

### Data Structures (15B11CI311)

Odd Semester 2020



3<sup>rd</sup> Semester, Computer Science and Engineering

Jaypee Institute Of Information Technology (JIIT), Noida



# Sorting Techniques in Data Structure



#### Outlines:

- Divide and Conquer
- Merge Sort
- Quick Sort
- Radix Sort
- Count Sort
- Bucket Sort



## Divide-and-Conquer

Divide and conquer is a strategy to solving a large problem by:

- Divide the problem into a number of sub-problems
  - Similar sub-problems of smaller size
- Conquer the sub-problems
  - Solve the sub-problems recursively
  - Sub-problem size small enough solve the problems in straight forward manner
- Combine the solutions of the sub-problems
  - Obtain the solution for the original problem



#### Merge Sort

- Merge sort is a one of the efficient sorting algorithm.
- Merge sort work on a principle of divide-and-conquer algorithm.
- Merge sort repeatedly breaks down an array into several sub array until each sub array consists of a single element
- Merge the sub-problem solutions together:
  - Compare the sub-array's first elements
  - Remove the smallest element and put it into the result array
  - Continue the process until all elements have been put into the result array

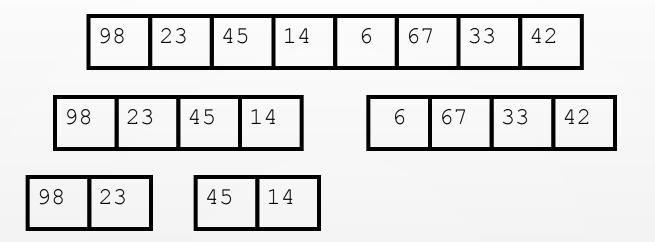
Illustration of Merge Sort



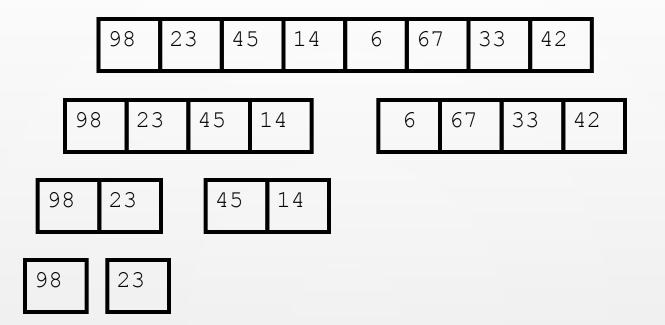


9	8	23	45	14	6	)	67	33	3 4	12	
98	23	4.5	5 14	1		(	ó (	67	33	4.	2

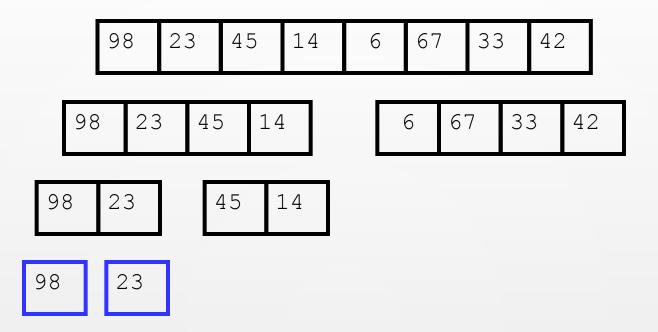






