



Data Structure and Algorithms

Session-22

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Counting sort

- Counting sort assumes that each of the n input elements is an integer in the range 0 to k . that is n is the number of elements and k is the highest value element.
- Consider the input set : 4, 1, 3, 4, 3. Then $n=5$ and $k=4$
- 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:** $C[1..k]$ store data temporarily

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
2	1	1	0	2	5	4	0	2	8	7	7	9	2	0	1	9

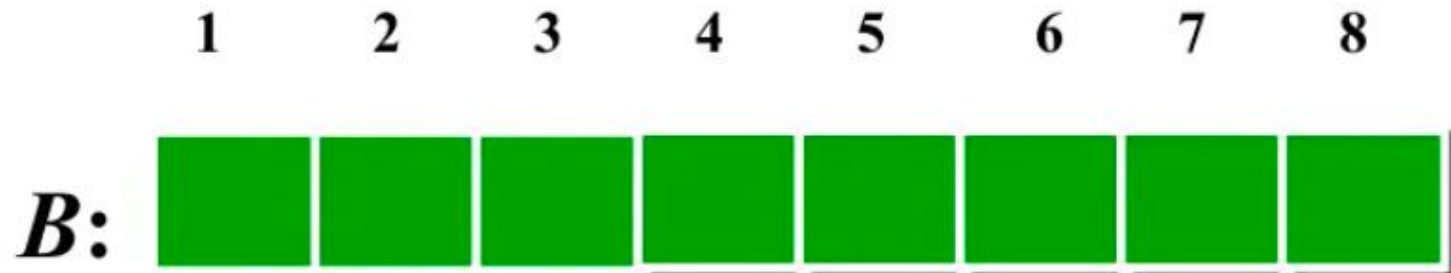
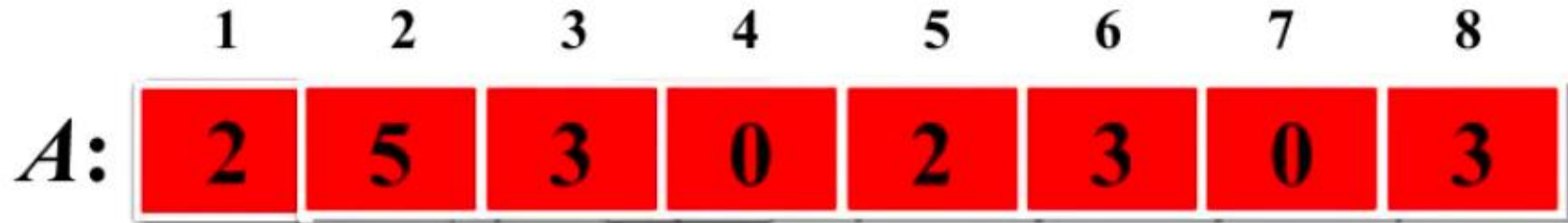
Counting Sort

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. $B[C[A[j]]] = A[j];$
11. $C[A[j]] = C[A[j]] - 1;$

Counting Sort

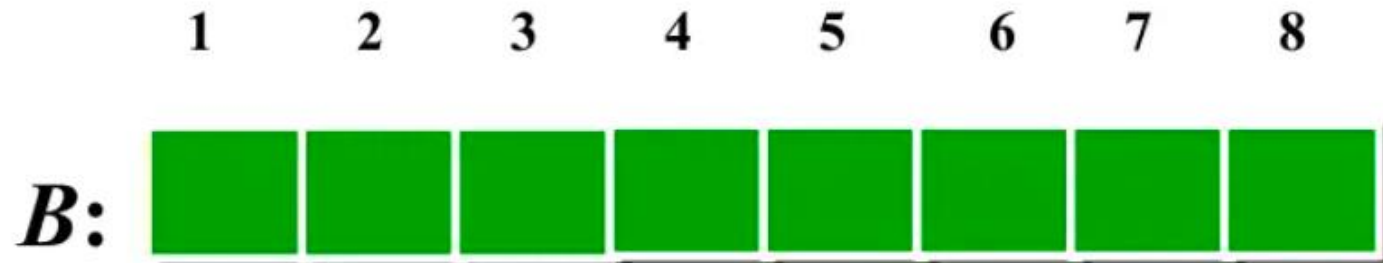
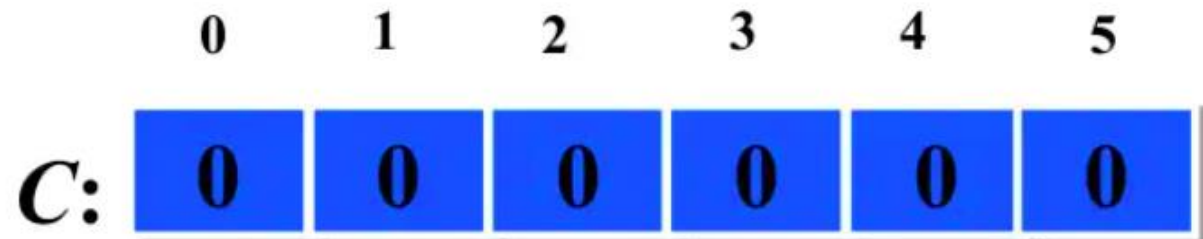
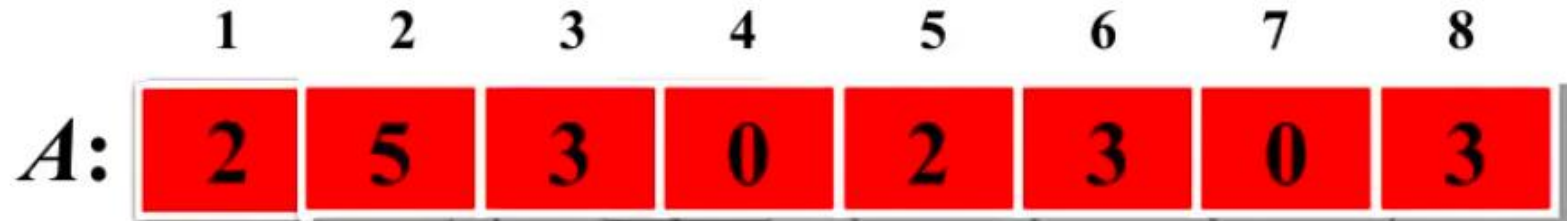
1. Counting-Sort(A, B, k)
2. Let $C[0 \dots k]$ be a new array
3. for $i=0$ to k **[Loop 1]**
4. $C[i] = 0;$
5. for $j=1$ to $A.length$ (or n) **[Loop 2]**
6. $C[A[j]] = C[A[j]] + 1;$
7. for $i=1$ to k **[Loop 3]**
8. $C[i] = C[i] + C[i-1];$
9. for $j=n$ or $A.length$ down to 1 **[Loop 4]**
10. $B[C[A[j]]]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

Counting-sort example



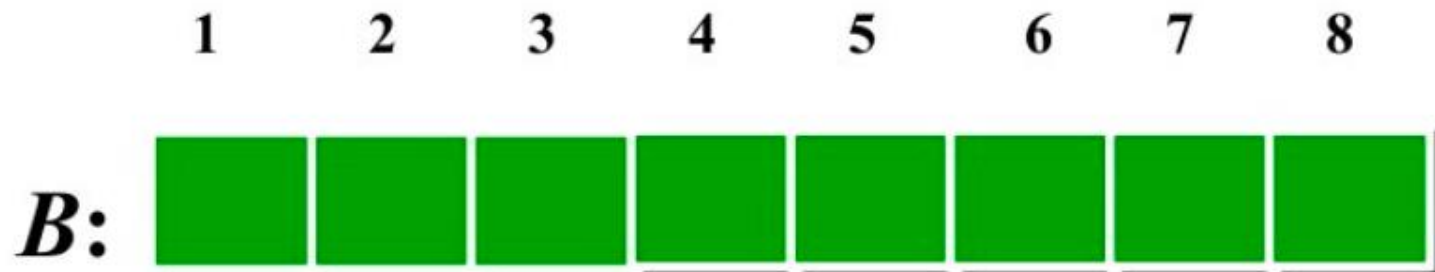
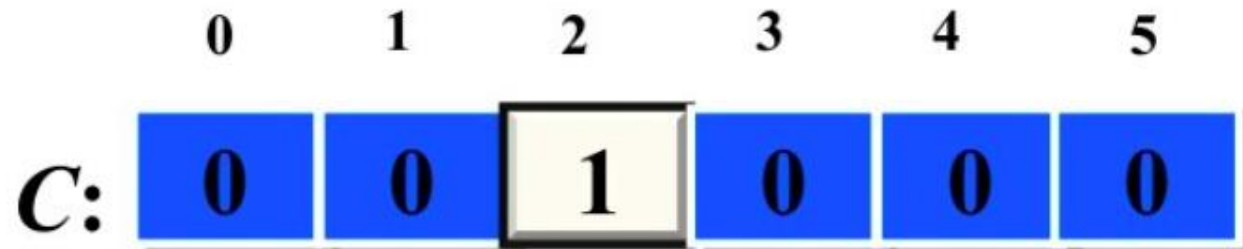
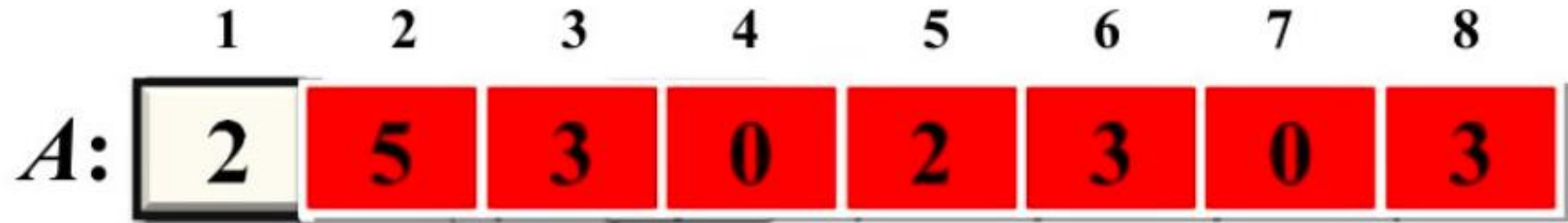
Executing Loop 1

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



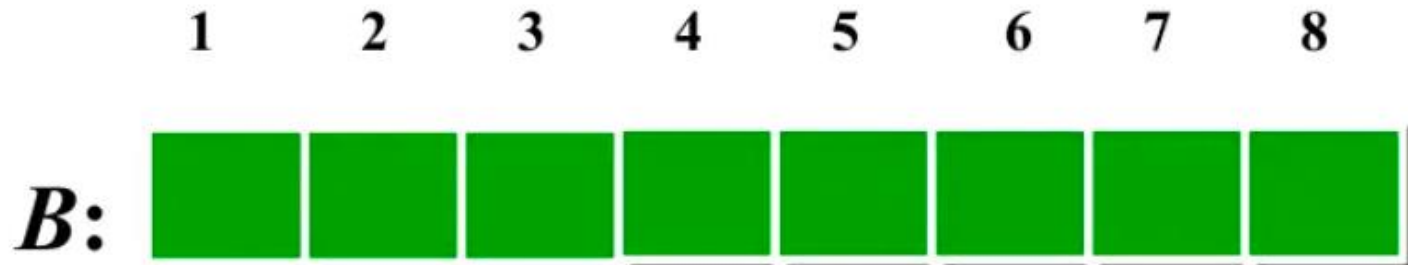
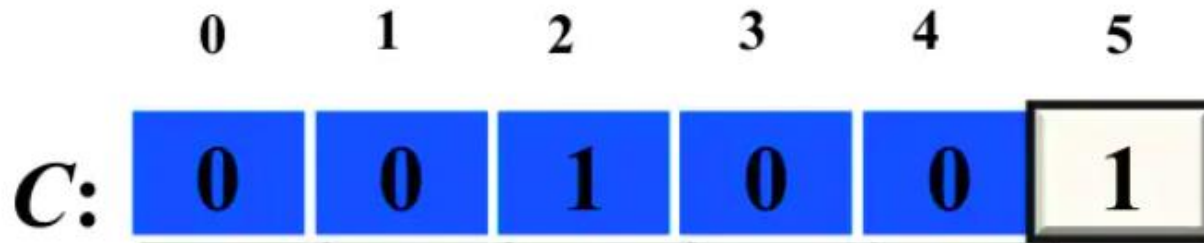
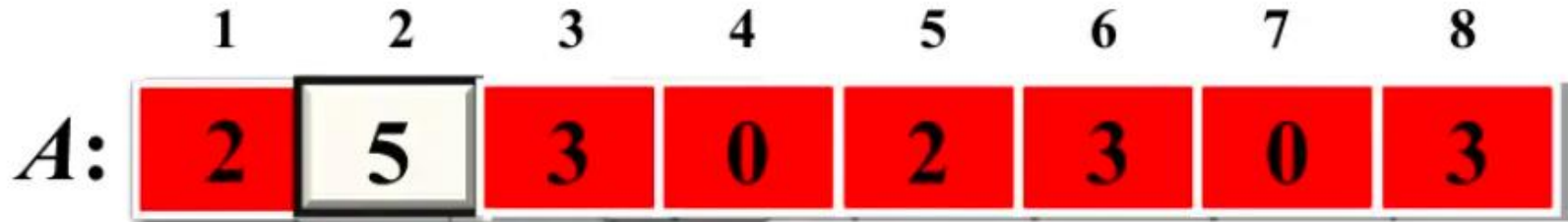
Executing Loop 2

