

## Counting and Radix Sort

### Sorting So Far

- **Insertion sort**
  - Easy to code
  - Fast on small inputs (less than ~50 elements)
  - Fast on nearly-sorted inputs
  - $O(n^2)$  worst case
  - $O(n^2)$  average

### Sorting So Far

- **Merge sort**
  - Divide-and-conquer:
    - Split array in half
    - Recursively sort subarrays
    - Linear-time merge step
  - $O(n \lg n)$  worst case

### Sorting So Far

- **Quick sort**
  - Divide-and-conquer
    - Partition array into two subarrays, recursively sort
    - All of first subarray < all of second subarray
    - **No merge step needed**
  - $O(n \lg n)$  average case
  - $O(n^2)$  worst case
  - Fast in practice

### Counting sort

- Counting sort assumes that **each of the  $n$  input elements** is an integer in the **range 0 to  $k$** .
- **$n$  is the number of elements** and  **$k$  is the highest value element**.
- Consider the input set : 4, 1, 3, 4, 3  $\rightarrow n=5$  and  $k=4$
- Counting sort determines for each input element  $x$ , **the number of elements less than  $x$** . And it uses this information to place element  **$x$  directly into its position in the output array**.
- For example if there exists 20 elements less than  $x$  then  $x$  is placed into the 21<sup>st</sup> position into the output array.
- The algorithm uses three array:
  - Input Array:**  $A[1..n]$  store input data where  $A[j] \in \{1, 2, 3, \dots, k\}$
  - Output Array:**  $B[1..n]$  finally store the sorted data
  - Temporary Array:**  $C[1..k]$

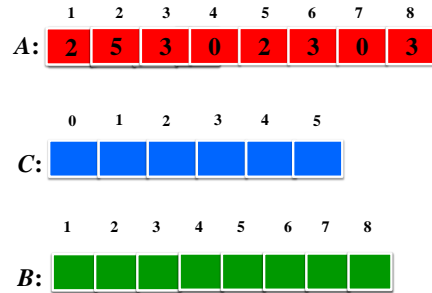
### Counting Sort

1. Counting-Sort( $A, B, k$ )
2. Let  $C[0..k]$  be a new array
3. for  $i=0$  to  $k$
4.      $C[i] = 0$ ;
5. for  $j=1$  to  $A.length$  (or  $n$ )
6.      $C[A[j]] = C[A[j]] + 1$ ;
7. for  $i=1$  to  $k$
8.      $C[i] = C[i] + C[i-1]$ ;
9. for  $j=n$  or  $A.length$  down to 1
10.     $B[C[A[j]]] = A[j]$ ;
11.     $C[A[j]] = C[A[j]] - 1$ ;

## Counting Sort

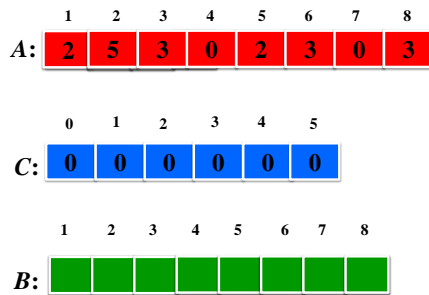
1. Counting-Sort(A, B, k)
2. Let  $C[0 \dots k]$  be a new array
3. for  $i=0$  to  $k$  [Loop 1]
4.    $C[i] = 0$ ;
5. for  $j=1$  to A.length ( or n) [Loop 2]
6.    $C[A[j]] = C[A[j]] + 1$ ;
7. for  $i=1$  to  $k$  [Loop 3]
8.    $C[i] = C[i] + C[i-1]$ ;
9. for  $j=n$  or A.length down to 1 [Loop 4]
10.    $B[C[A[j]]] = A[j]$ ;
11.    $C[A[j]] = C[A[j]] - 1$ ;