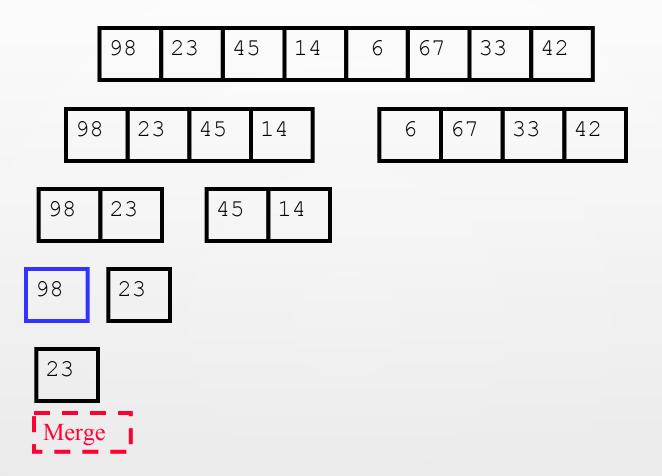




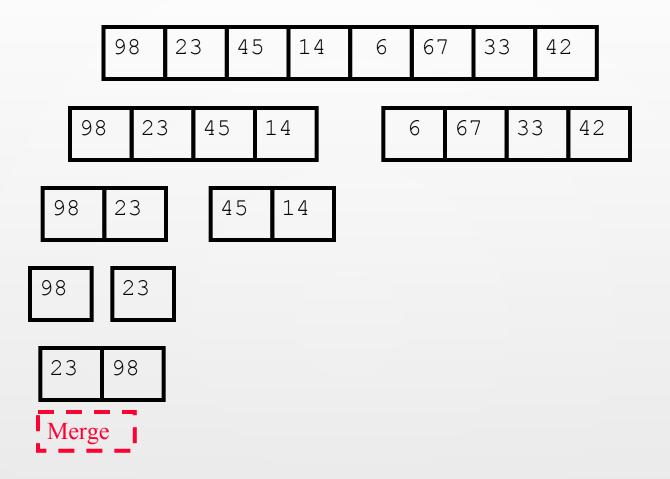




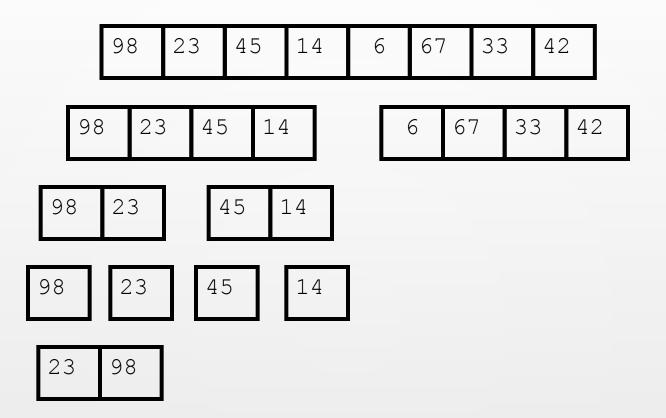




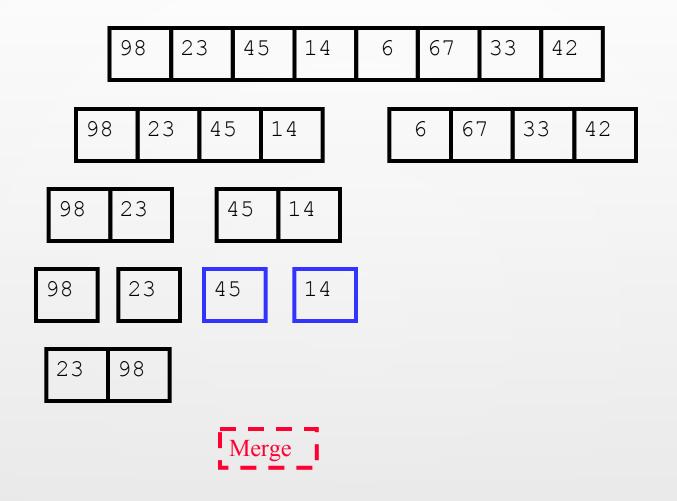




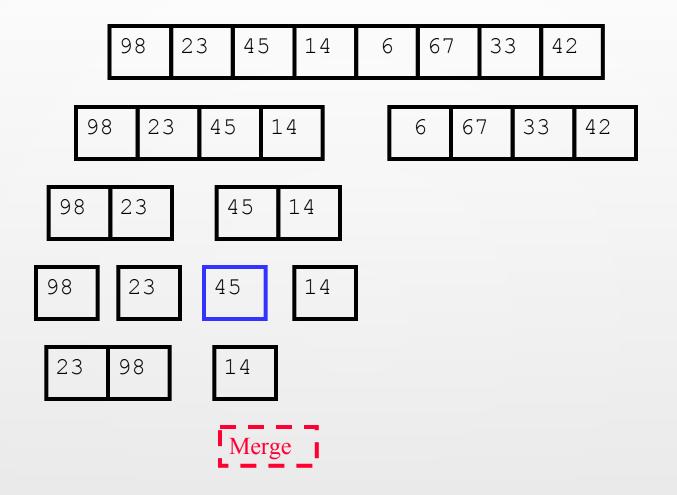




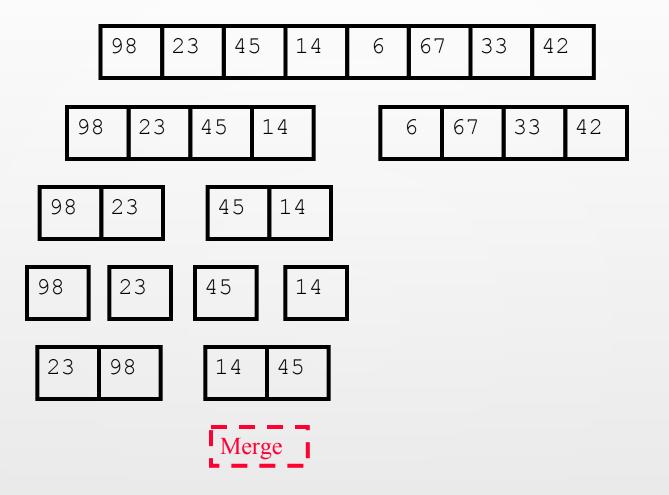








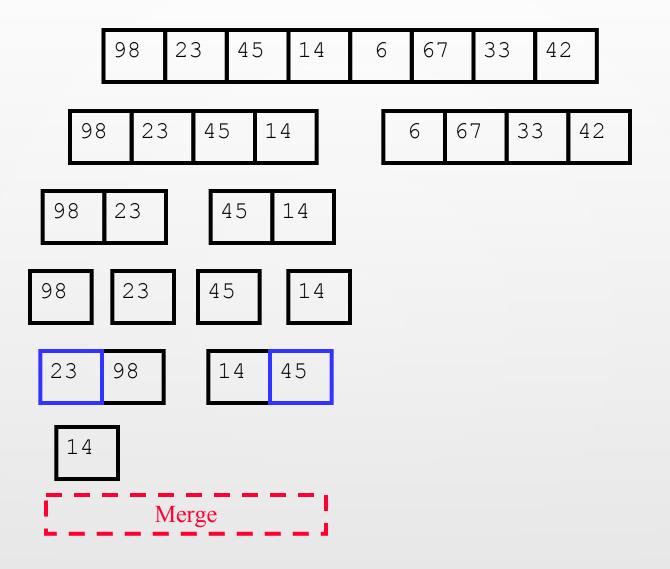




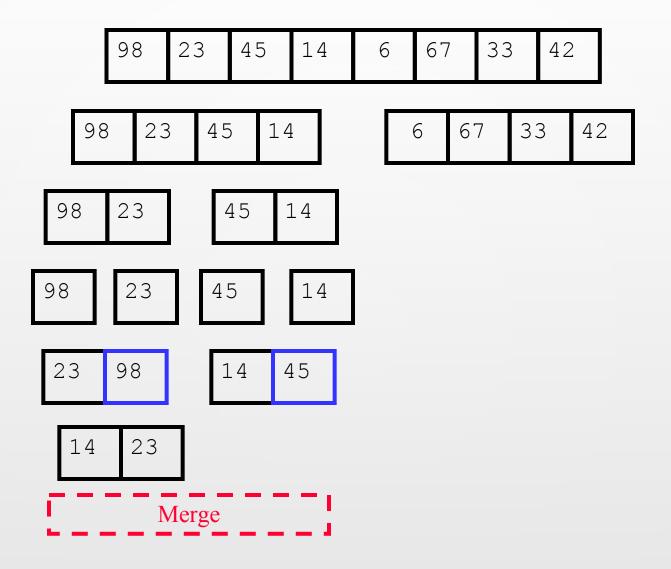




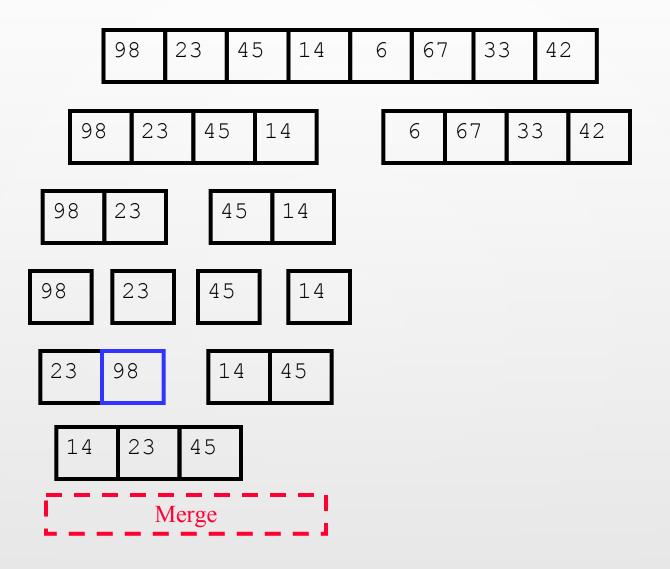








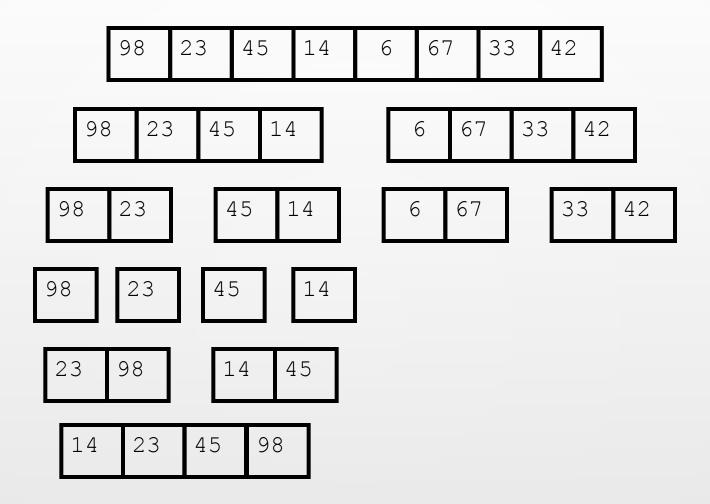




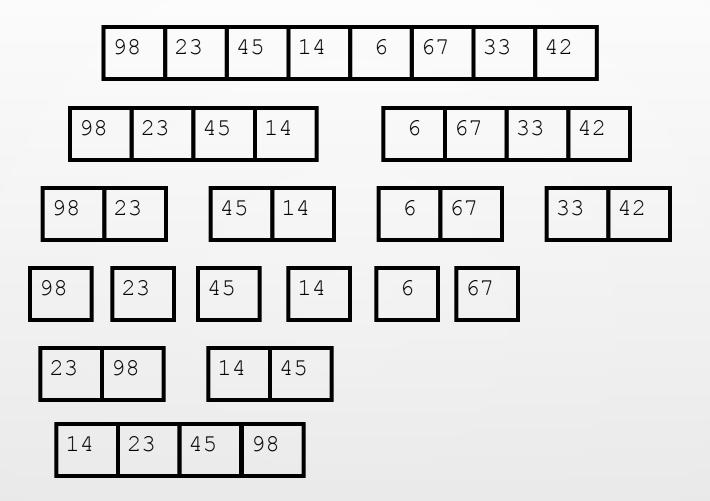




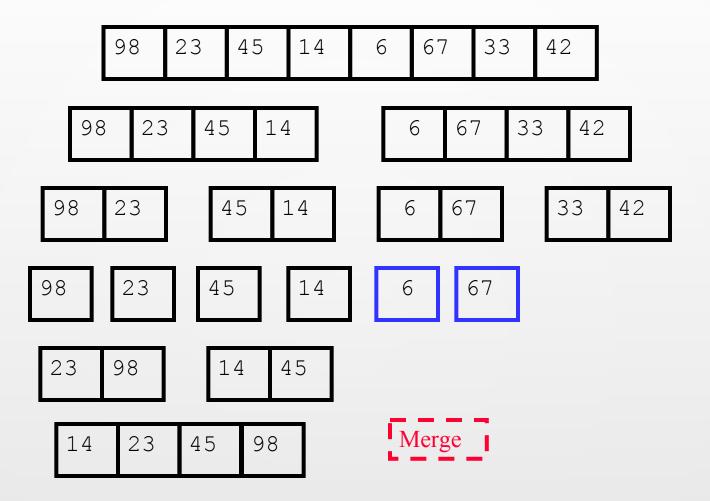




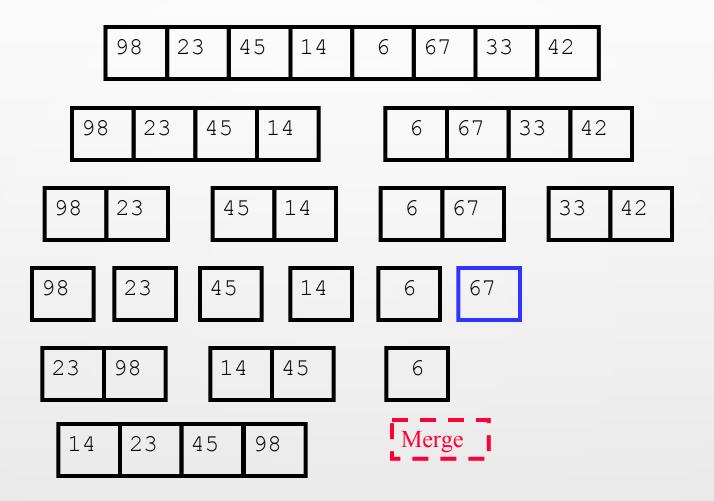




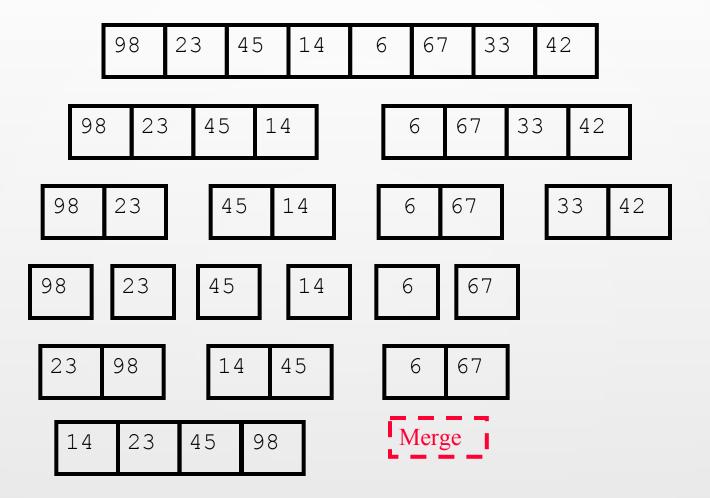




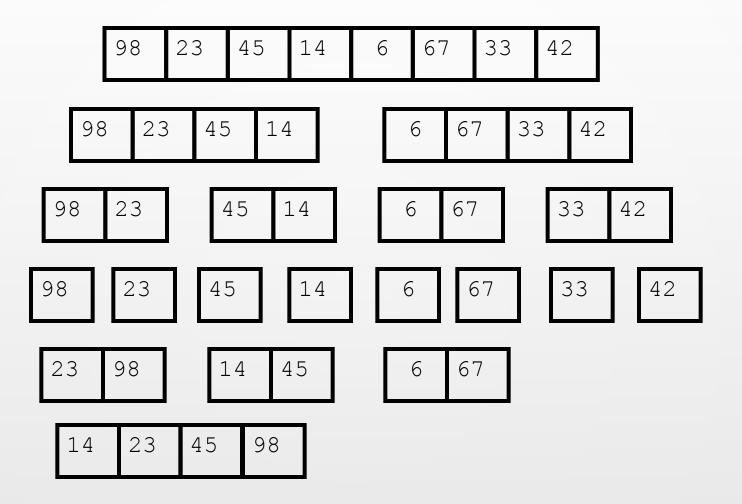








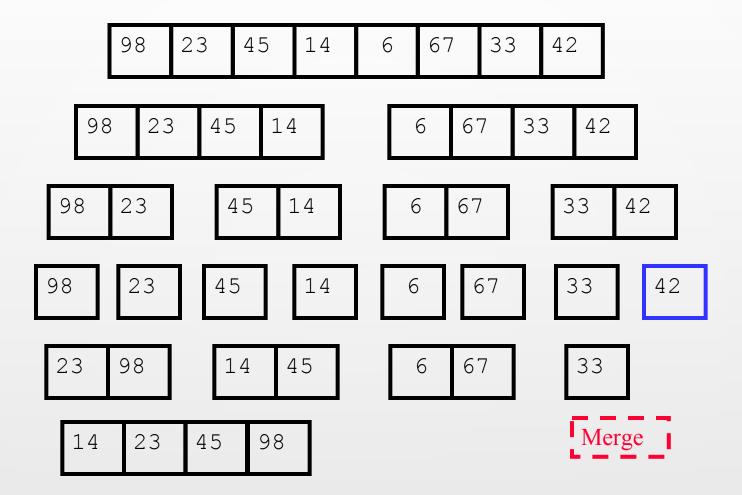