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



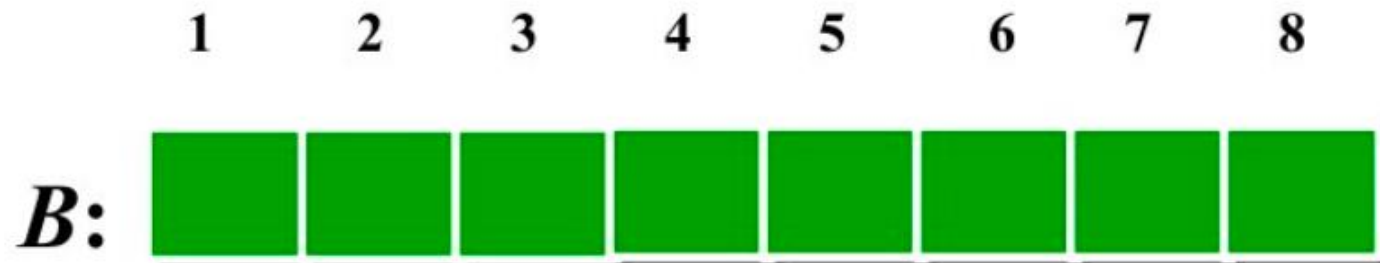
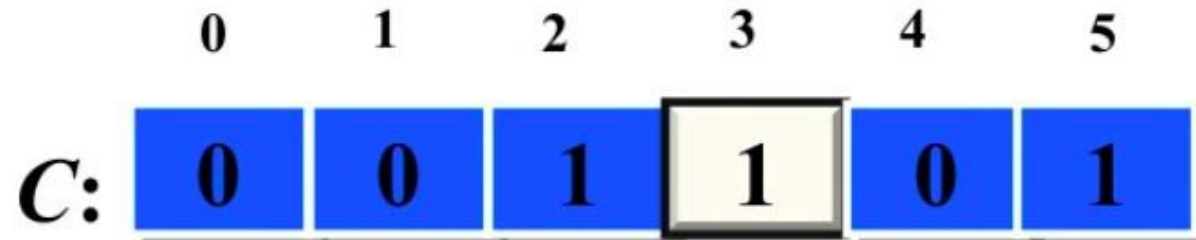
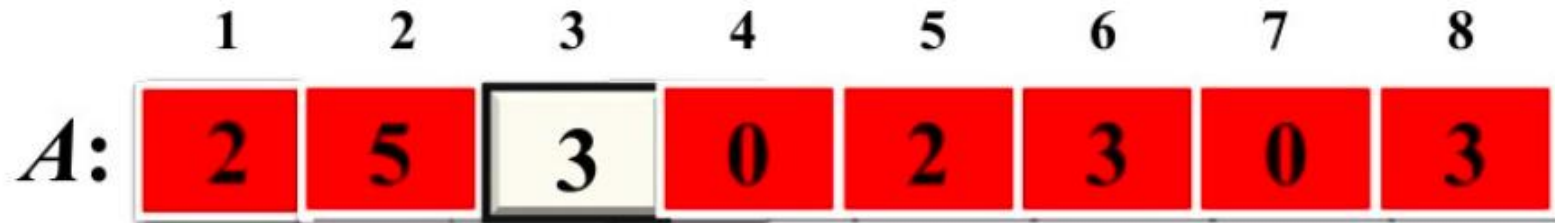
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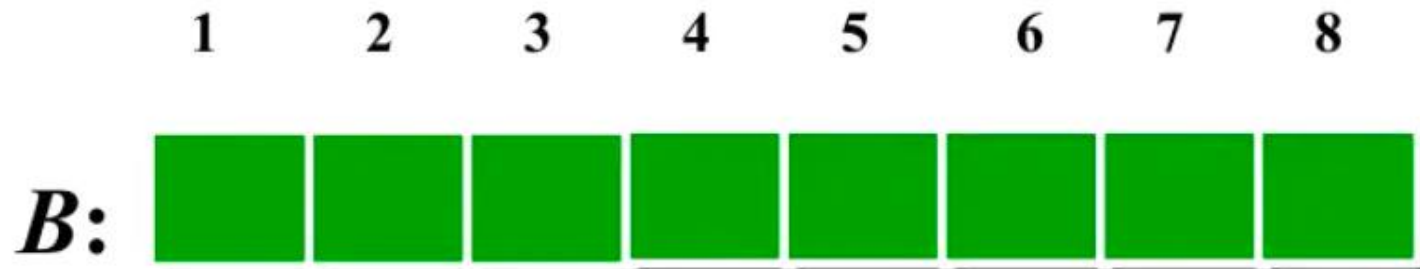
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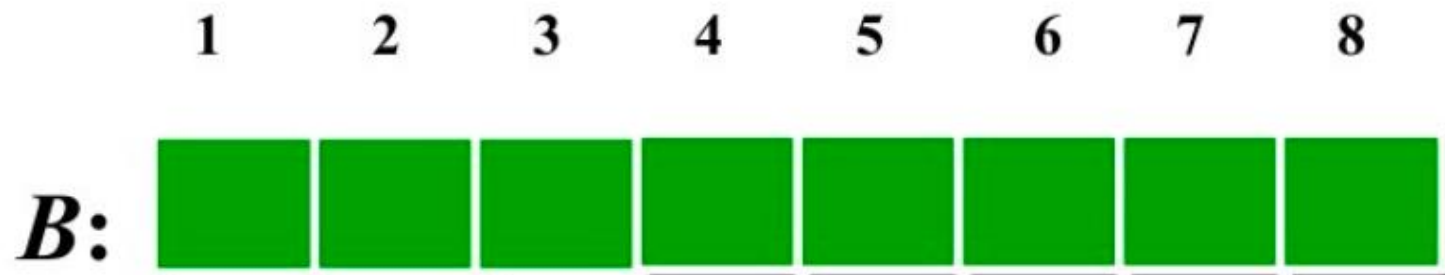
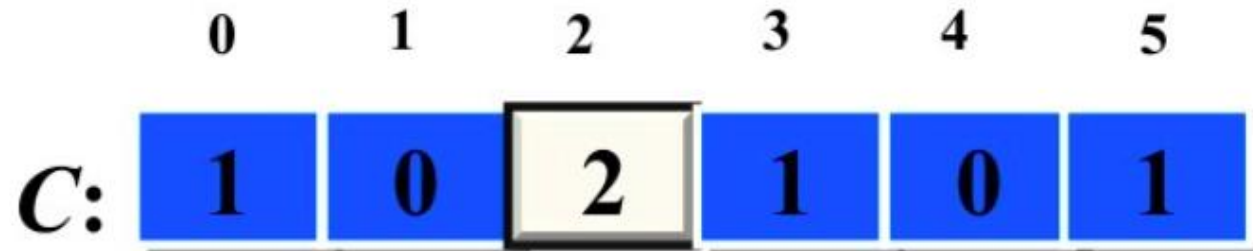
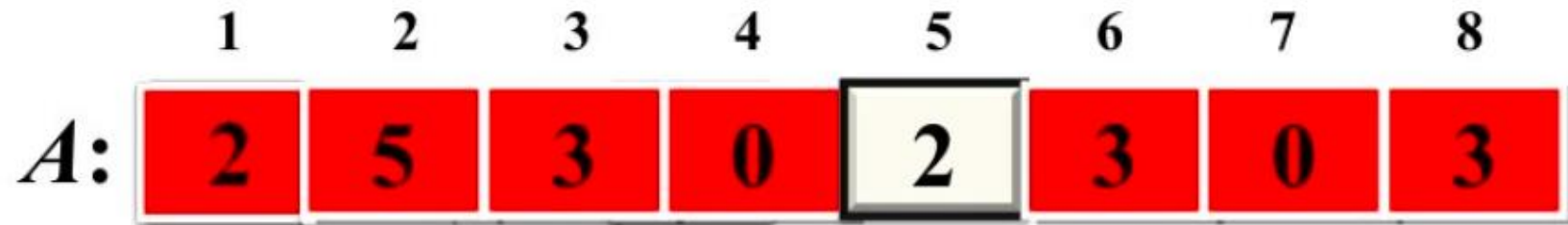
Executing Loop 2

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Executing Loop 2

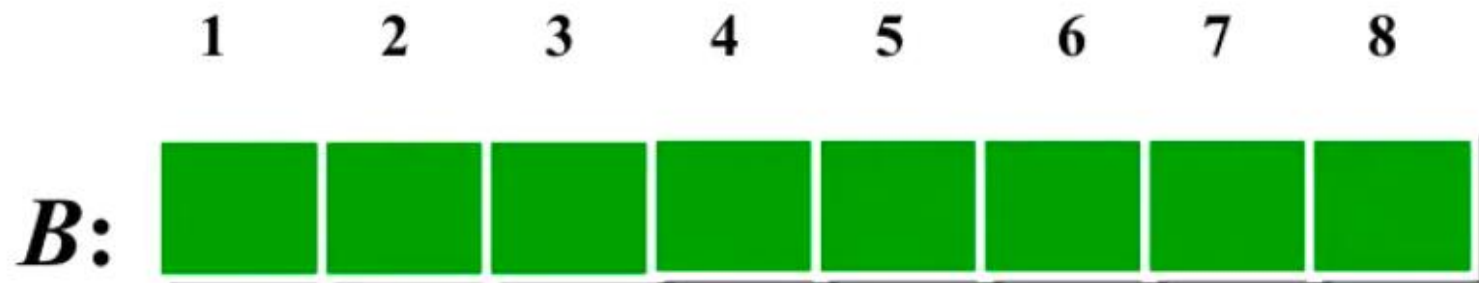
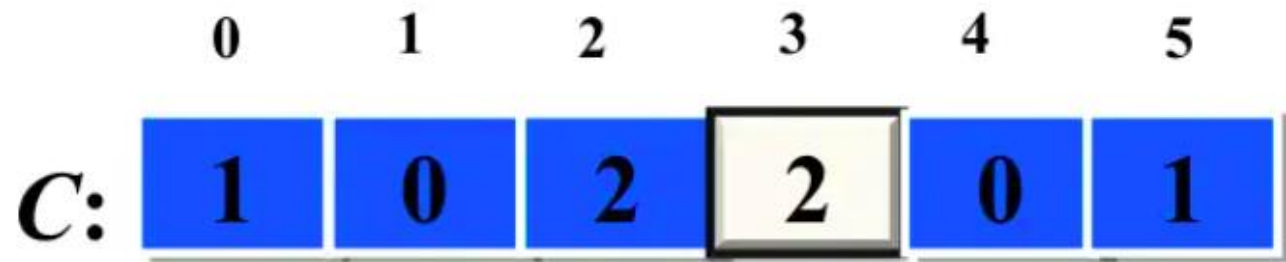
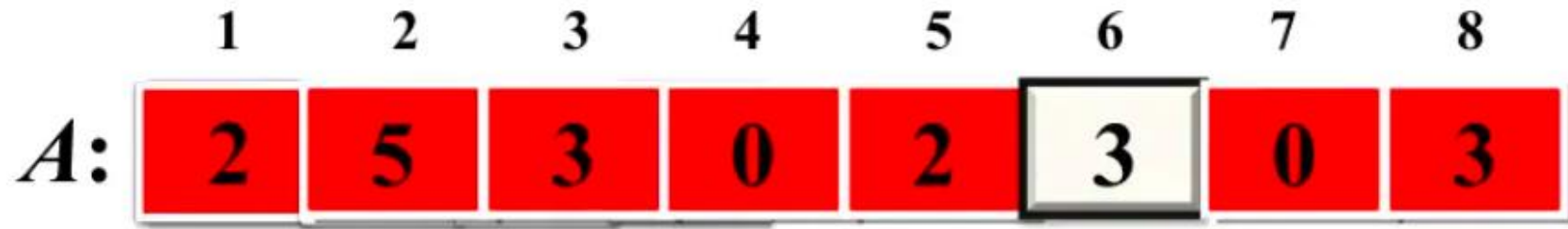
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Executing Loop 2

