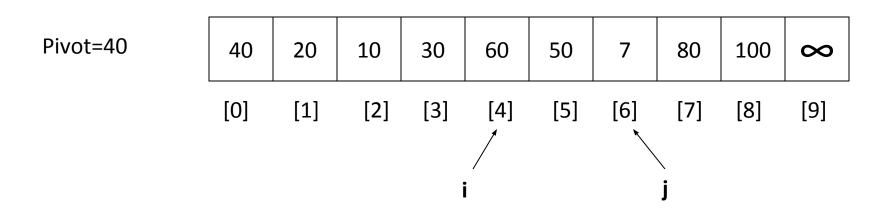
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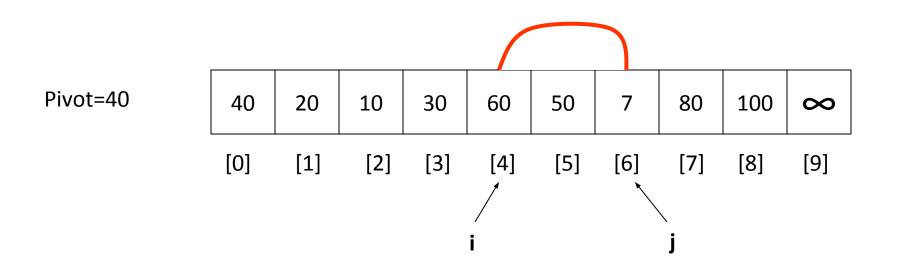
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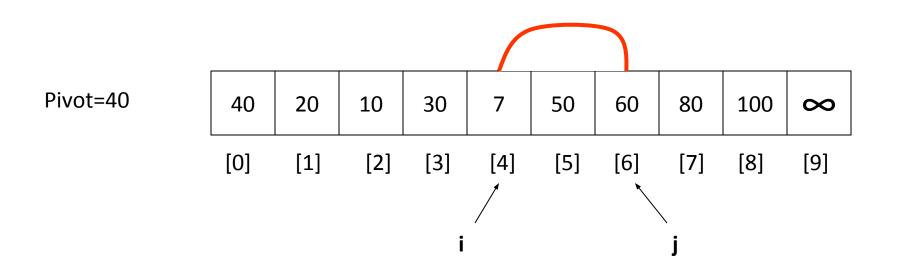
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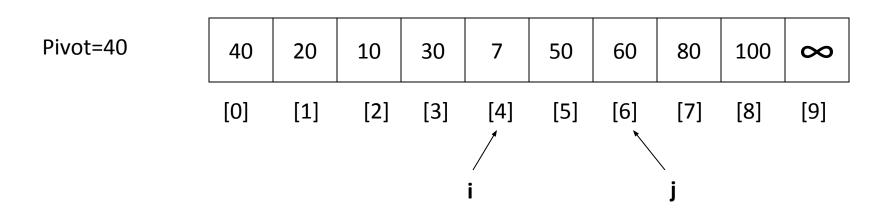
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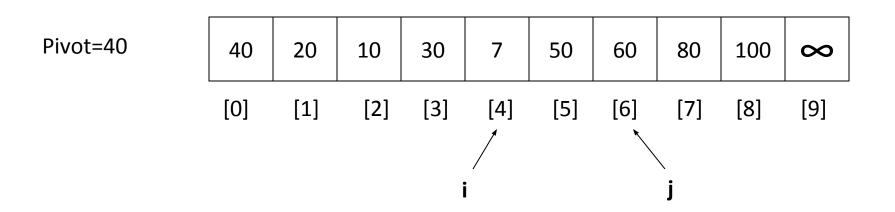
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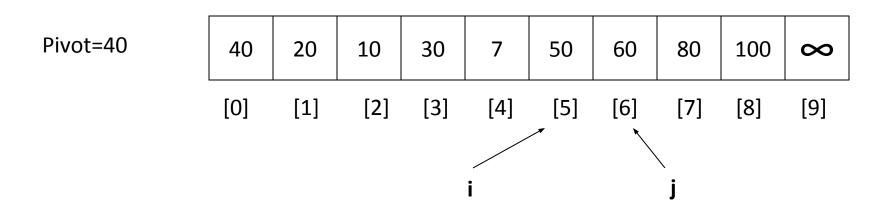
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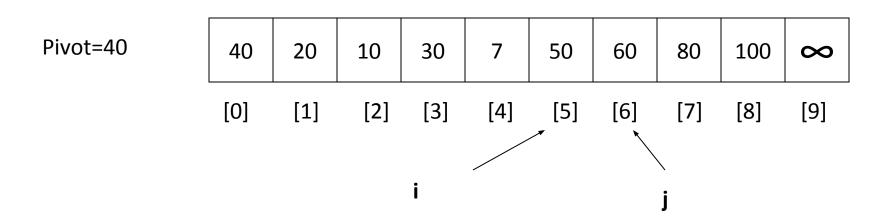