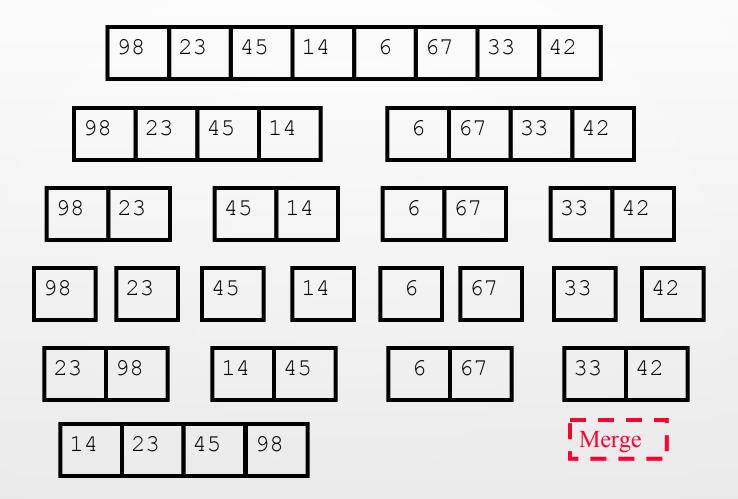








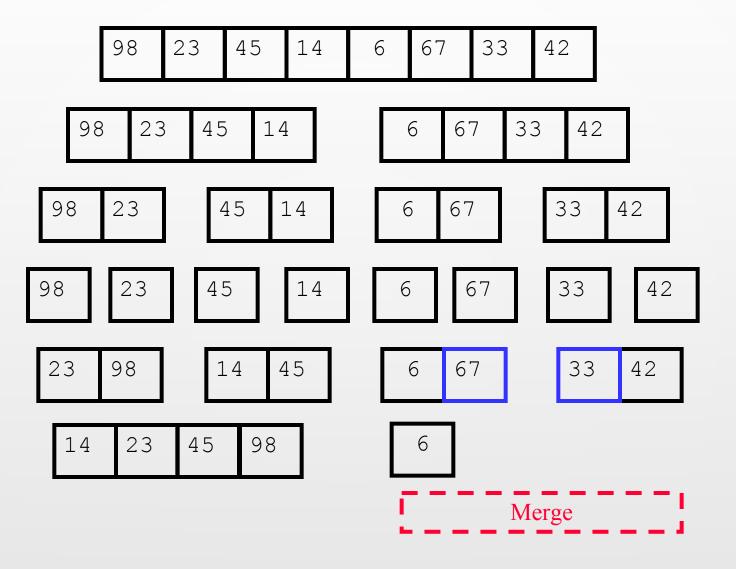




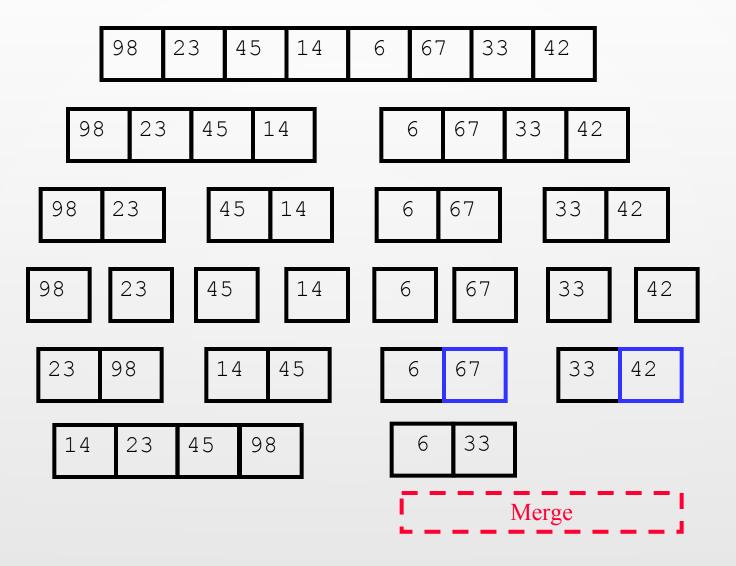












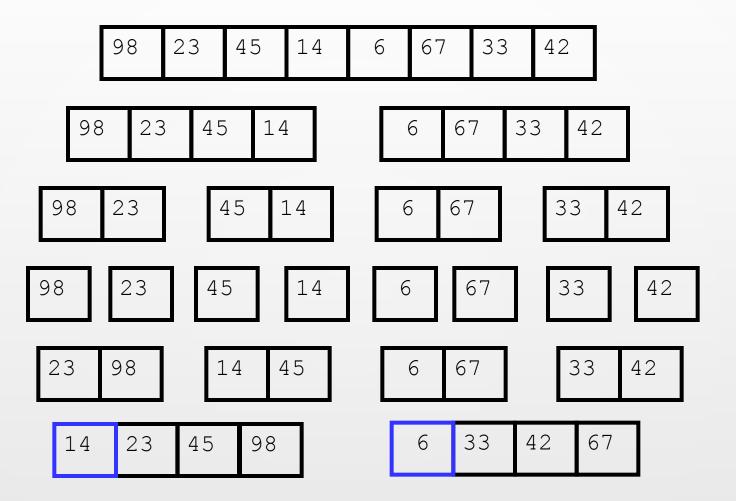






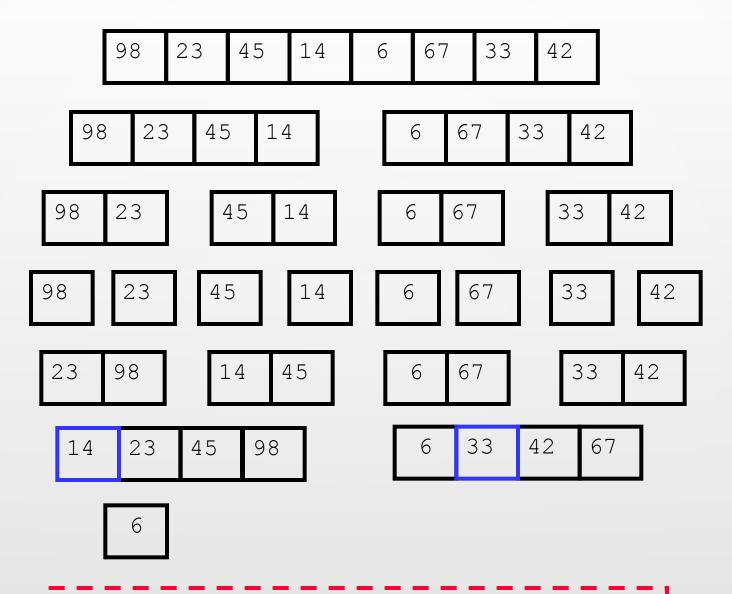




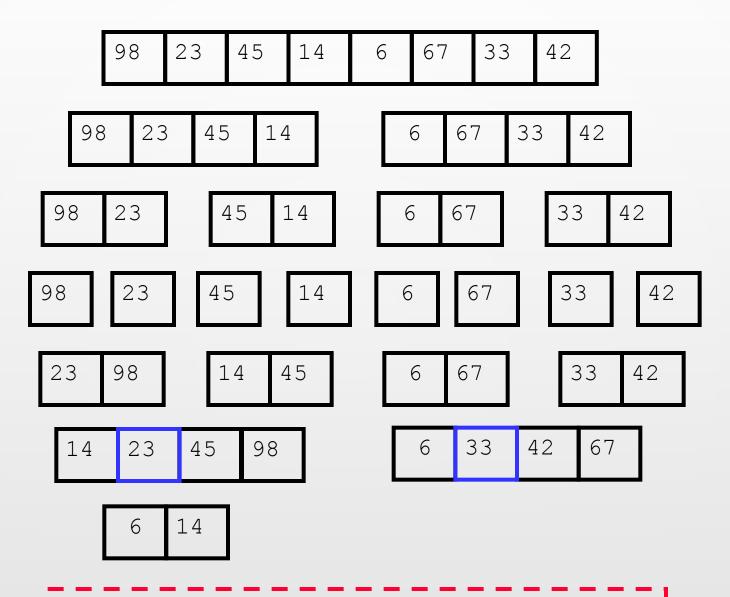


Merge

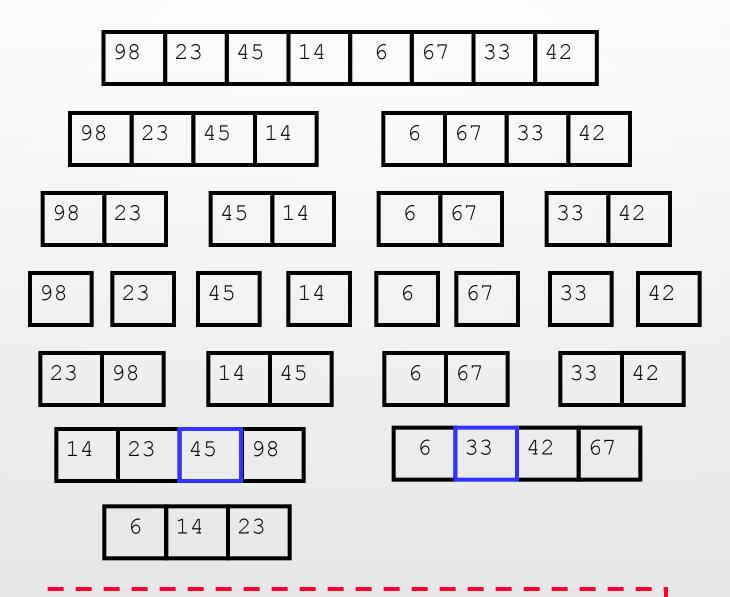




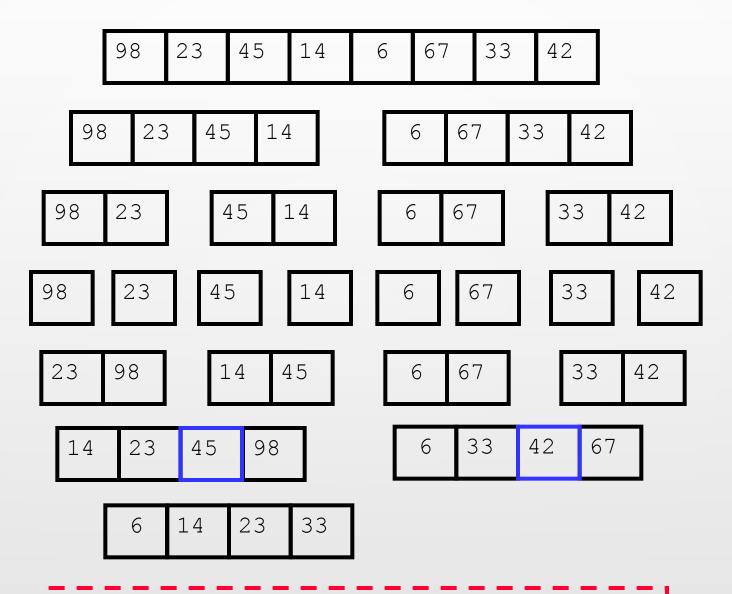




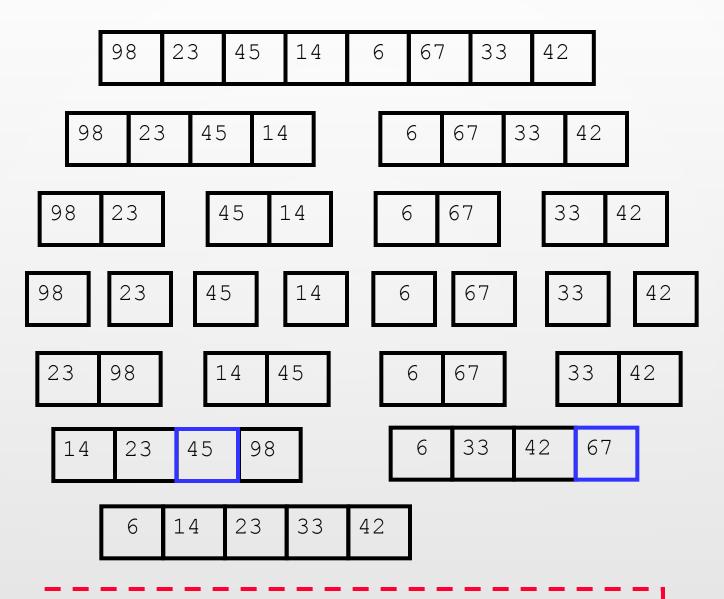




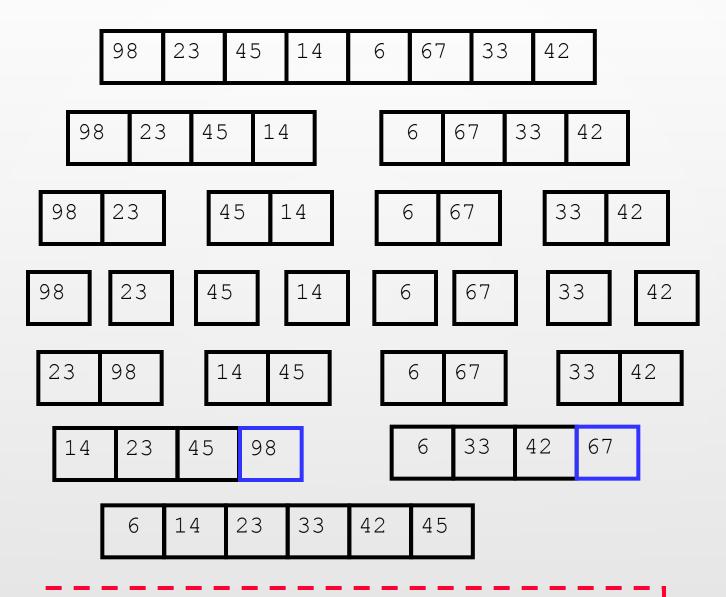




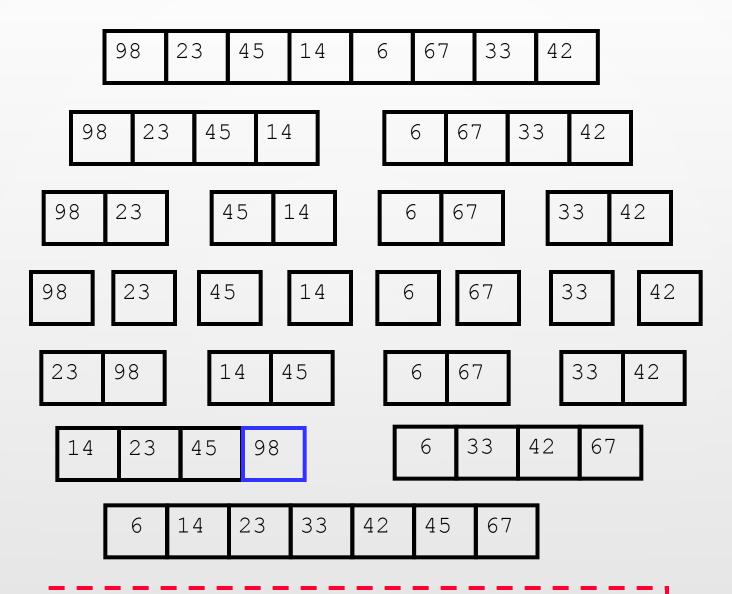




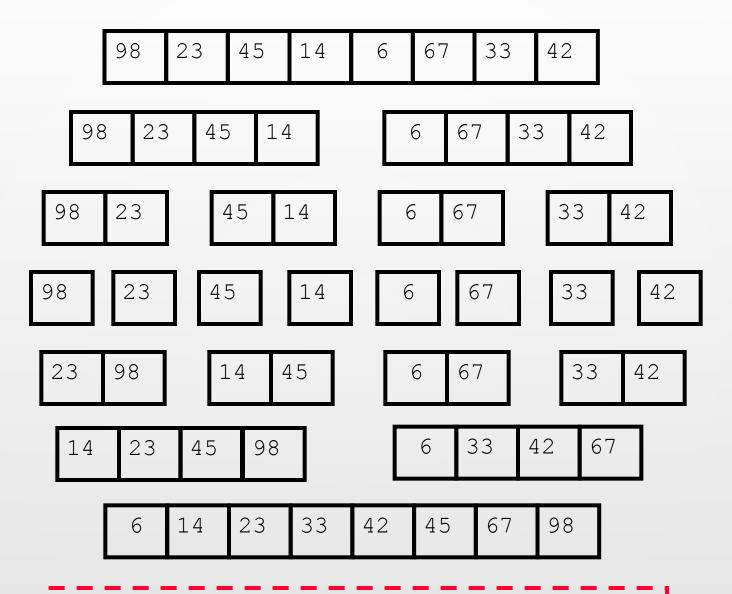




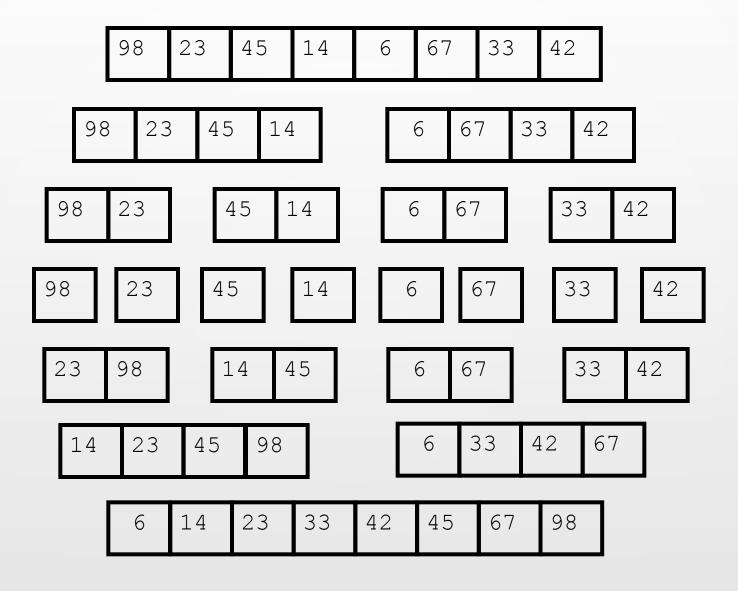




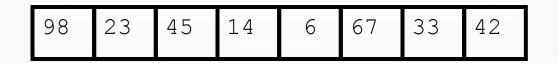














6 14 23 33 42 45 67 98





**INPUT**: a sequence of n numbers stored in array A

OUTPUT: an sorted sequence of n numbers

l points first element index of array and r points the index of last element

**Initial Call:** MergeSort(A, 1, n)

#### Merge(A, p, q, r)

# Procedure Merge



```
n_1 \leftarrow q - p + 1
  n_2 \leftarrow r - q
for i \leftarrow 1 to n_1
    do L[i] \leftarrow A[p + i - 1]
for j \leftarrow 1 to n_2
    do R[i] \leftarrow A[q + i]
L[n_1+1] \leftarrow \infty
R[n_2+1] \leftarrow \infty
i ← 1
j ← 1
for k \leftarrow p to r
    do if L[i] \leq R[i]
        then A[k] ← L[i]
                 i ← i + 1
        else A[k] \leftarrow R[j]
                 i \leftarrow i + 1
```

Input: Array containing sorted subarrays A[p..q] and A[q+1..r].

Output: Merged sorted subarray in A[p..r].