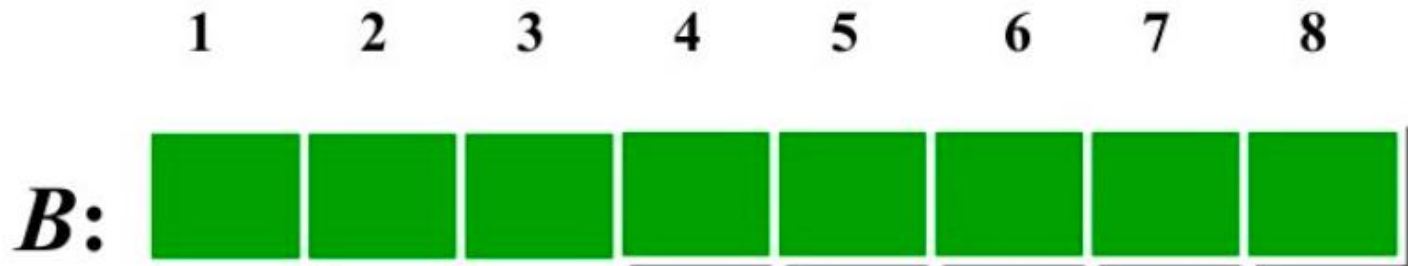
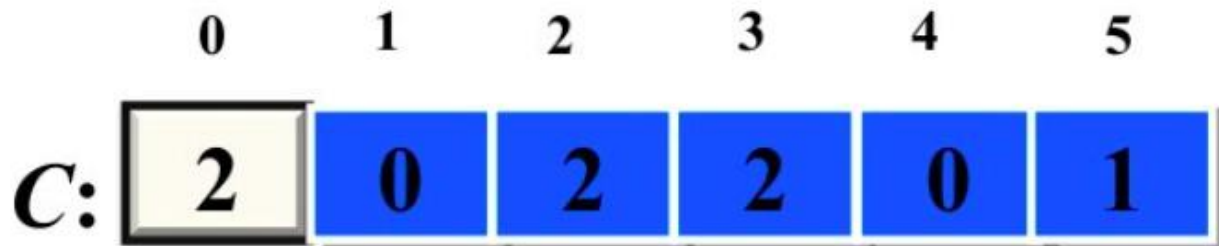
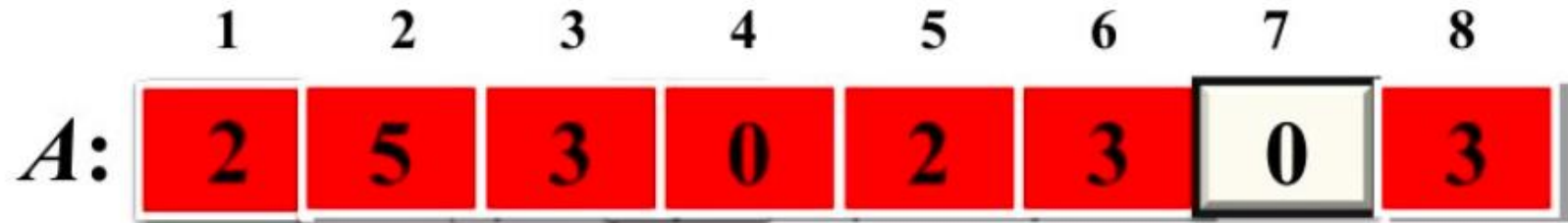
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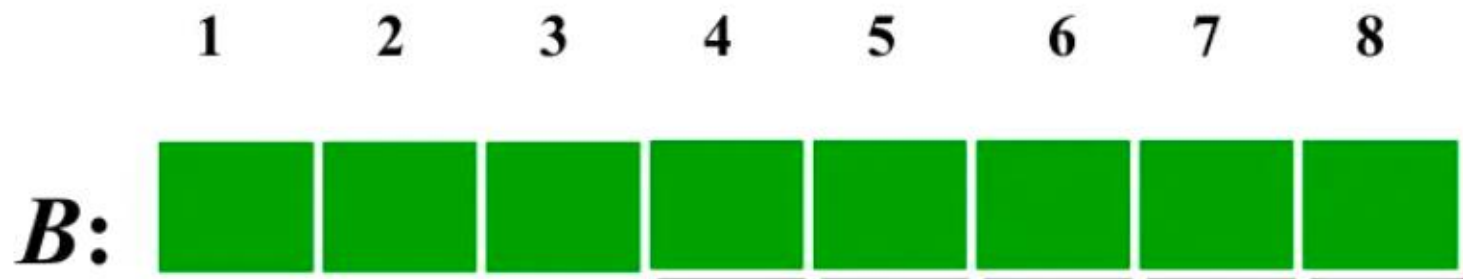
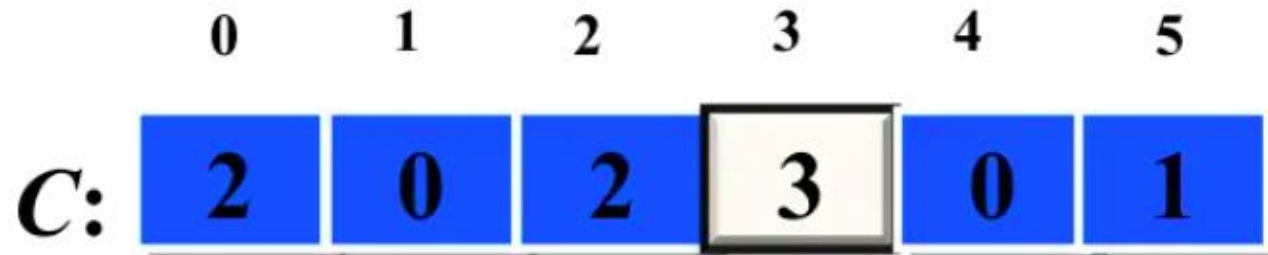
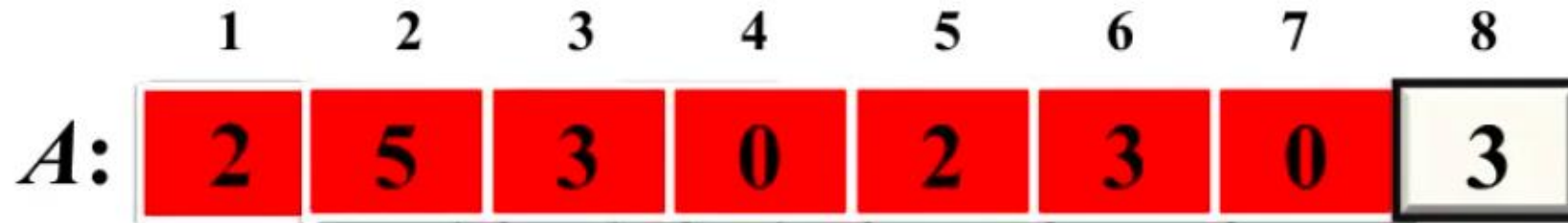
Executing Loop 2

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



Executing Loop 2

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



End of Loop 2

A:

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

C:

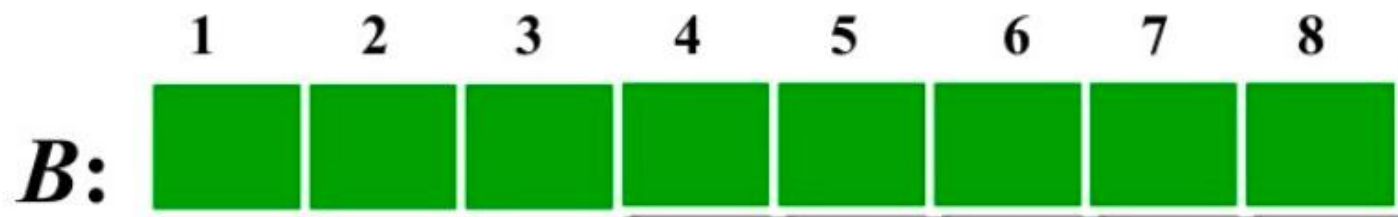
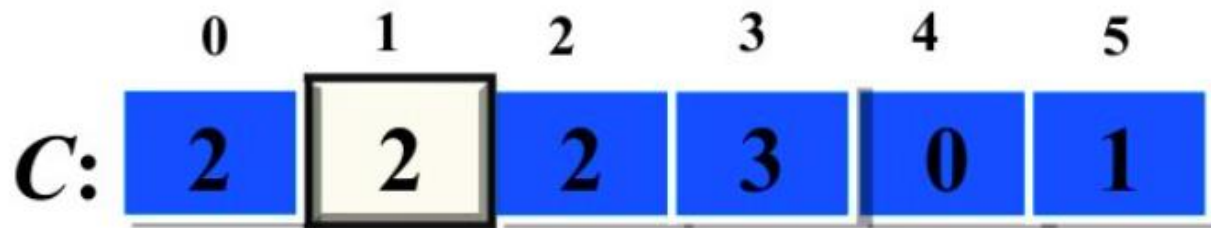
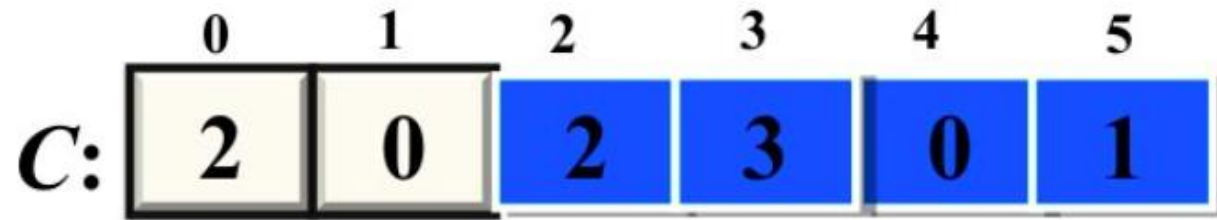
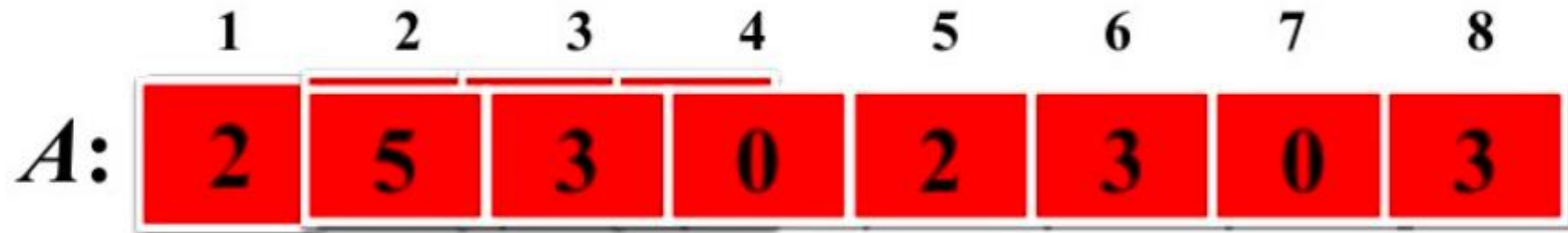
0	1	2	3	4	5
2	0	2	3	0	1