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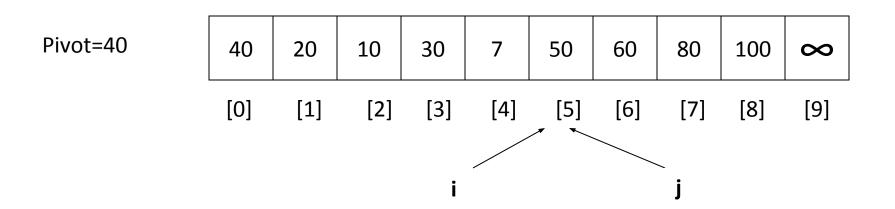
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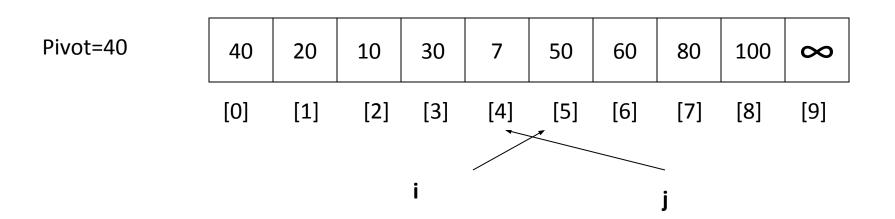
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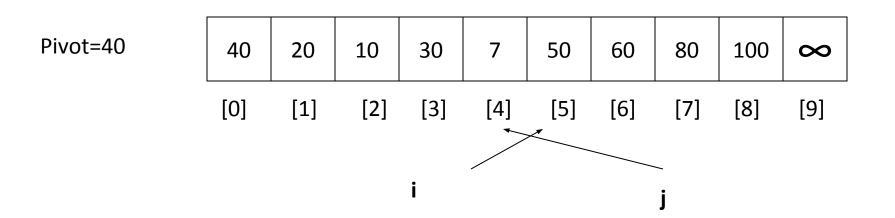
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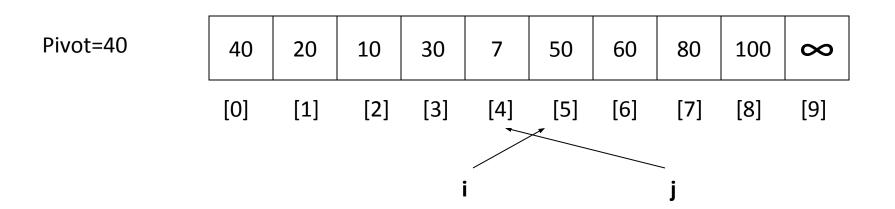
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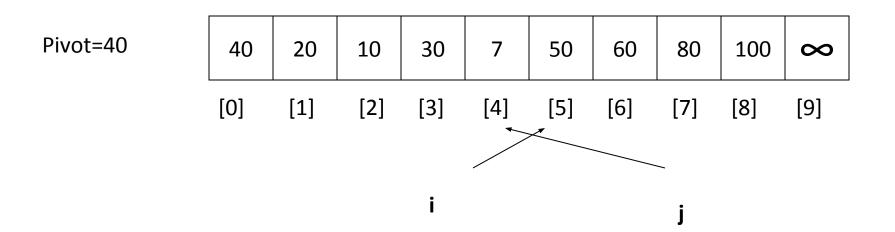
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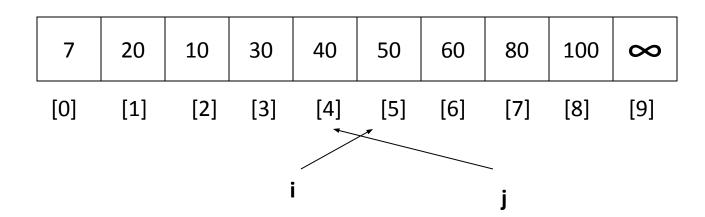