O(n)

**Sentinels**, to avoid having to check if either subarray is fully copied at each step.

Source: Cormen Thomas H.; Leiserson, Charles E.; Rivest, Ronald L.; Stein, Clifford (2009) [1990]. Introduction to Algorithms (3rd ed.):



- Merge-sort is out-place sorting technique. During merge procedure it takes O(n) extra space. Therefore merge sort is not in-place sorting algorithm.
- Merge sort is not better approach for smaller size array.
- For smaller size array use insertion sort only.
- Merge sort is a stable sorting algorithm. After sorting if repeated element order not changes is called stable sorting technique.



#### Time Analysis of Merge Sort

```
MergeSort (A, p, r) // sort A[p...r] by divide & conquer

1 if p < r

2 then q \leftarrow \lfloor (p+r)/2 \rfloor \longrightarrow O(1)

3 MergeSort (A, p, q) \longrightarrow T(n/2)

4 MergeSort (A, q+1, r) \longrightarrow T(n/2)

5 Merge (A, p, q, r) // merges A[p...q] with A[q+1...r]
```

So, we can write the overall running time of MERGE-SORT function in the form of recurrence relation as:

$$T(n)=2T(n/2)+\Theta(n)+\Theta(1)$$

So Solving above equation:  $\Theta(nlgn)$ 



## Quicksort Algorithm

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

- If array contains only one element, return
- Else
  - pick one element to use as pivot.
  - Partition elements into two sub-arrays:
    - •Left sub-array contains elements less than or equal to pivot
    - •Right sub-array contains elements greater than pivot
  - Then repeatedly apply above two steps on the two sub-arrays until array become sorted
  - Return results



# 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.



### Partition – Choosing the pivot

- First, we have to select a pivot element among the elements of the given array.
- Which array item should be selected as pivot?
  - Somehow we have to select a pivot, and we hope that we will get a good partitioning.
  - If the items in the array arranged randomly, we choose a pivot randomly.
  - We can choose the first or last element as a pivot (it may not give a good partitioning).
  - We can use different techniques to select the pivot.

### Quick Sort Algorithm



As the name implies, it is quick, and it is the algorithm generally preferred for sorting.

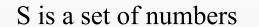
#### **Basic Ideas**

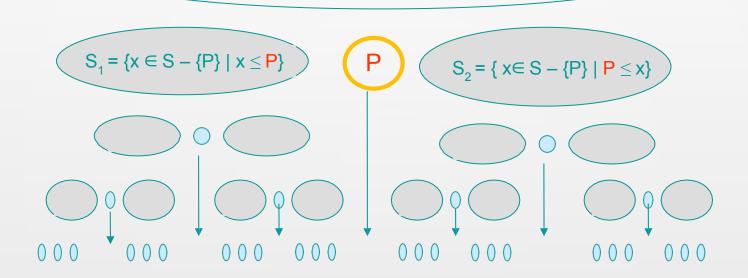
(Another divide-and-conquer algorithm)

- ☐ Pick an element, say P(the pivot)
- ☐ Re-arrange the elements into 3 sub-blocks,
  - 1.those less than or equal to  $(\leq)$  P(the left-block S1)
  - 2. P(the only element in the **middle**-block)
  - 3. those greater than or equal to  $(\geq)$  P(the **right-block S2**)
- □ Repeat the process recursively for the left-and right-sub-blocks. Return {quicksort(S1), P, quicksort(S2)}.(That is the results of quicksort(S1), followed by P, followed by the results of quicksort(S2))



#### Basic Ideas

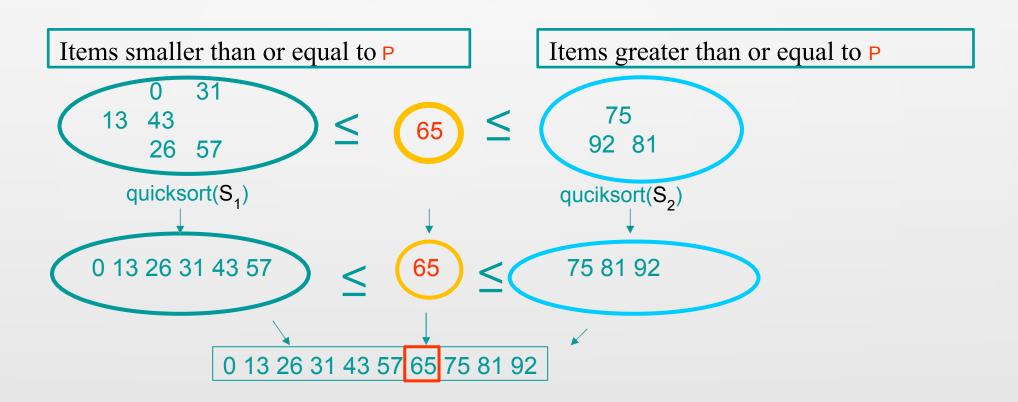






#### Basic Ideas

Pick a "Pivot" value, P Create 2 new sets without P





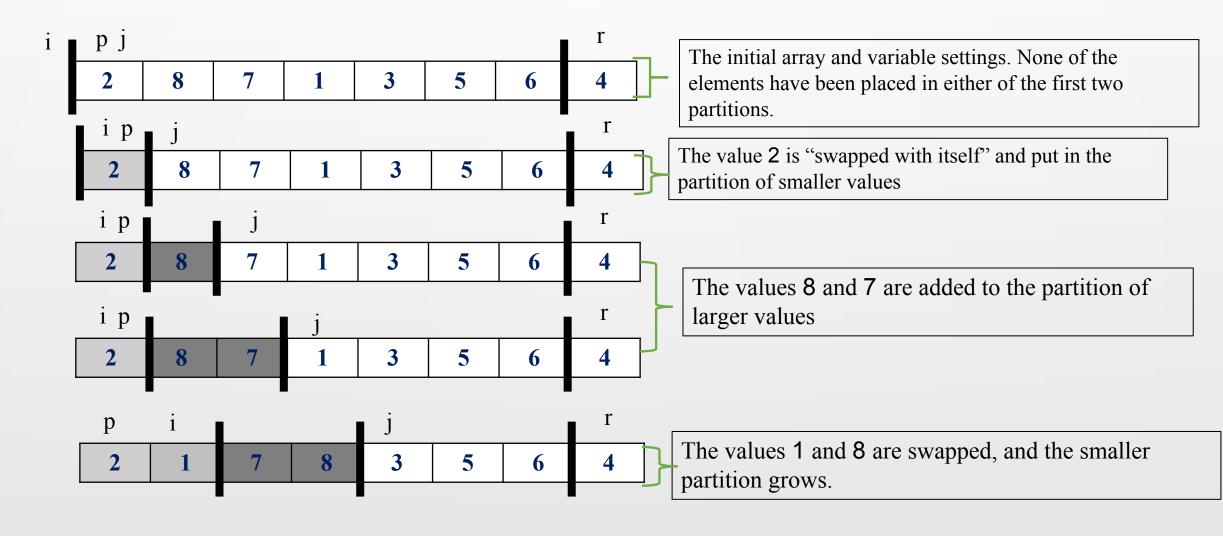
#### Quicksort Code:

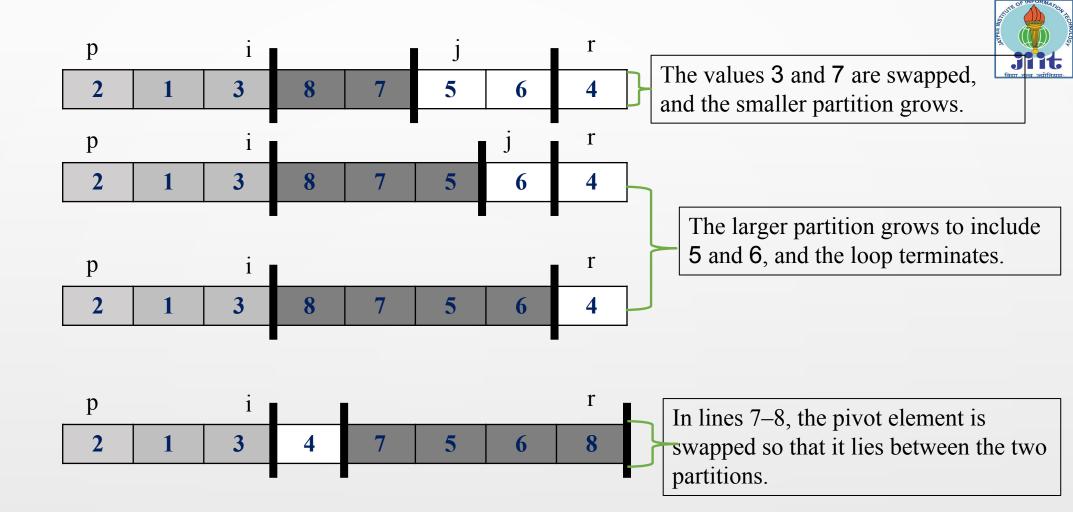


```
int pivot = arr[high]; // selecting last element as pivot
       int i = (low - 1); // index of smaller element
      for (int j = low; j \le high-1; j++)
 // If the current element is smaller than or equal to pivot
     if (arr[j] <= pivot)
       { i++; // increment index of smaller element
       swap(&arr[i], &arr[j]);
swap(\&arr[i+1], \&arr[high]);
  Return (i + 1);
```