B:

1	2	3	4	5	6	7	8

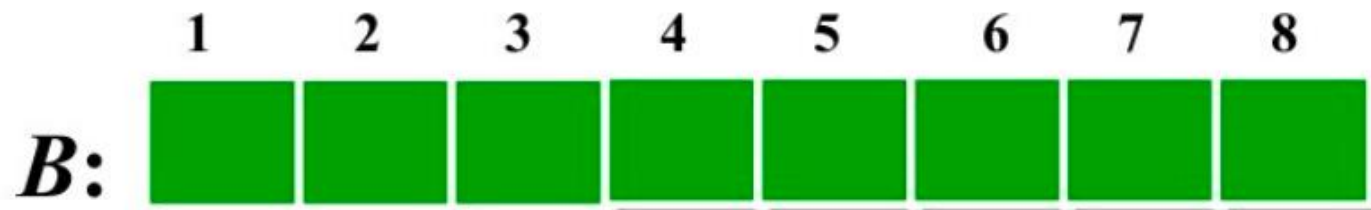
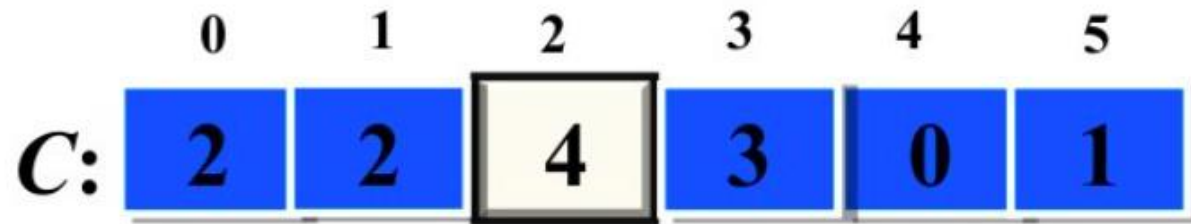
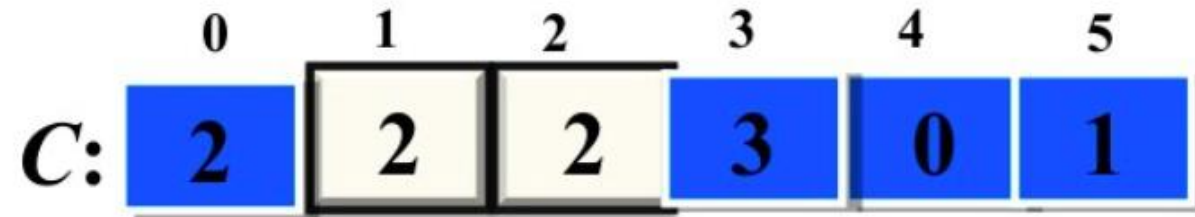
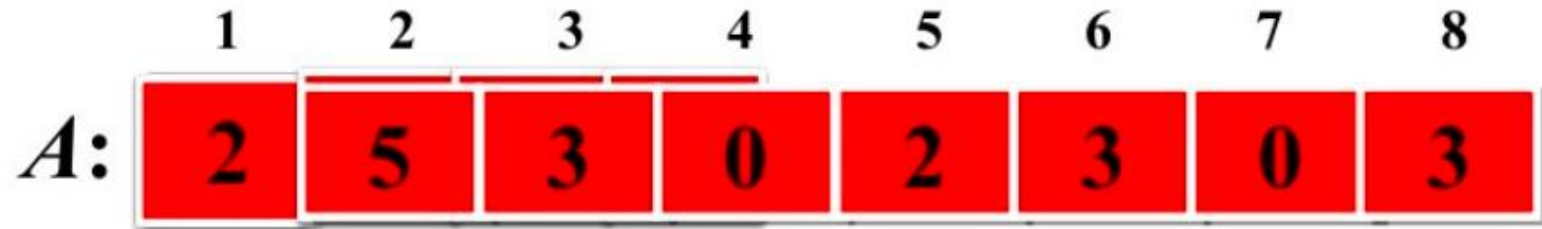
Executing Loop 3

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



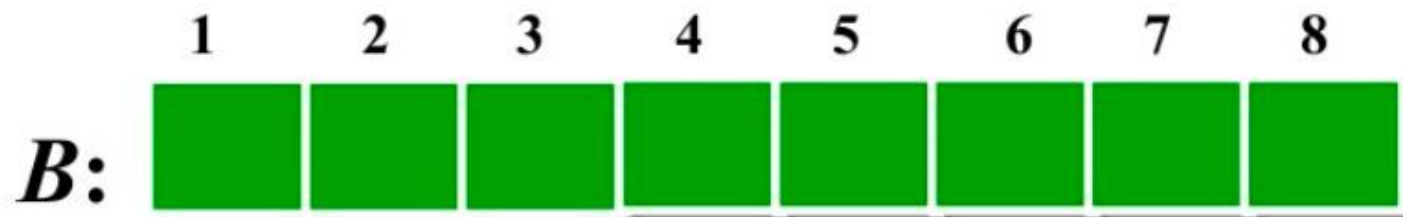
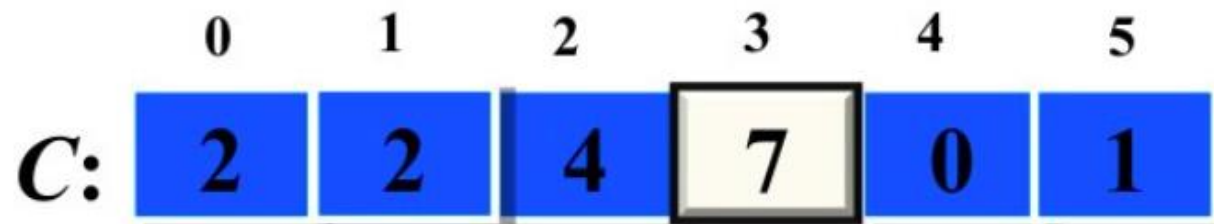
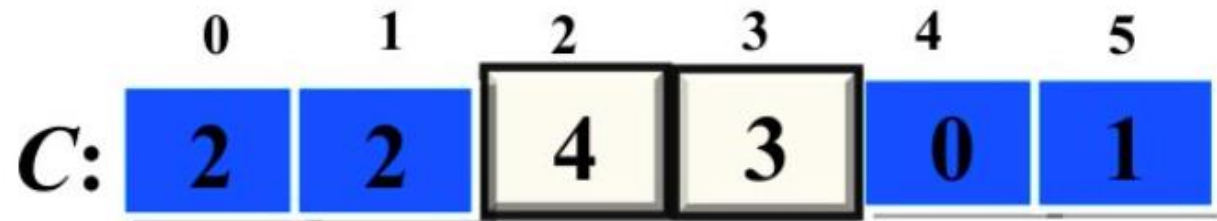
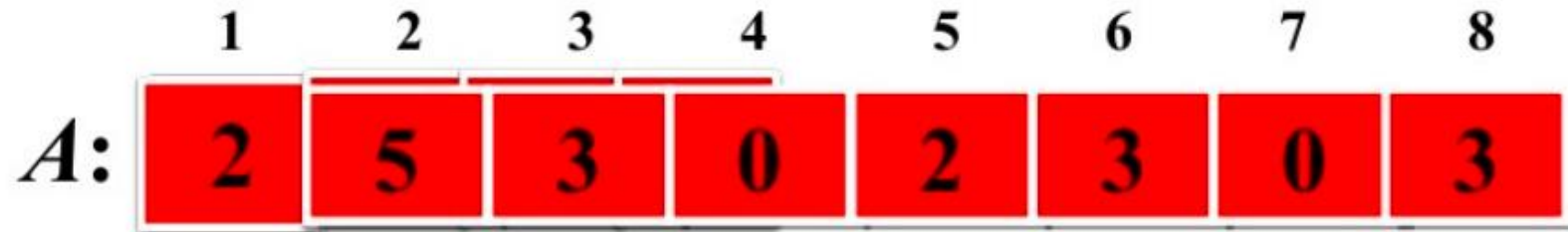
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Executing Loop 3

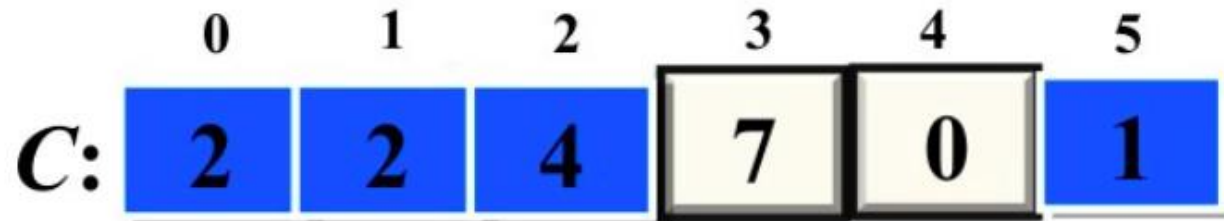
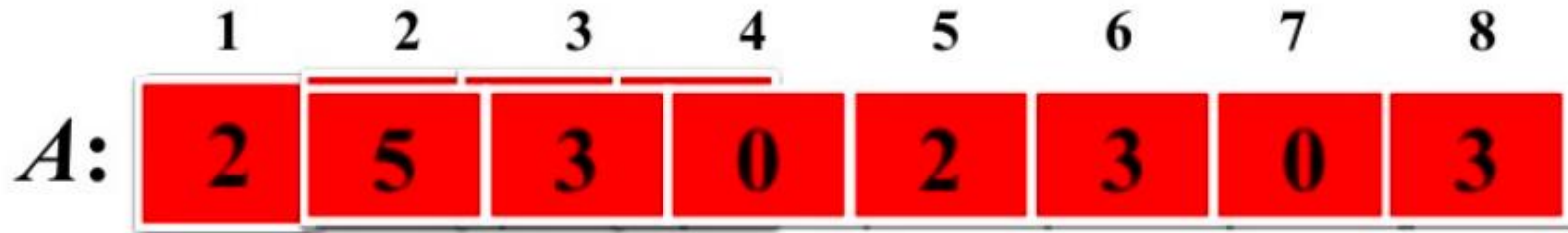
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Executing Loop 3