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- 5. If i>=j then swap a[low] and a[j]

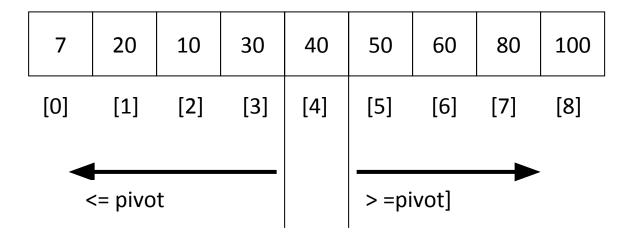


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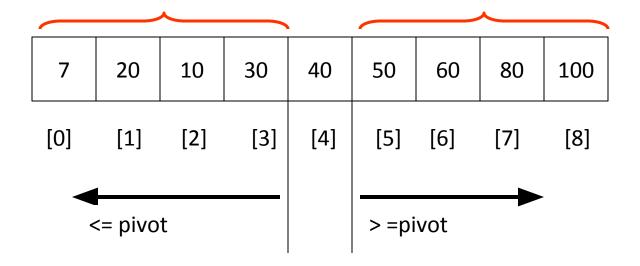
Pivot=40 and returns 4



Partition Result



Recursion: Quicksort Sub-arrays



Algorithm for Partition

```
Alg. PARTITION (A[l...r])
     pivot\leftarrow A[I]
      i \leftarrow l
     i \leftarrow r + 1
     repeat
                repeat
                   i \leftarrow i + 1
                 until A[i] \ge pivot
           repeat
             j \leftarrow j - 1
                         until A[j] \leq pivot
           if i < j
                      then exchange A[i] \leftrightarrow A[j]
      until i>=j
      exchange A/I \rightarrow A[j]
return j
```

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

- Assume that keys are random, uniformly distributed.
- What is best case running time?
 - Recursion:
 - 1. Partition splits array in two sub-arrays of size n/2
 - 2. Quicksort applied recursively to each sub-array
- The recurrence relation to compute time complexity

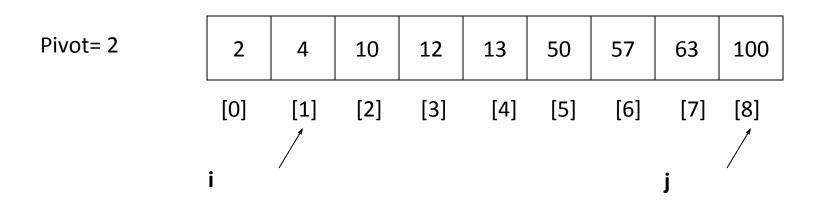
$$C_{best}(n) = 2C_{best}(n/2) + n$$
 for $n > 1$, $C_{best}(1) = 0$.

- Assume that keys are random, uniformly distributed.
- Best case running time: $\Omega(n \log_2 n)$

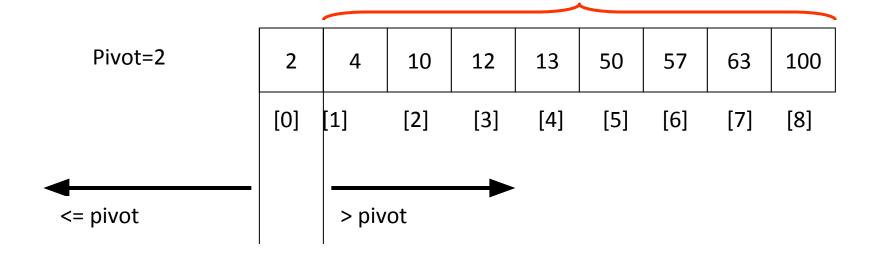
- Assume that keys are random, uniformly distributed.
- Best case running time: $\Omega(n \log_2 n)$
- Worst case running time?