Lightly shaded array elements are all in the first partition with values no greater than x. Heavily shaded elements are in the second partition with values greater than x. The unshaded elements have not yet been put in one of the first two partitions, and the final white element is the pivot x









- The advantage of this quicksort is that we can sort "in-place", i.e., without the need for a temporary buffer depending on the size of the inputs.
- Partitioning Step: Time Complexity is O(n).

Recall that quicksort involves partitioning, and 2 recursive calls.

Thus, giving the basic quicksort relation:

$$T(n) = O(n) + T(i) + T(n-i-1) = cn + T(i) + T(n-i-1)$$

where i is the size of the first sub-block after partitioning.

We shall take T(0) = T(1) = 1 as the initial conditions.

To find the solution for this relation, we'll consider three cases:

- 1. The Worst-case
- 2. The Best-case
- 3. The Average-case

All above cases depends on the value of the pivot!!



Worst-Case (Data is sorted already)

- When the pivot is the smallest (or largest) element at partitioning on a block of size
   n, the result
  - yields one empty sub-block, one element (pivot) in the "correct" place and one sub-block of size (*n*-1)
  - takes O(n) times.
- Recurrence Equation:

$$T(1) = 1$$

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

Solution:  $O(n^2)$ 

Worse than Mergesort!!!



#### **Best case:**

- The pivot is in the middle (median) (at each partition step), i.e. after each partitioning, on a block of size *n*, the result
  - yields two sub-blocks of approximately equal size and the pivot element in the "middle" position
  - $\bullet$  takes n data comparisons.
- Recurrence Equation becomes

$$T(1) = 1$$
$$T(n) = 2T(n/2) + cn$$

Solution: **O**(*n* logn)

Comparable to Mergesort!!



#### Average case:

The average case input would be any randomly generated arrays and the pivot is at the random position every step.

It turns out the average case running time also is  $O(n \log n)$ . More in detail will discuss offline.



### So the trick is to select a good pivot

Different ways to select a good pivot.

- First element
- Last element
- Median-of-three elements
  - Pick three elements, and find the median x of these elements. Use that median as the pivot.
- Random element
  - Randomly pick a element as a pivot.



### Selection problem: a 'faster' answer.

We can borrow the idea from the partition algorithm. Suppose we want to find a element of rank i in A[1..n]. After the 1<sup>st</sup> partition call (use **a random element** as pivot):

- 1. If the return index 'q' = i, then A[q] is the element we want. (Since there is exactly i-1 elements smaller than or equal to A[q]).
- If the return index 'q' > i, then the target element can NOT be in A[q .. right]. The target element is rank i in A[1.. q-1] Recursive call with parameters (A, 1, q-1, i).
- 3. If the return index 'q' < i, then the target element can NOT be in A[1 .. q]. The target element is rank i-q in A[q+1 ..n] Recursive call with parameters (A, q+1,n, i-q).

**→** 



### A 'faster' selection algorithm: Codes

```
RANDOMIZED-SELECT(A, p, r, i)
                                                   return a element of rank i in A[p..r]
    if p = r
       then return A[p]
   q \leftarrow \text{RANDOMIZED-PARTITION}(A, p, r)
                                                          A[p..q-1] \le A[\mathbf{q}] \le A[q+1..r]
   k \leftarrow q - p + 1
                                                          size of A[p..q]= k
   if i = k
                    be the pivot value is the answer
6
      then return A[q]
   elseif i < k
8
      then return RANDOMIZED-SELECT (A, p, q - 1, i)
9
   else return RANDOMIZED-SELECT(A, q + 1, r, i - k)
```

Source: Cormen Thomas H.; Leiserson, Charles E.; Rivest, Ronald L.; Stein, Clifford (2009) [1990]. Introduction to Algorithms (3rd ed.).



## Different sorting algorithms

Sorting Algorithm	Worst-case time	Average- case time	Space overhead ⊕(1)	
Bubble Sort	$\Theta$ (n <sup>2</sup> )	$\Theta$ (n <sup>2</sup> )		
Bassic cort	O (11 )	O (11 )	O(1)	
Insertion Sort	$\Theta$ (n <sup>2</sup> )	$\Theta$ (n <sup>2</sup> )	Θ(1)	
Merge Sort	Θ (n log n)	Θ (n log n)	Θ(n)	
Quick Sort	$\Theta$ (n <sup>2</sup> )	⊕ (n log n)	Θ(1)	

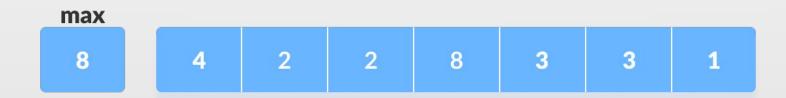


#### Counting Sort

- Counting sort is a sorting algorithm that sorts the elements of an array by counting the number of occurrences of each unique element in the array.
- The count is stored in an auxiliary array and the sorting is done by mapping the count as an index of the auxiliary array.

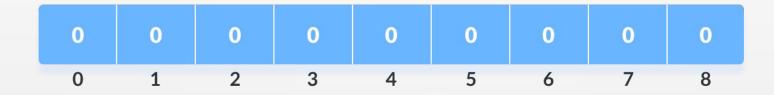
#### **How Counting Sort Works?**

Find out the maximum element (let it be max) from the given array:





2. Initialize an array of length max+1 with all elements 0. This array is used for storing the count of the elements in the array.

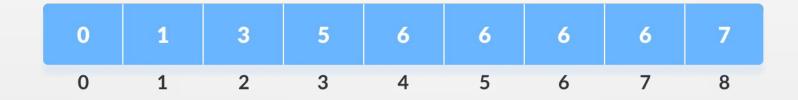


3. Store the count of each element at their respective index in count array. For example: if the count of element 3 is 2 then, 2 is stored in the 3rd position of count array. If element "5" is not present in the array, then 0 is stored in 5th position.

0	1	2	2	1	0	0	0	1
	1							

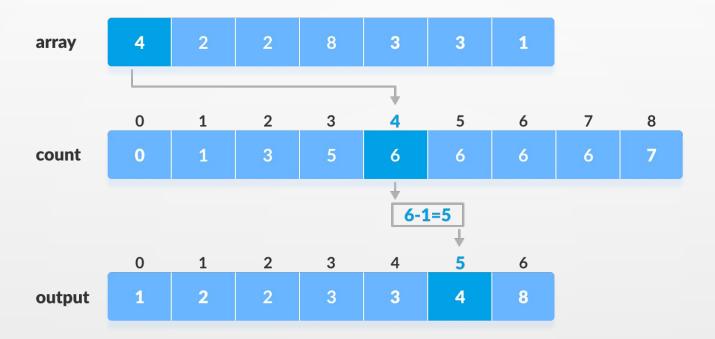


4. Store cumulative sum of the elements of the count array. It helps in placing the elements into the correct index of the sorted array.