## Counting-sort example



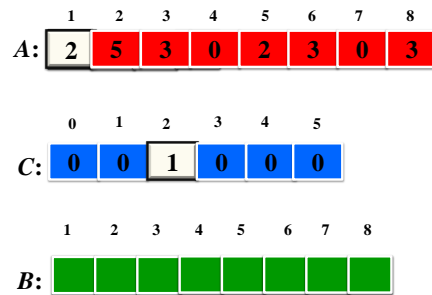
### Executing Loop 1

3. for  $i=0$  to  $k$
4.    $C[i] = 0$ ;



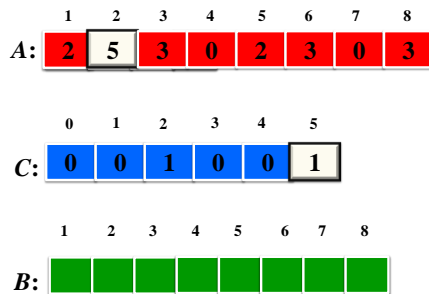
### Executing Loop 2

5. for  $j=1$  to A.length or n
6.    $C[A[j]] = C[A[j]] + 1$ ;



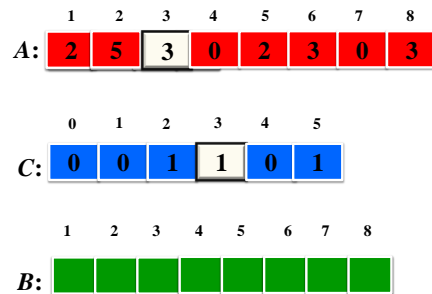
### Executing Loop 2

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**Executing Loop 2**

5. for j=1 to A.length or n  
 6.  $C[A[j]] = C[A[j]] + 1;$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
1	0	1	1	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 2**

5. for j=1 to A.length or n  
 6.  $C[A[j]] = C[A[j]] + 1;$

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1	2	3	4	5	6	7	8
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C: 

0	1	2	3	4	5
1	0	2	1	0	1

B: 

1	2	3	4	5	6	7	8

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1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
1	0	2	2	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 2**

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A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	0	2	2	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 2**

5. for j=1 to A.length or n  
 6.  $C[A[j]] = C[A[j]] + 1;$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	0	2	3	0	1

B: 

1	2	3	4	5	6	7	8

**End of Loop 2**

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	0	2	3	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 3**

7. for i=1 to k  
8.  $C[i] = C[i] + C[i-1];$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	0	2	3	0	1

C: 

0	1	2	3	4	5
2	2	2	3	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 3**

7. for i=1 to k  
8.  $C[i] = C[i] + C[i-1];$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	2	2	3	0	1

C: 

0	1	2	3	4	5
2	2	4	3	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 3**

7. for i=1 to k  
8.  $C[i] = C[i] + C[i-1];$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	2	4	3	0	1

C: 

0	1	2	3	4	5
2	2	4	7	0	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 3**

7. for i=1 to k  
8.  $C[i] = C[i] + C[i-1];$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	2	4	7	0	1

C: 

0	1	2	3	4	5
2	2	4	7	7	1

B: 

1	2	3	4	5	6	7	8

**Executing Loop 3**

7. for i=1 to k  
8.  $C[i] = C[i] + C[i-1];$

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

C: 

0	1	2	3	4	5
2	2	4	7	7	1

C: 

0	1	2	3	4	5
2	2	4	7	7	8

B: 

1	2	3	4	5	6	7	8

**End of Loop 3**

A: 

1	2	3	4	5	6	7	8
2	5	3	0	2	3	0	3

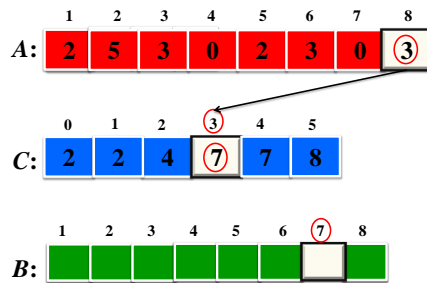
C: 