Quicksort: Worst Case

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



- 1. Repeat i++ until a[i]>=pivot
- 2. Repeat j-- until a[j]<=pivot
- 3. If i<j swap a[i] and a[j]
- 4. if i < j, go to 1.
- 5. If i>=j then swap a[low] and a[j]



- Assume that keys are random, uniformly distributed.
- Best case running time: Ω(n log₂n)
- Worst case running time?
 - Recursion:
 - 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.
 - The recurrence relation to compute time complexity T(n) =T(n-1)+T(0)+cn

The recurrence relation to compute time complexity

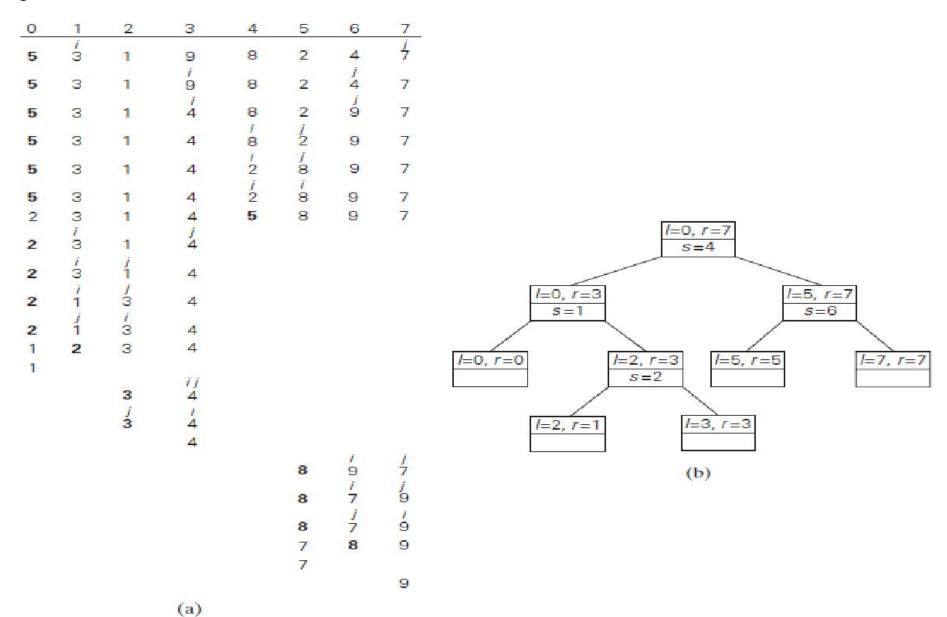
$$T(n) = T(n-1) + T(0) + n$$

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

Solving the recurrence relation using backward substitution method will result to $T(n)=O(n^2)$

- Assume that keys are random, uniformly distributed.
- Best case running time: $\Omega(n \log_2 n)$
- Worst case running time: O(n²)

Example of quicksort operation. (a) Array's transformations with pivots shown in bold. (b) Tree of recursive calls to Quicksort with input values l and r of subarray bounds and split position s of a partition obtained



0	1	2	3	4	5	6	7	
	i						j	I=0,r=7
5	3	1	9	8	2	4	7	1 0,1 7
			i			j		s=4
5	3	1	9	8	2	4	7	
			i			j		
5	3	1	4	8	2	9	7	
	ì	1	1	i i	j	ì	1	
5	3	1	4	8	2	9	7	
				i	j			
5	3	1	4	2	8	9	7	s=1 s=6
				j	i			
5	3	1	4	2	8	9	7	
2	3	1	4	5	8	9	7	
			-					
	i		j				i	
2	i 3	1	j 4			i	j	s=2 I=5,r=5 I=7,r=7
2		1 j	j		8	i 9	j 7	s=2 I=5,r=5 I=7,r=7
2	3	1 j	j		8	i 9 i	j 7 j	s=2 I=5,r=5 I=7,r=7
2	3 i	j	j 4			i 9	j	s=2
	3 i 3	j 1	j 4		8	i 9 i 7	j 7 j	s=2 l=5,r=5 l=7,r=7
2	3 i 3 i 1	j 1 j 3	j 4 4		8	i 9 i 7	j 7 j 9	s=2
2 2 2	3 i 3 i 1 j	j 1 j 3 i	j 4 4 4		8 8 8 7	i 9 i 7 j	j 7 j 9 i 9	s=2
2	3 i 3 i 1	j 1 j 3	j 4 4		8 8	i 9 i 7 j	j 7 j 9 i	s=2
2 2 2	3 i 3 i 1 j	j 1 j 3 i	j 4 4 4		8 8 8 7	i 9 i 7 j	j 7 j 9 i 9	s=2
2 2 2 1	3 i 3 i 1 j	j 1 j 3 i	j 4 4 4 4		8 8 8 7	i 9 i 7 j	j 7 j 9 i 9	s=2
2 2 2 1	3 i 3 i 1 j	j 1 j 3 i	j 4 4 4		8 8 8 7	i 9 i 7 j	j 7 j 9 i 9	s=2