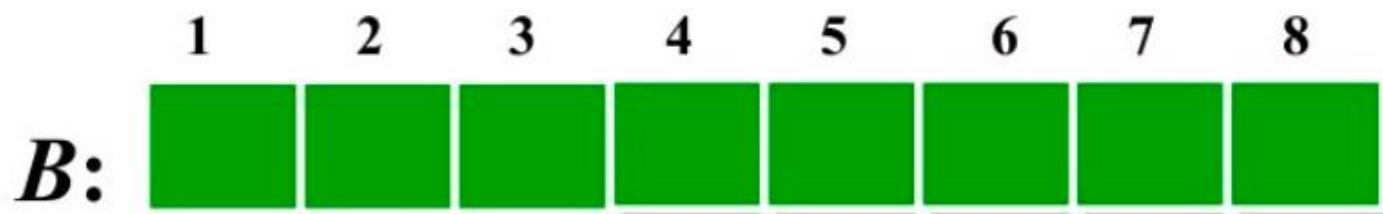
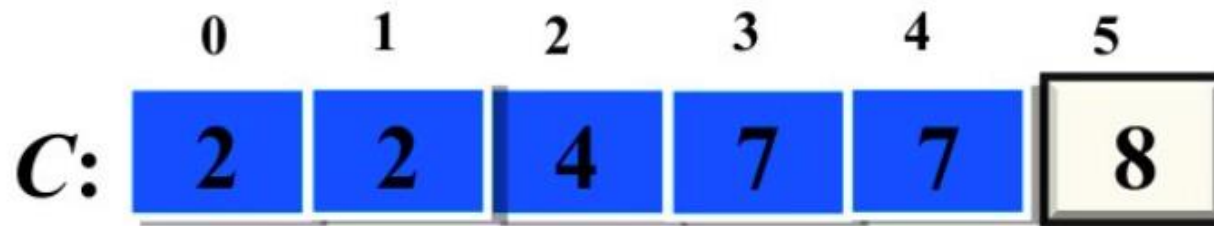
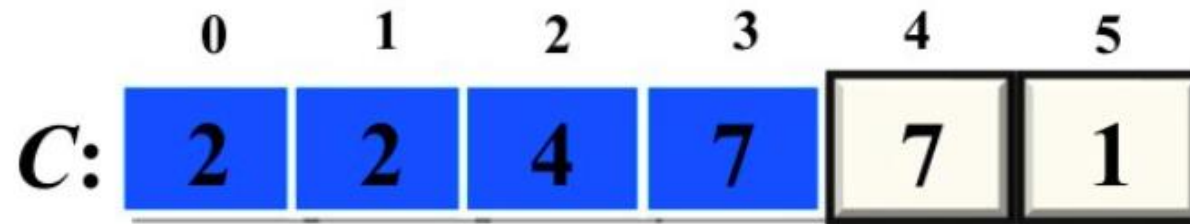
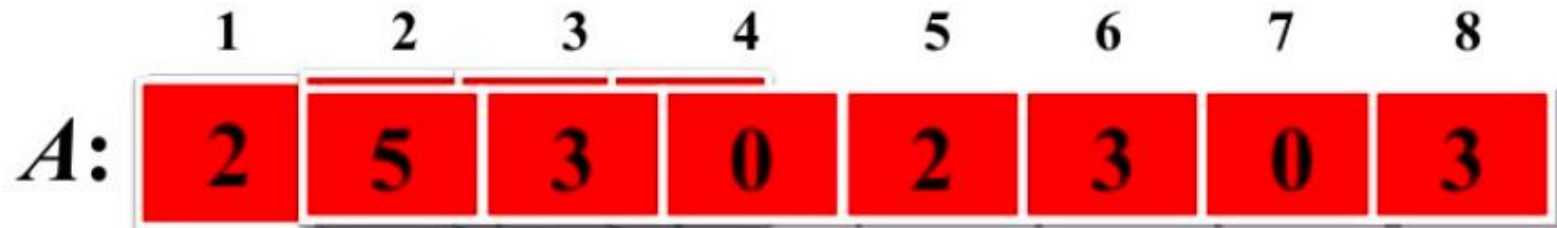
7. for i=1 to k

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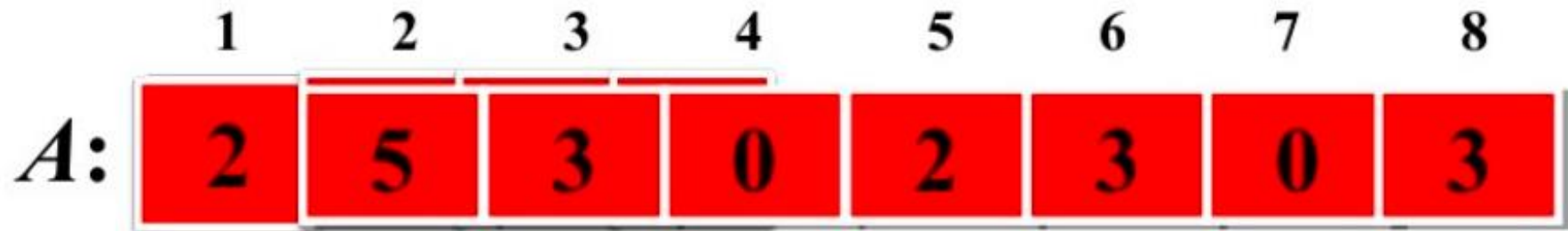


Executing Loop 3

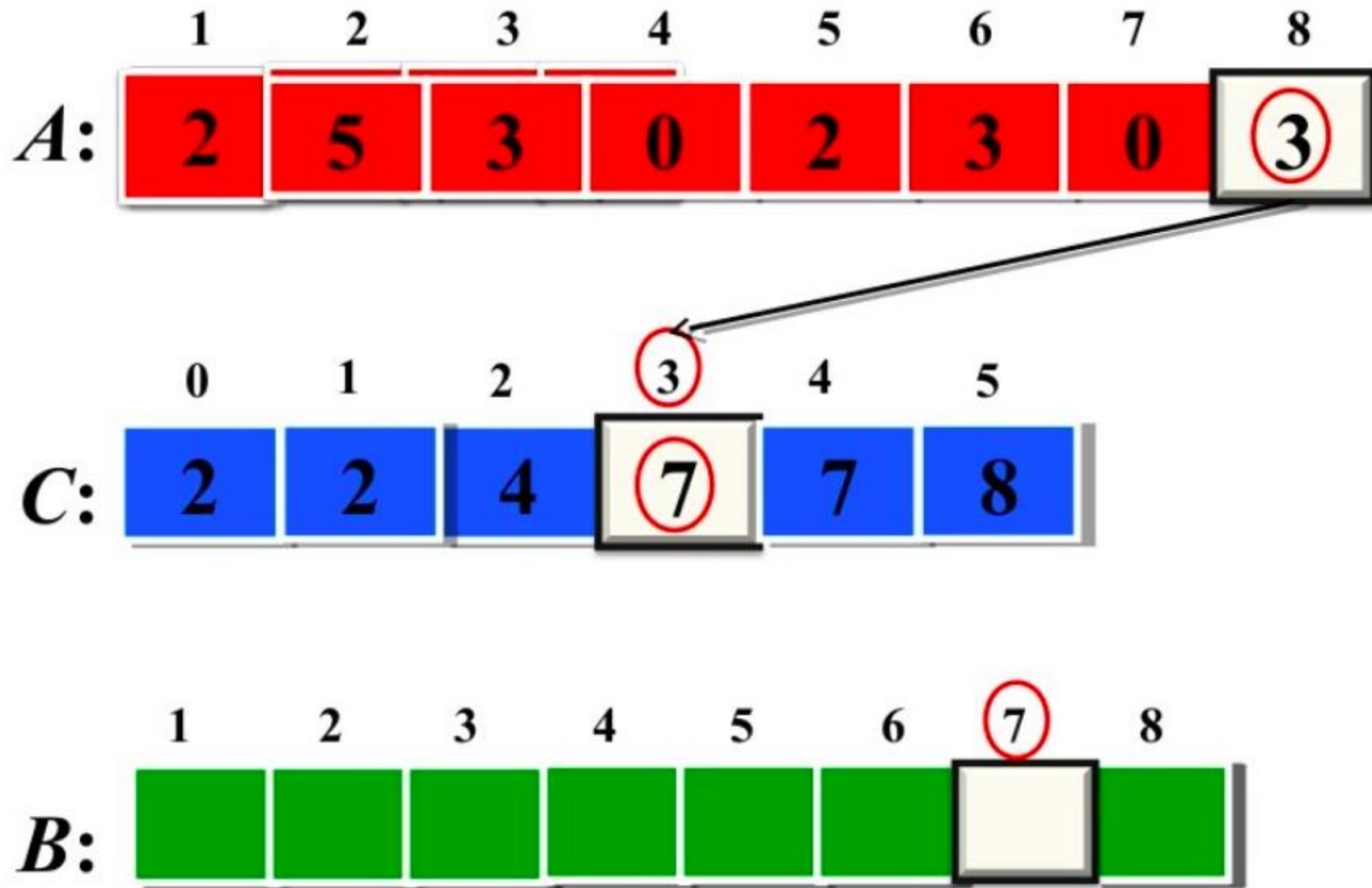
7. for $i=1$ to k
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End of Loop 3

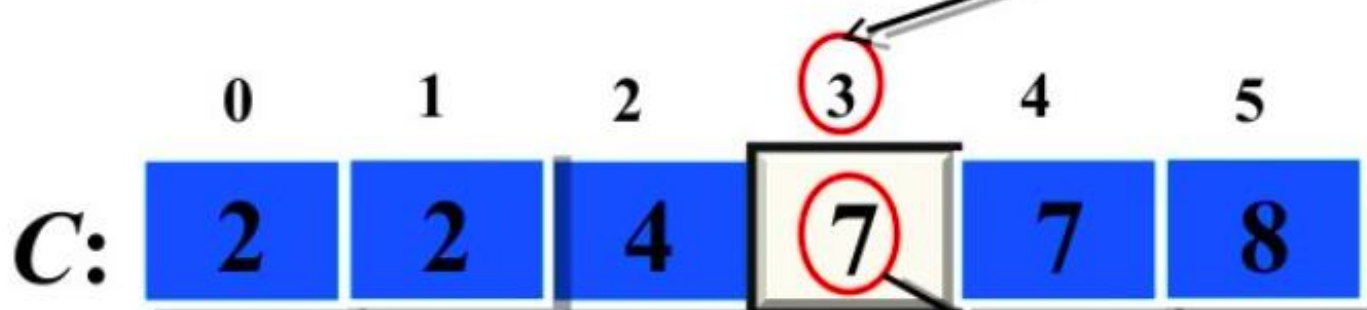
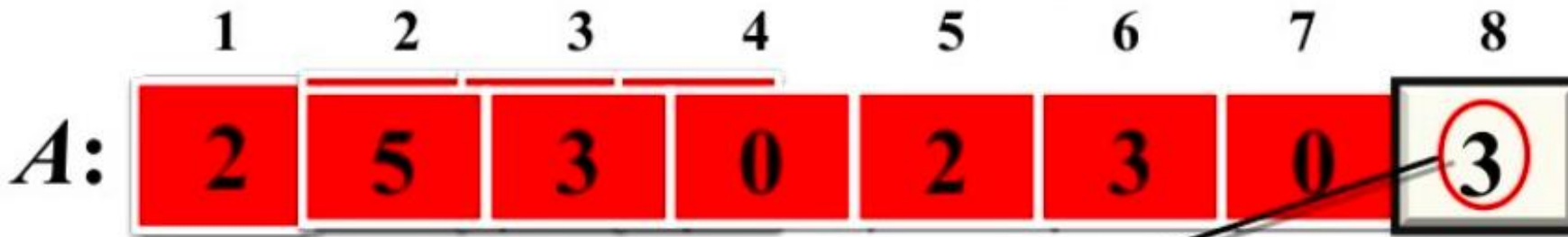


Executing Loop 4

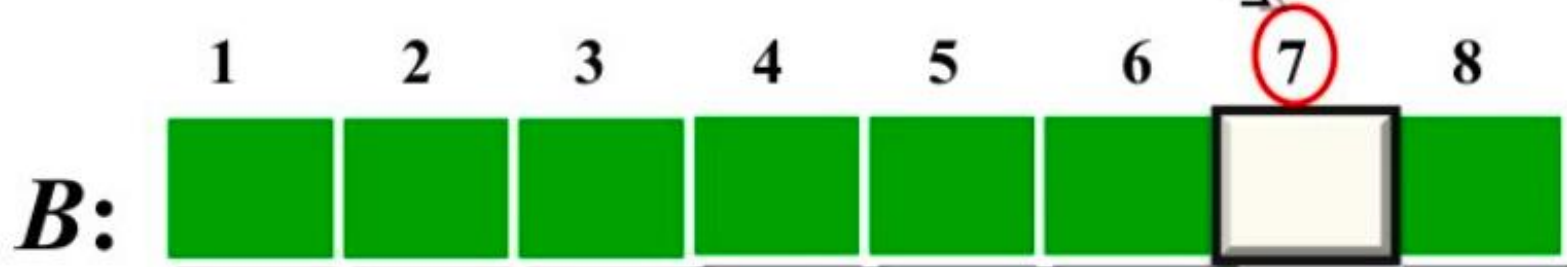


Executing Loop 4

```
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;
```

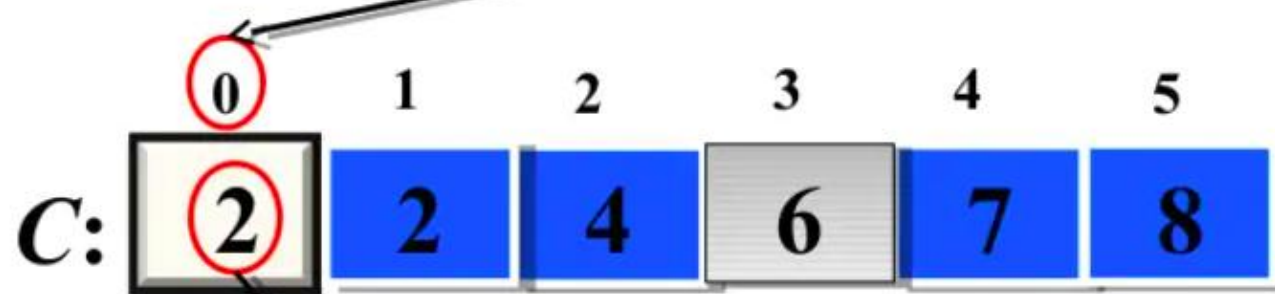
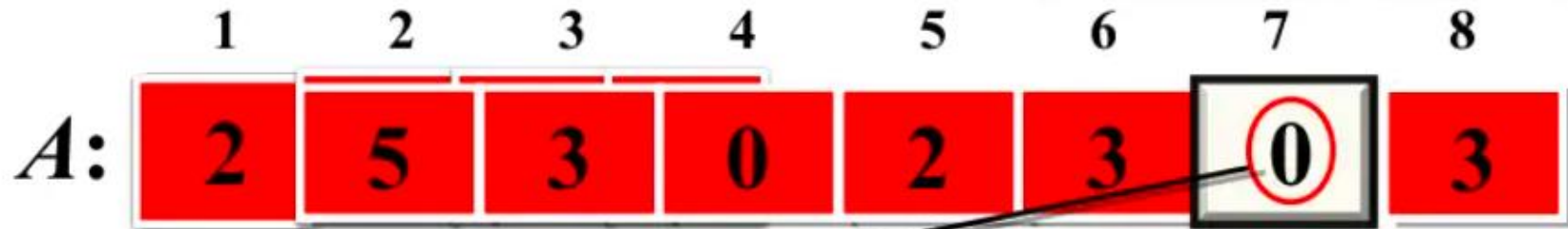