0	1	2	3	4	5
2	2	4	7	7	8

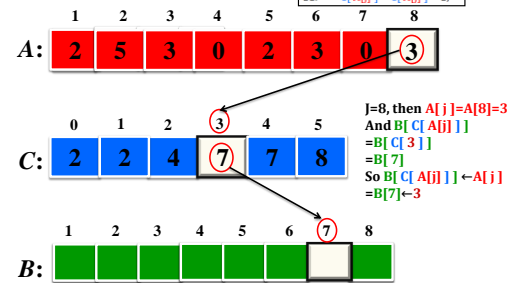
B: 

1	2	3	4	5	6	7	8

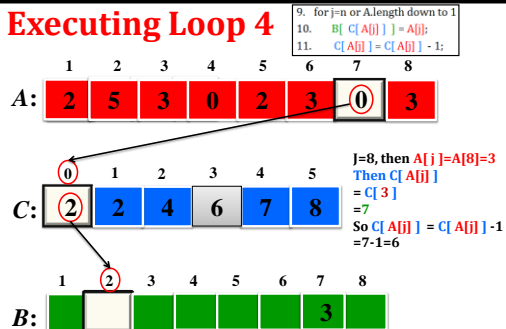
### Executing Loop 4



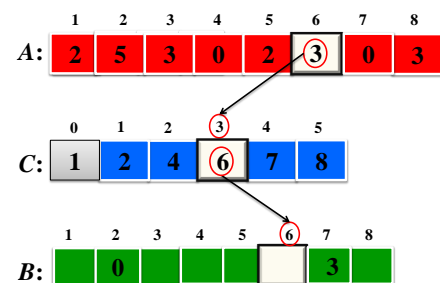
### Executing Loop 4



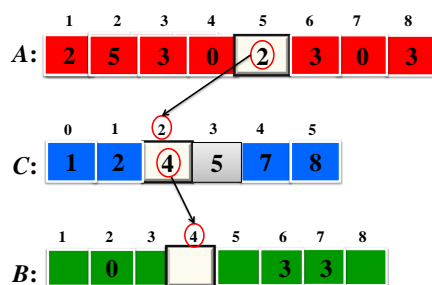
### Executing Loop 4



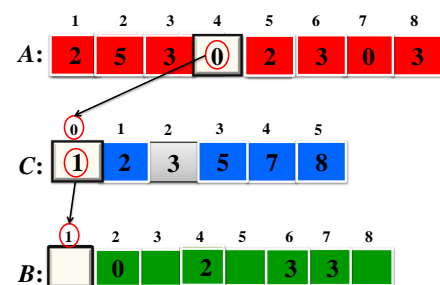
### Executing Loop 4



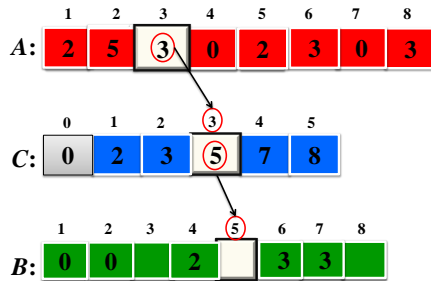
### Executing Loop 4



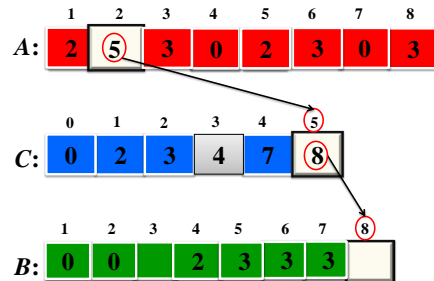
### Executing Loop 4



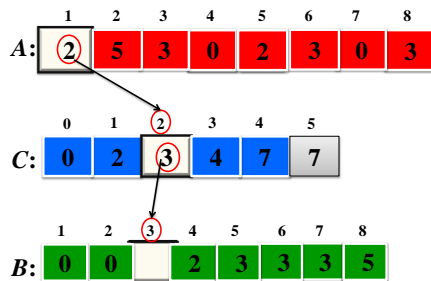
### Executing Loop 4



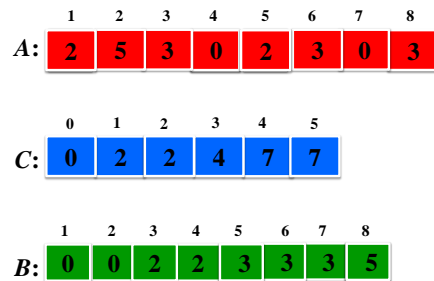
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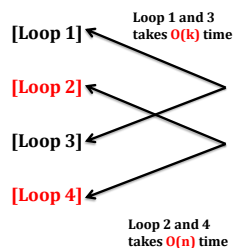


### End of Loop 4



### Time Complexity Analysis

1. Counting-Sort(A, B, k)
2. Let  $C[0 \dots k]$  be a new array
3. for  $i=0$  to  $k$
4.  $C[i] = 0$ ;
5. for  $j=1$  to A.length or  $n$
6.  $C[A[j]] = C[A[j]] + 1$ ;
7. for  $i=1$  to  $k$
8.  $C[i] = C[i] + C[i-1]$ ;
9. for  $j=n$  or A.length down to 1
10.  $B[C[A[j]]] = A[j]$ ;
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```

void CountingSort (int A[], int n, int B[], int K) {
    int C[K], i, j;
    //Complexity: O(K)
    for (i = 0; i < K; i++)
        C[i] = 0;

    //Complexity: O(n)
    for (j = 0; j < n; j++)
        C[A[j]] = C[A[j]] + 1;

    //C[i] now contains the number of elements equal to i
    //Complexity: O(K)
    for (i = 1; i < K; i++)
        C[i] = C[i] + C[i-1];