5. Find the index of each element of the original array in the count array. This gives the cumulative count. Place the element at the index calculated as shown in figure below





6. After placing each element at its correct position, decrease its count by one.



## Counting Sort Algorithm

```
countingSort(array, size)

max <- find largest element in array
initialize count array with all zeros
for j <- 0 to size

find the total count of each unique element and
store the count at jth index in count array
for i <- 1 to max

find the cumulative sum and store it in count array itself
for j <- size down to 1
restore the elements to array decrease count of each element restored by 1
```



# Radix Sort

- This sort is unusual because it does not directly compare any of the elements
- We instead create a set of buckets and repeatedly separate the elements into the buckets
- On each pass, we look at a different part of the elements.
- Assuming decimal elements and 10 buckets, we would put the elements into the bucket associated with its units digit
- The buckets are actually queues so the elements are added at the end of the bucket
- At the end of the pass, the buckets are combined in increasing



### Radix Sort

- On the second pass, we separate the elements based on the "tens" digit, and on the third pass we separate them based on the "hundreds" digit
- Each pass must make sure to process the elements in order and to put the buckets back together in the correct order.
- Radix sort use counting sort in the intermediate steps.



### **Working of Radix Sort:**

1. Find the largest element in the array, i.e. max. Let X be the number of digits n max. X is calculated because we have to go through all the significant places of all elements.

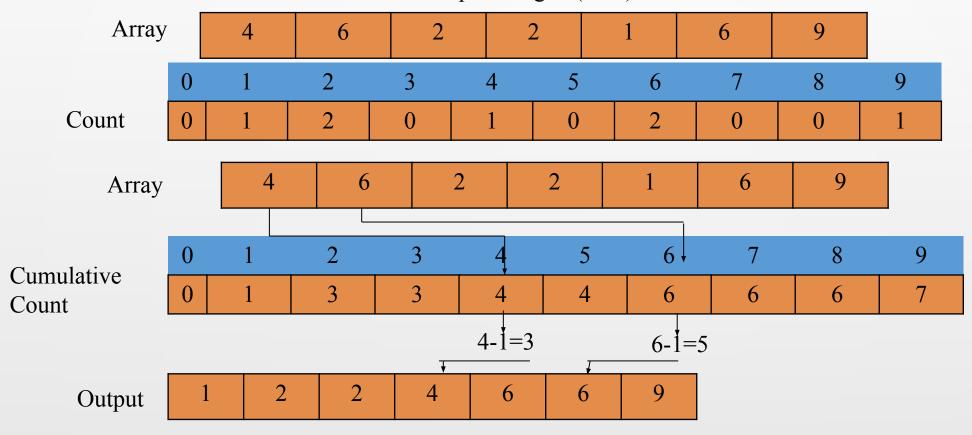
In the below array, we have the largest number 799. It has 3 digits. Therefore, the loop should go up to hundreds place (3 times).

2. Now, go through each significant place one by one. Use any stable sorting technique(Counting Sort) to sort the digits at each significant place. We have used counting sort for this





3. Sort the elements based on the unit place digits (X=0).



Original array sorted at unit place

|--|



3. Now, sort the elements based on digits at tens place.

1   <b>2</b> 2   1 <b>2</b> 4   <b>2</b> 6   4 <b>3</b> 6   5 <b>6</b> 2   7 <b>9</b> 9
---

Sort elements based on tens place

4. Finally, sort the elements based on the digits at hundreds place

001	022	<b>0</b> 26	<b>1</b> 24	<b>4</b> 36	<b>5</b> 62	<b>7</b> 99
-----	-----	-------------	-------------	-------------	-------------	-------------

Sort elements based on hundreds place



```
void radixsort(int array[], int size)
  { // Get maximum element
     int max = getMax(array, size);
 // Apply counting sort to sort elements based on place value.
  for (int place = 1; max / place > 0; place *= 10)
       countingSort(array, size, place);
void countingSort(int array[], int size, int place)
const int max = 10;
 int output[size]; int count[max];
for (int i = 0; i < max; ++i)
```



```
count[i] = 0; // Calculate count of elements
 for (int i = 0; i < size; i++)
 count[(array[i] / place) % 10]++;
for (int i = 1; i < max; i++)
count[i] += count[i - 1];
for (int i = size - 1; i >= 0; i--)
 { output[count[(array[i] / place) % 10] - 1] = array[i];
count[(array[i] / place) % 10]--;
for (int i = 0; i < size; i++)
array[i] = output[i]; }
```



### Radix Sort Analysis

- Each element is examined once for each of the digits it contains, so if the elements have at most M digits and there are N elements this algorithm has order O(M\*N)
- This means that sorting is linear based on the number of elements
- Why then isn't this the only sorting algorithm used?



## Radix Sort Analysis

- Though this is a very time efficient algorithm it is not space efficient
- If an array is used for the buckets and we have B buckets, we would need N\*B extra memory locations because it's possible for all of the elements to wind up in one bucket
- If linked lists are used for the buckets you have the overhead of pointers



#### **Bucket Sort**

- Bucket sort is mainly useful when input is uniformly distributed over a range. For example, consider the following problem.
  - Sort a large set of floating point numbers which are in range from 0.0 to 1.0 and are uniformly distributed across the range. How do we sort the numbers efficiently?
- A simple way is to apply a comparison based sorting algorithm. The <u>lower bound for comparison based sorting</u> algorithm (Merge Sort, Heap Sort, Quick-Sort .. etc) is  $\Omega(n \log n)$ , i.e., they cannot do better than nlogn.
- Can we sort the array in linear time? <u>Counting sort</u> can not be applied here as we use keys as index in counting sort. Here keys are floating point numbers.

The idea is to use bucket sort. Following is bucket algorithm.

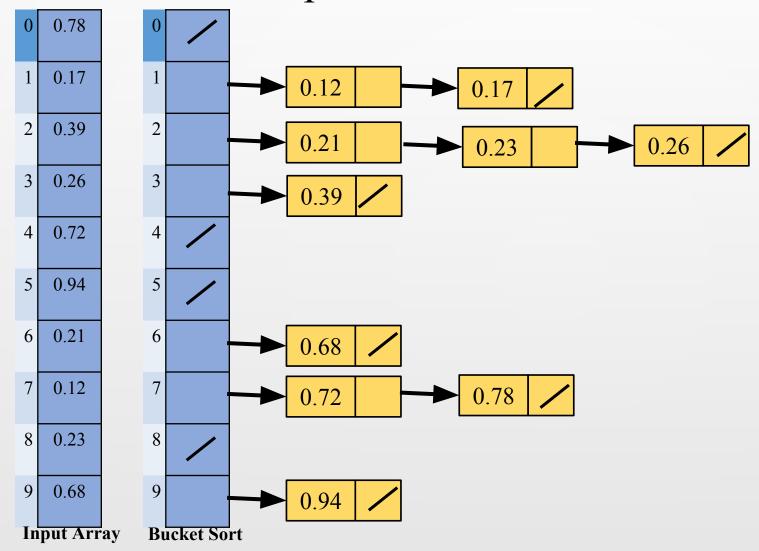


Bucket\_Sort(arr[], n)

- 1)Create n empty buckets (Or lists).
- 2) Do following for every array element arr[i].
  - a) Insert arr[i] into bucket[n\*array[i]]
- 3) Sort individual buckets using insertion sort.
- 4) Concatenate all sorted buckets.



### Bucket Sort Example





# Complexity of Bucket Sort

**Worst Case Complexity:** O(n<sup>2</sup>)

When there are elements of close range in the array, they are likely to be placed in the same bucket. This may result in some buckets having more number of elements than others.

#### **Best Case Complexity:** O(n+k)

It occurs when the elements are uniformly distributed in the buckets with a nearly equal number of elements in each bucket.

#### **Average Case Complexity:** O(n)

It occurs when the elements are distributed randomly in the array. Even if the elements are not distributed uniformly, bucket sort runs in linear time



### References:

- <u>Cormen, Thomas H.</u>; <u>Leiserson, Charles E.</u>; <u>Rivest, Ronald L.</u>; <u>Stein, Clifford</u> (2009) [1990]. Introduction to Algorithms (3rd ed.). MIT Press and McGraw-Hill. <u>ISBN 0-262-03384-4</u>. 1320 pp.
- <a href="https://home.cse.ust.hk/~dekai/271/notes/L01a/quickSort.pdf">https://home.cse.ust.hk/~dekai/271/notes/L01a/quickSort.pdf</a>
- <a href="https://www.programiz.com/dsa/counting-sort">https://www.programiz.com/dsa/counting-sort</a>