J=8, then $A[j] = A[8] = 3$
And $B[C[A[j]]]$
 $= B[C[3]]$
 $= B[7]$
So $B[C[A[j]]] \leftarrow A[j]$
 $= B[7] \leftarrow 3$



Executing Loop 4

9. for j=n or A.length down to 1
10. $B[C[A[j]]] = A[j];$
11. $C[A[j]] = C[A[j]] - 1;$



J=8, then $A[j] = A[8] = 3$

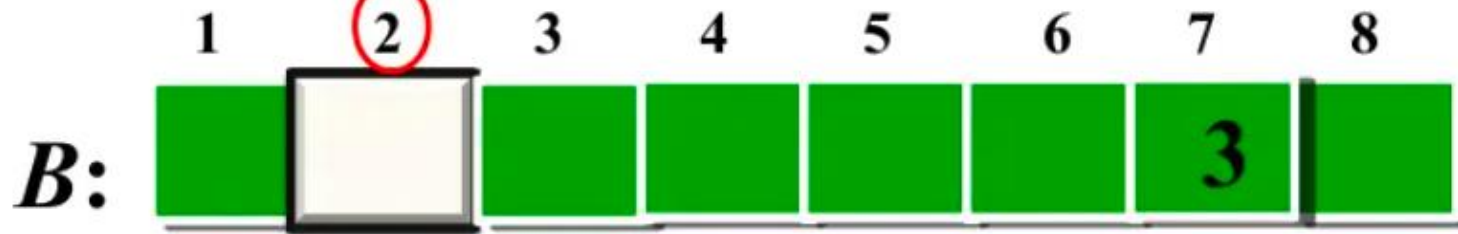
Then $C[A[j]]$

$= C[3]$

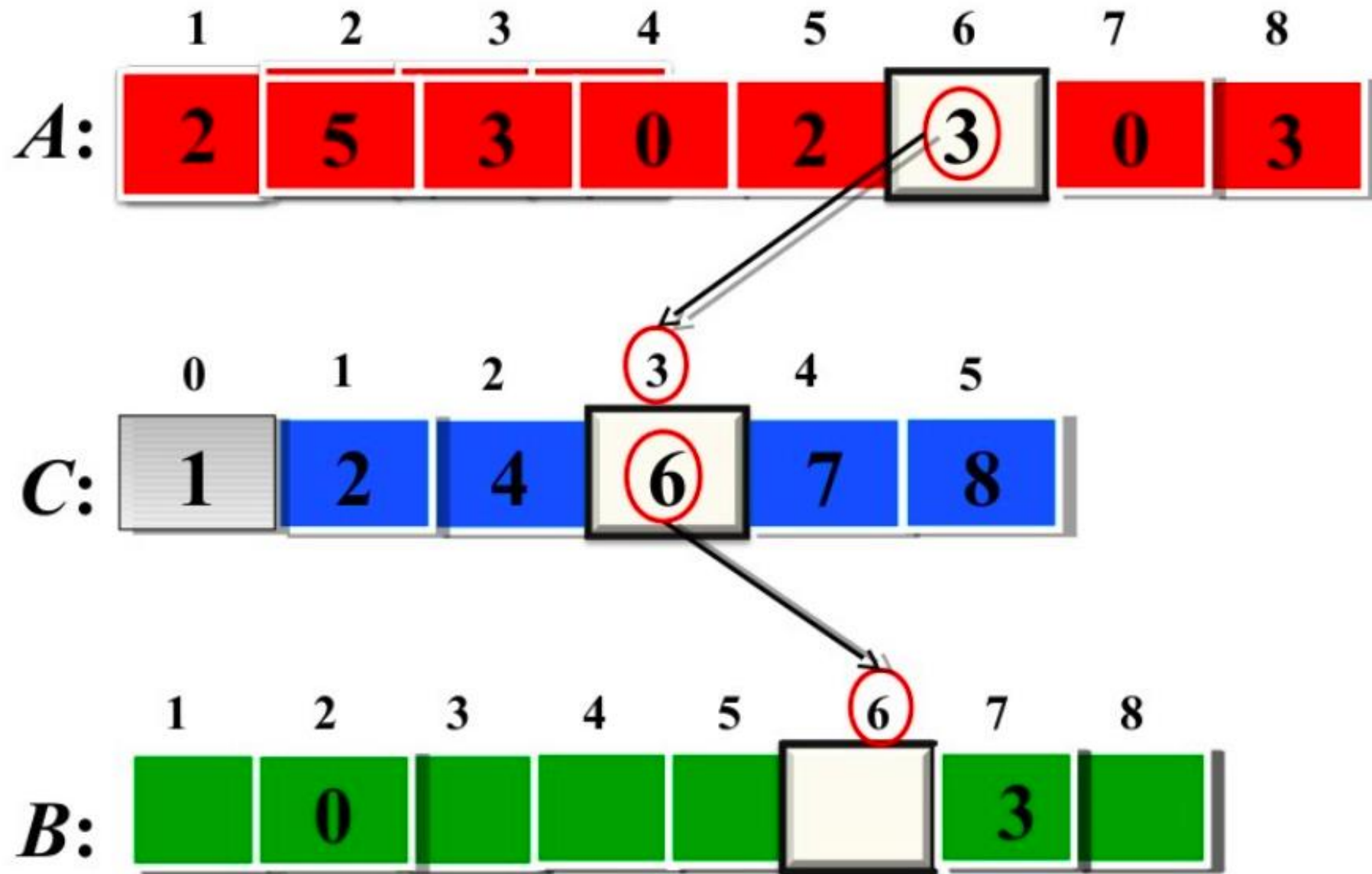
$= 7$

So $C[A[j]] = C[A[j]] - 1$

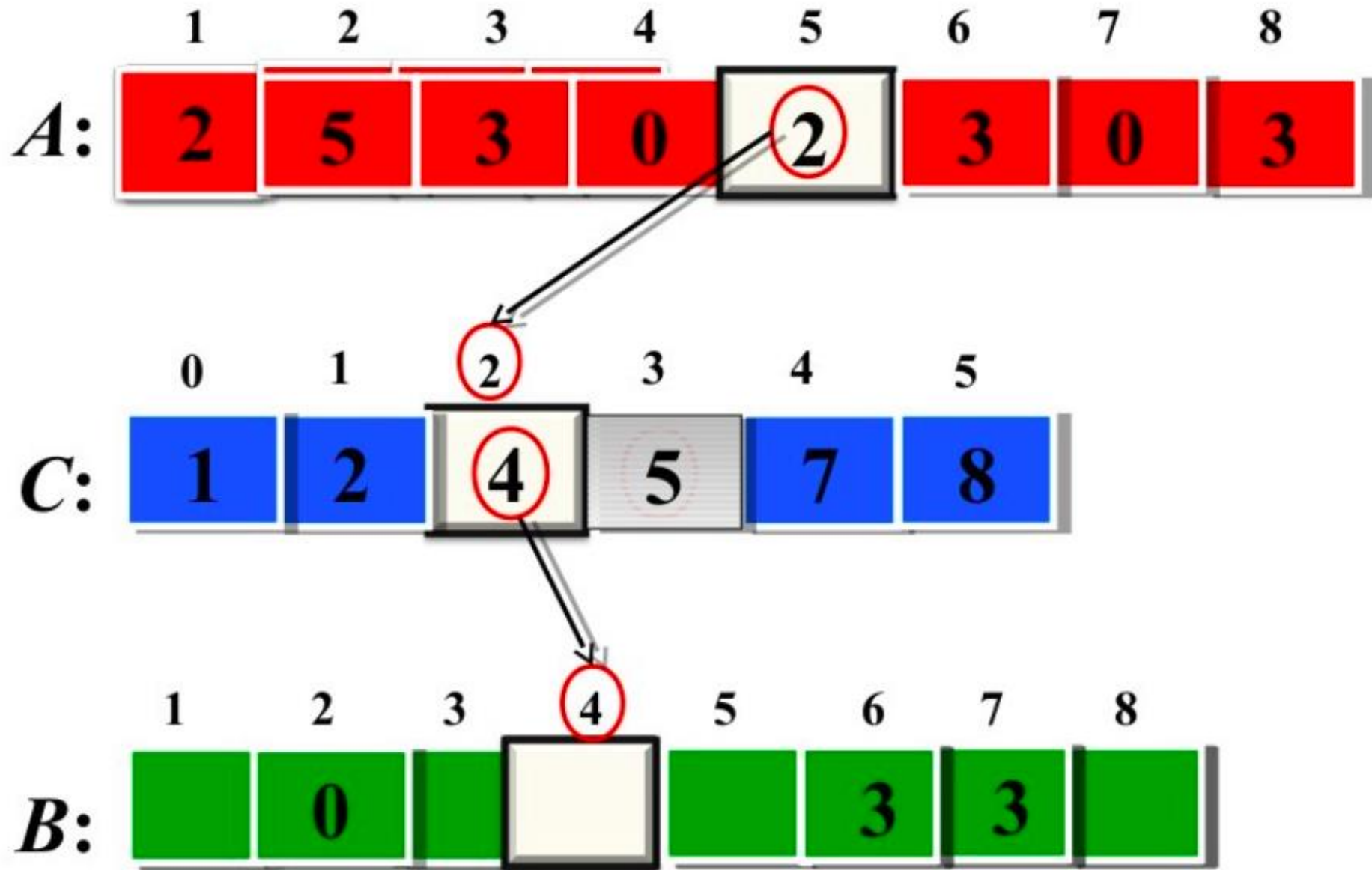
$= 7 - 1 = 6$



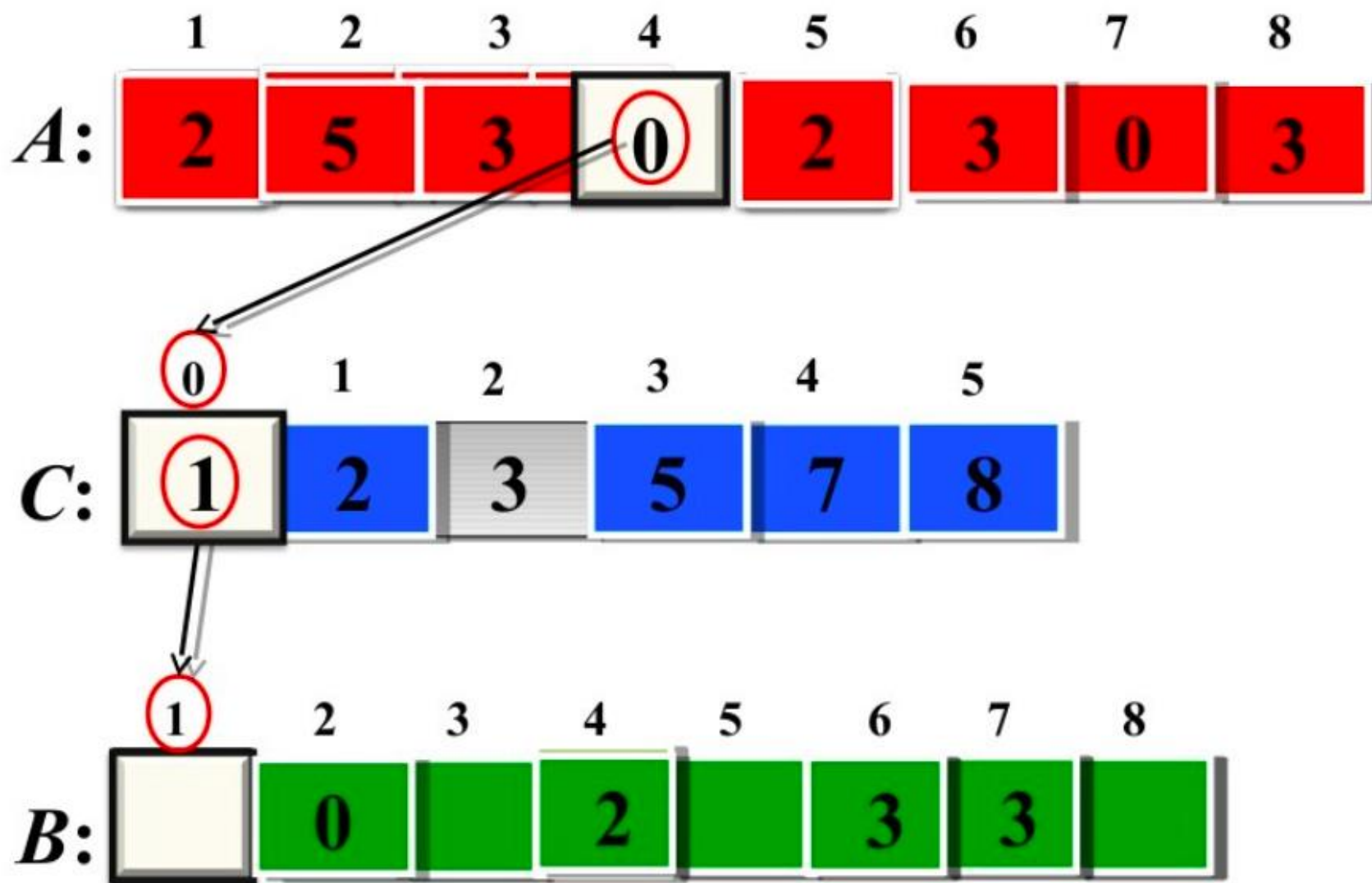
Executing Loop 4



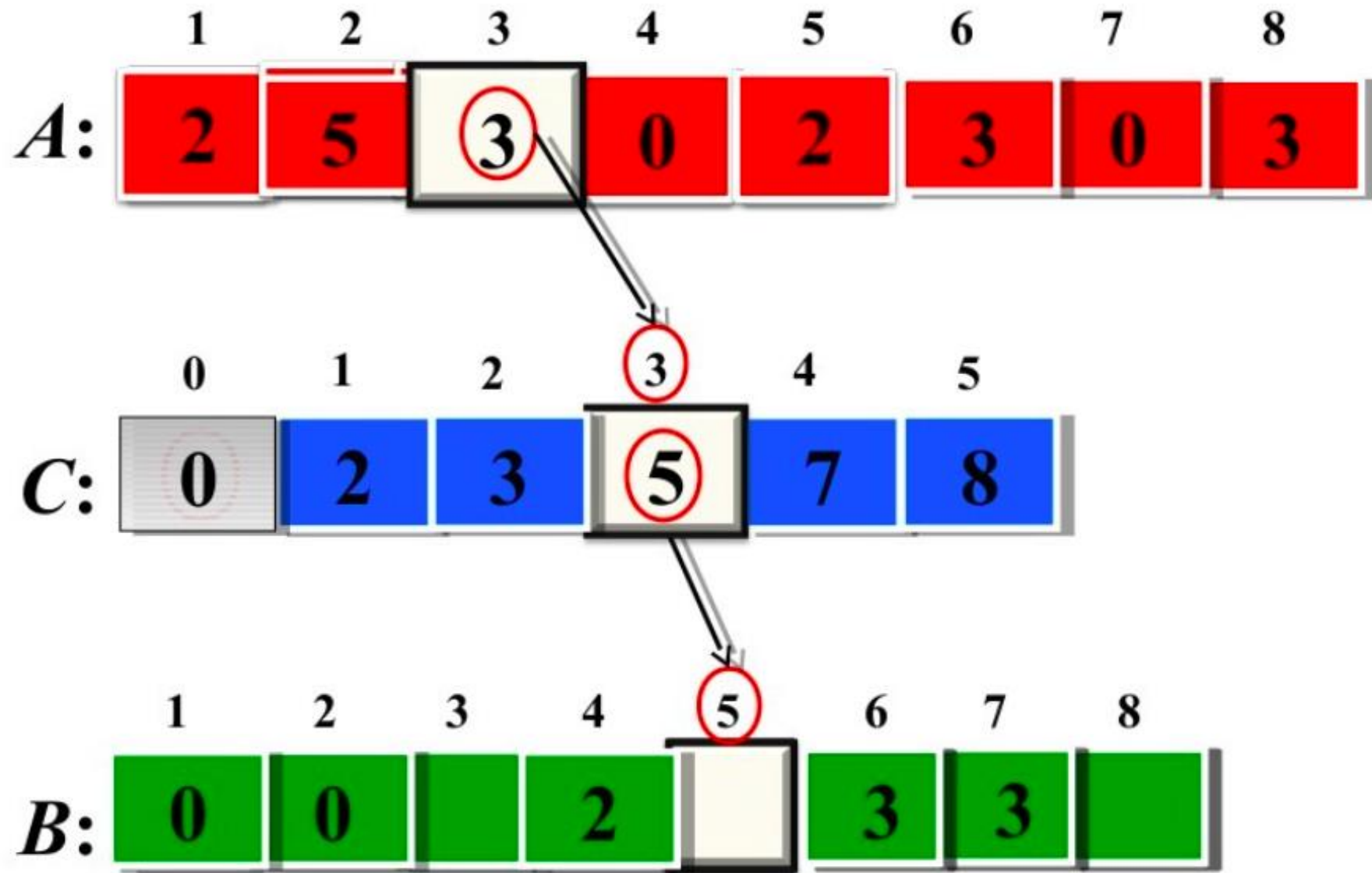
Executing Loop 4



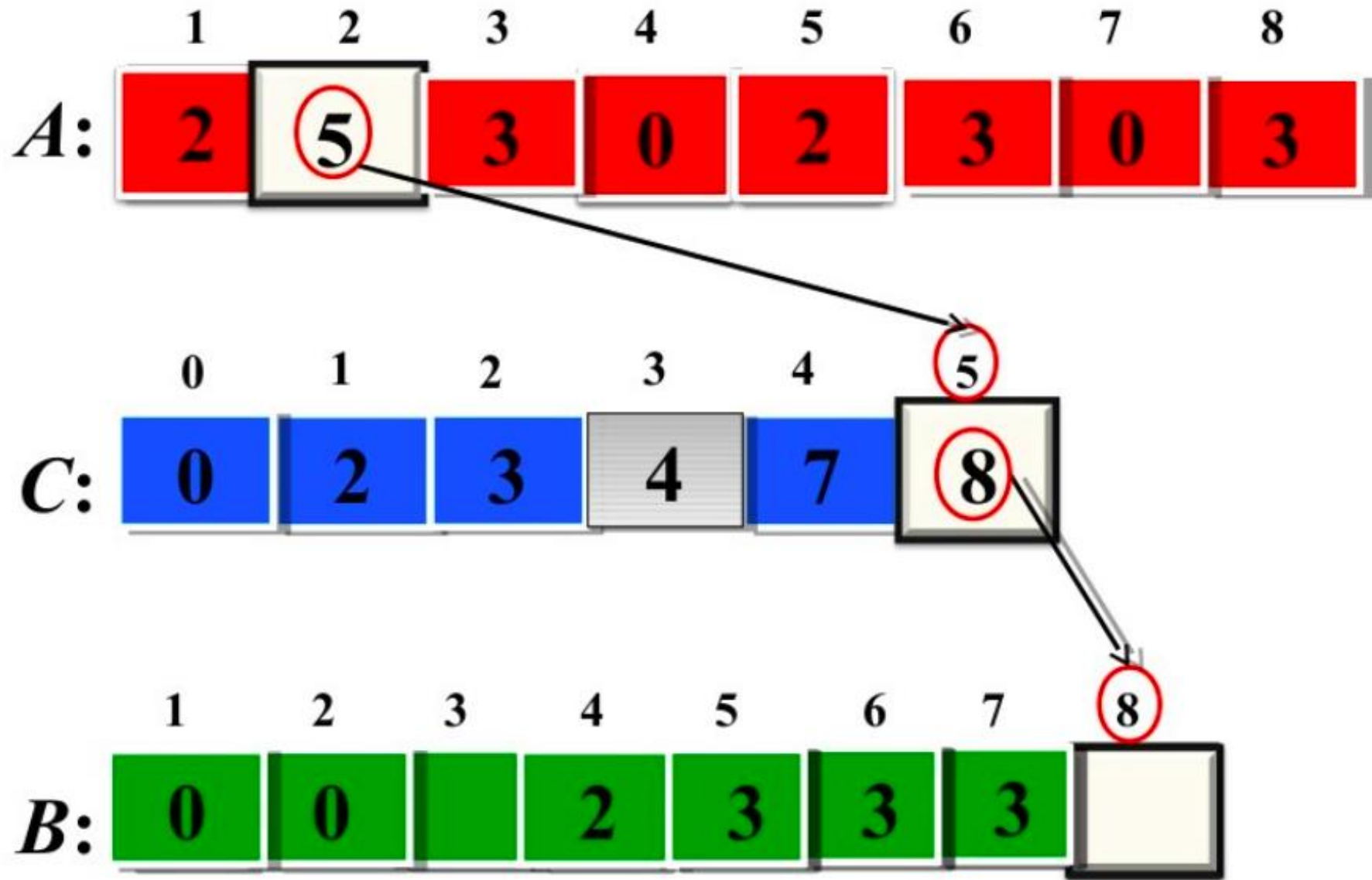
Executing Loop 4



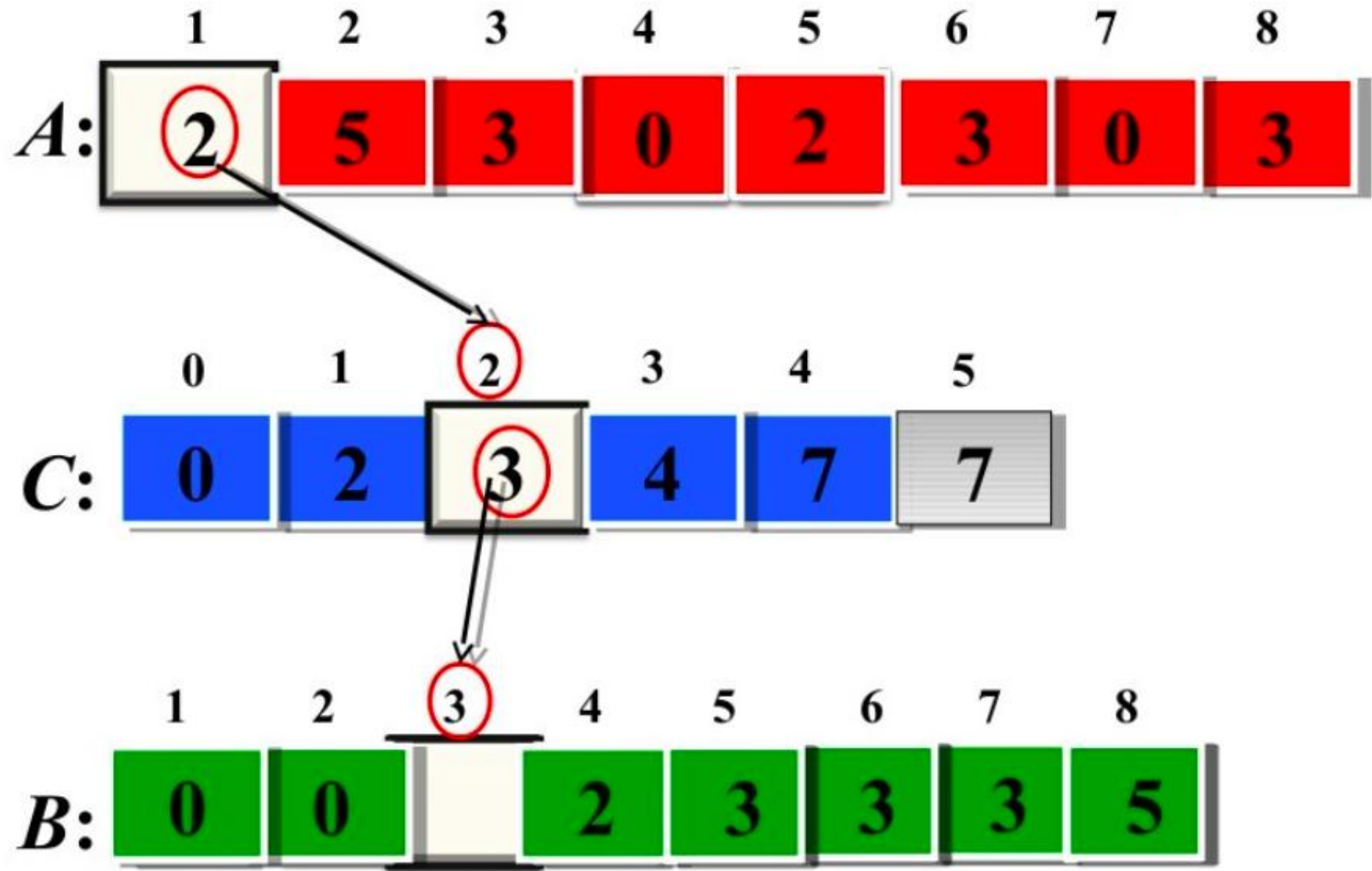
Executing Loop 4



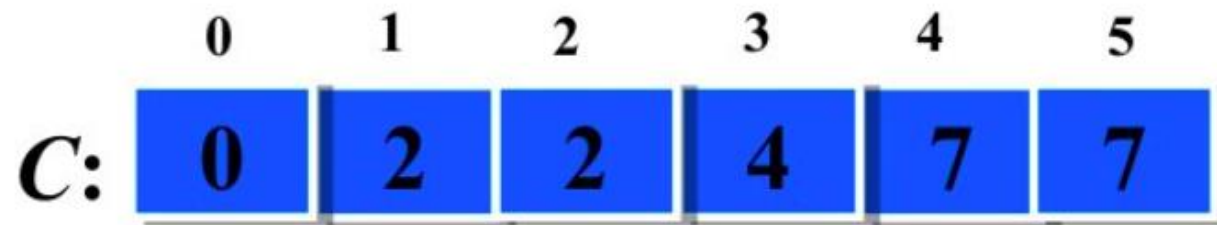
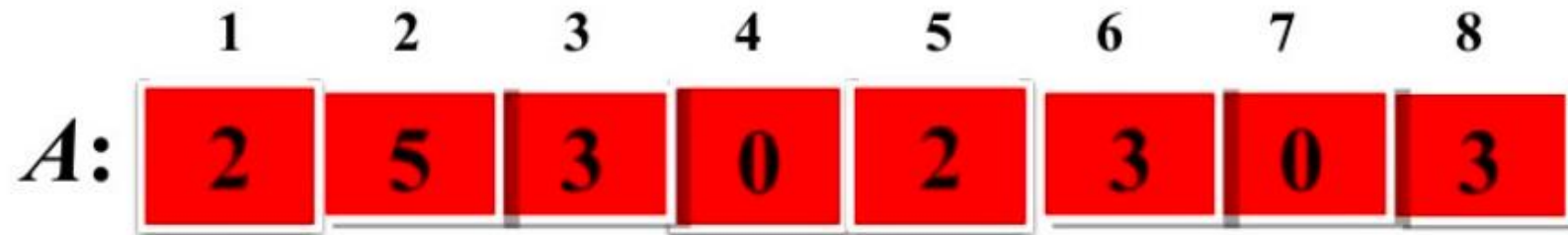
Executing Loop 4



Executing Loop 4



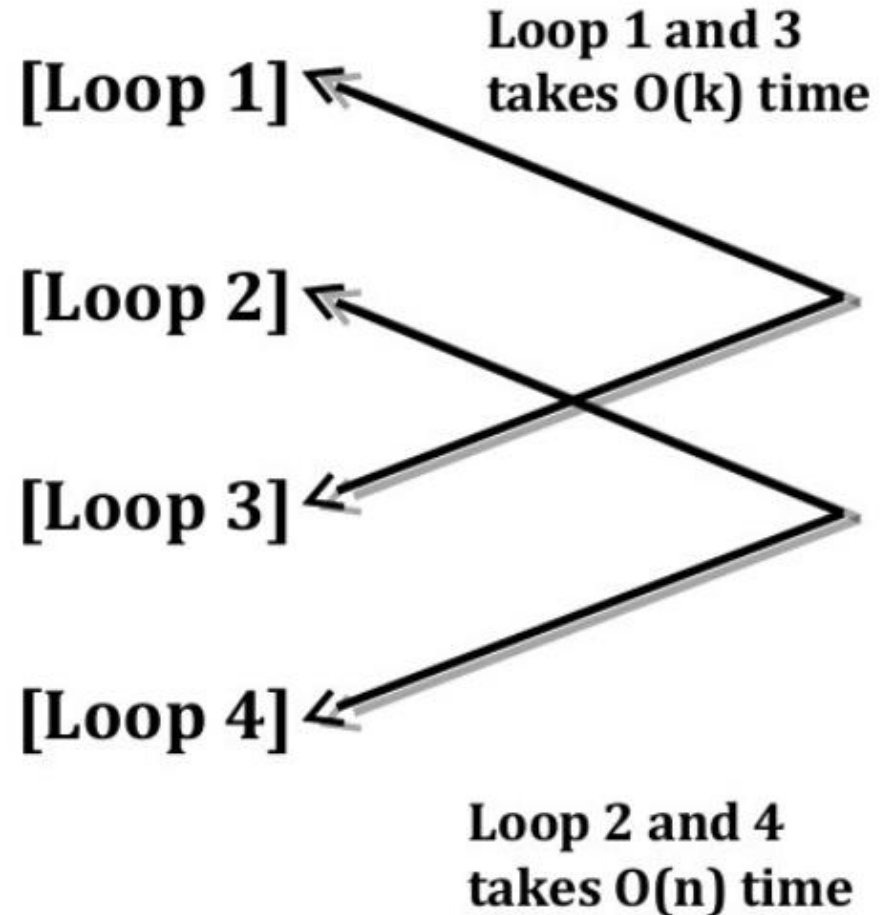
End of Loop 4



Sorted data in Array B

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. $C[i] = C[i] + C[i-1];$
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Time Complexity Analysis