    //C[i] now contains the number of elements <= i
    //Complexity: O(n)
    for (j = n-1; j >= 0; j--) {
        B[C[A[j]]] = A[j];
        C[A[j]] = C[A[j]] - 1;
    }
}

```

## Time Complexity Analysis

- **So the counting sort takes a total time of:**
- $=O(k) + O(n) + O(k) + O(n) = O(n + k)$
- Usually,  $k = O(n)$ 
  - Thus counting sort runs in  $O(n)$  time
- Counting sort is called stable sort.

## Radix Sort

- Radix sort is non comparative sorting method
- Two classifications of radix sorts are **least significant digit (LSD)** radix sorts and most significant digit (**MSD**) radix sorts.
- **LSD radix sorts** process the integer representations starting from the least digit and move towards the most significant digit.

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## Radix Sort

In input array  $A$ , each element is a number of  $d$  digit.

Radix -Sort( $A, d$ )

for  $i \leftarrow 1$  to  $d$

do "use a stable sort to sort array  $A$  on digit  $i$ ;

329	720	720	329
457	355	329	355
657	436	436	436
839	457	839	457
436	657	355	657
720	329	457	720
355	839	657	839

## Radix Sort

- **What sort will we use to sort on digits?**
- Counting sort is obvious choice:
  - Sort  $n$  numbers on digits that range from  $1..k$
  - Time:  $O(n + k)$
- Each pass over  $n$  numbers with  **$d$  digits takes** time  $O(n+k)$ , so total time  **$O(dn+dk)$** 
  - When  $d$  is constant and  $k=O(n)$ , takes  $O(n)$  time
- **How many bits in a computer word?**

## Median Finding Algorithm

## Problem Definition

- **Given** a set of " $n$ " unordered numbers we want to find the " $k$ " smallest number. ( $k$  is an integer between 1 and  $n$ ).

## A Simple Solution

- A simple sorting algorithm (like heapsort) will take Order of  $O(n \lg_2 n)$  time.

Step	Running Time
Sort n elements (using heapsort)	$O(n \lg_2 n)$
Return the $k^{\text{th}}$ smallest element	$O(1)$
Total running time	$O(n \lg_2 n)$

## Linear Time selection algorithm

- Also called **Median Finding Algorithm**.
- Find  $k^{\text{th}}$  smallest element in  $O(n)$  time in worst case.
- Uses Divide and Conquer strategy.
- Uses elimination in order to cut down the running time substantially.

## Steps to solve the problem

- Step 1: If n is small, for example  $n < 6$ , just sort and return the  $k^{\text{th}}$  smallest number in constant time i.e;  $O(1)$  time.
- Step 2: Group the given number in subsets of 5 in  $O(n)$  time.
- Step 3: Sort each of the group in  $O(n)$  time. Find median of each group.

## Algorithm

- Given array A of size n and integer  $k \leq n$ ,
  - Group the array into  $n/5$  groups of size 5 and find the median of each group. (For simplicity, we will ignore integrality issues.)
  - Recursively, find the true median of the medians. Call this p
  - Use p as a pivot to split the array into subarrays LESS and GREATER.
  - Recursive on the appropriate piece.
- $O(n)$  comparisons to find the  $k^{\text{th}}$  smallest in an array of size n

Theorem 4.2 *DeterministicSelect makes  $O(n)$  comparisons to find the  $k^{\text{th}}$  smallest in an array of size n.*

Proof: Let  $T(n, k)$  denote the worst-case time to find the  $k^{\text{th}}$  smallest out of n, and  $T(n) = \max_k T(n, k)$  as before.

Step 1 takes time  $O(n)$ , since it takes just constant time to find the median of 5 elements. Step 2 takes time at most  $T(n/5)$ . Step 3 again takes time  $O(n)$ . Now, we claim that at least  $3/10$  of the array is  $\leq p$ , and at least  $3/10$  of the array is  $\geq p$ . Assuming for the moment that this claim is true, Step 4 takes time at most  $T(7n/10)$ , and we have the recurrence:

$$T(n) \leq cn + T(n/5) + T(7n/10), \quad (4.1)$$

## Description of the Algorithm step

- If n is small, for example  $n < 6$ , just sort and return the  $k^{\text{th}}$  smallest number. (Bound time- 7)
- If  $n > 5$ , then partition the numbers into groups of 5. (Bound time  $n/5$ )
- Sort the numbers within each group. Select the middle elements (the medians). (Bound time-  $7n/5$ )
- Call your "Selection" routine recursively to find the median of  $n/5$  medians and call it m. (Bound time-  $T_{n/5}$ )
- Compare all  $n-1$  elements with the median of medians m and determine the sets L and R, where L contains all elements  $< m$ , and R contains all elements  $> m$ . (Bound time- n)



## Recursive formula

- $T(n) = O(n) + T(n/5) + T(7n/10)$
- $T(n) \leq c7n/10 + cn/5 + O(n)$
- $T(n) = \Theta(n)$

## Example

- Given a set  
(.....2,5,9,19,24,54,5,87,9,10,44,32,21,13,24,18,26,16,19,25,39,47,56,71,91,61,44,28.....)  
having n elements.

## Arrange the numbers in groups of five

.....	2	54	44	4	25	.....
.....	5	5	32	18	39	.....
.....	9	87	21	26	47	.....
.....	19	9	13	16	56	.....
.....	24	10	2	19	71	.....

## Find median of N/5 groups

.....	2	5	2	4	25	.....
.....	5	9	13	16	39	.....
.....	9	10	21	18	47	.....
.....	19	54	32	19	56	.....
.....	24	87	44	26	71	.....

Median of each group

## Find the Median of each group

.....	2	5	2	4	25	.....
.....	5	9	13	16	39	.....
.....	9	10	21	18	47	.....
.....	19	54	32	19	56	.....
.....	24	87	44	26	71	.....

Find **m**, the median of medians