- So the counting sort takes a total time of: $O(n + k)$
- Counting sort is called stable sort.
 - A sorting algorithm is ***stable*** when numbers with the same values appear in the output array in the same order as they do in the input array.

Longest Common Subsequence (LCS):

✓ Problem Statement:

- We are given two strings 's1' and 's2'.
- We need to find the length of the longest subsequence which is common in both the strings.
- subsequence is a sequence that can be derived from another sequence by deleting some elements without changing the order of the remaining elements

✓ Example 1:

- s1 = "elephant"
- s2 = "eretpat"
- Output: 5
- Explanation: The longest substring is "eepat".

✓ Example 2:

- s1 = "houdini "
- s2 = "hdupti"
- Output: 3
- Explanation: The longest substring is "hui".

$$\begin{aligned} 1. & \quad 1 + F(2.8, 2.7) \\ 2. & \quad 0 + F(3.8, 2.7) \\ & \quad 0 + F(2.8, 3.7) \end{aligned}$$

Longest Common Subsequence (Divide & Conquer):

```
int findLCSLengthAux(String s1, String s2, int i1, int i2) {  
    if (i1 == s1.length() || i2 == s2.length())//Base Case  
        return 0;  
  
    int c3 = 0;  
  
    if (s1.charAt(i1) == s2.charAt(i2)) {//If current character in both the string matches, then increase the index by 1 in both the strings.  
        c3 = 1 + findLCSLengthAux(s1, s2, i1 + 1, i2 + 1);  
  
        int c2 = findLCSLengthAux(s1, s2, i1 + 1, i2);//Increase index of 1st String  
  
        int c1 = findLCSLengthAux(s1, s2, i1, i2 + 1);//Increase index of 2nd String  
  
        return Max(c1, c2, c3)  
    }  
}
```

Closest-Pair Problem: Divide and Conquer

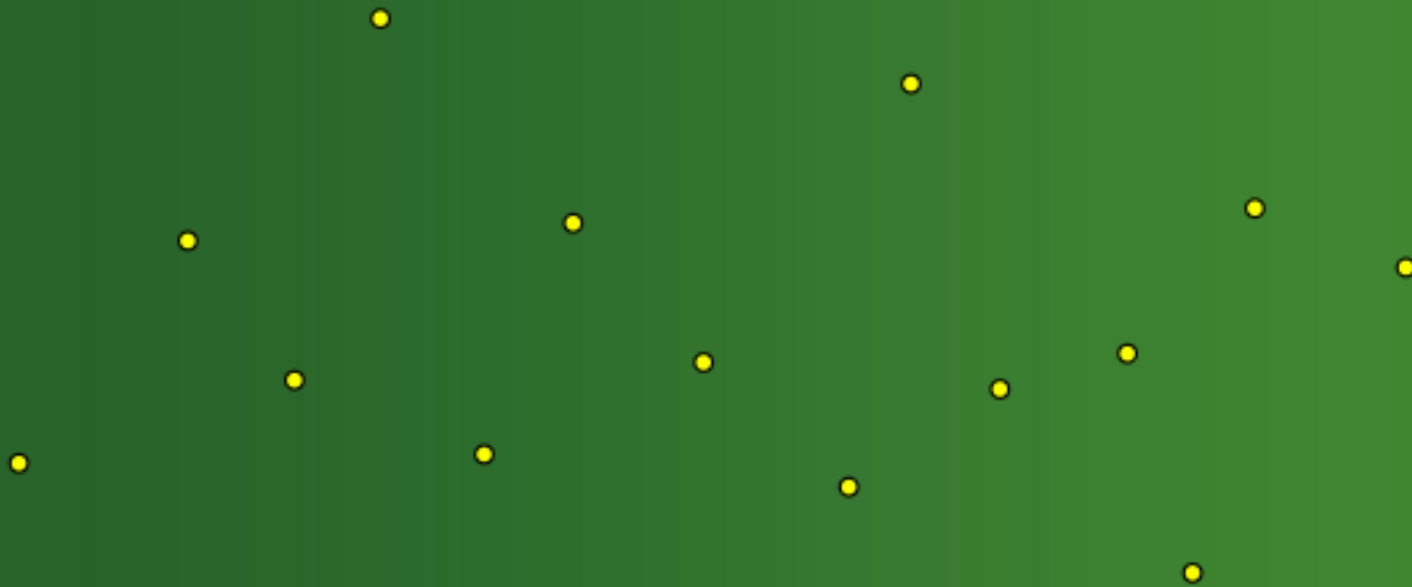
- Brute force approach requires comparing every point with every other point
- Given n points, we must perform $1 + 2 + 3 + \dots + n-2 + n-1$ comparisons.

$$\sum_{k=1}^{n-1} k = \frac{(n-1) \cdot n}{2}$$

- Brute force $\rightarrow O(n^2)$
- The Divide and Conquer algorithm yields $\rightarrow O(n \log n)$
- Reminder: if $n = 1,000,000$ then
 - $n^2 = 1,000,000,000,000$ whereas
 - $n \log n = 20,000,000$

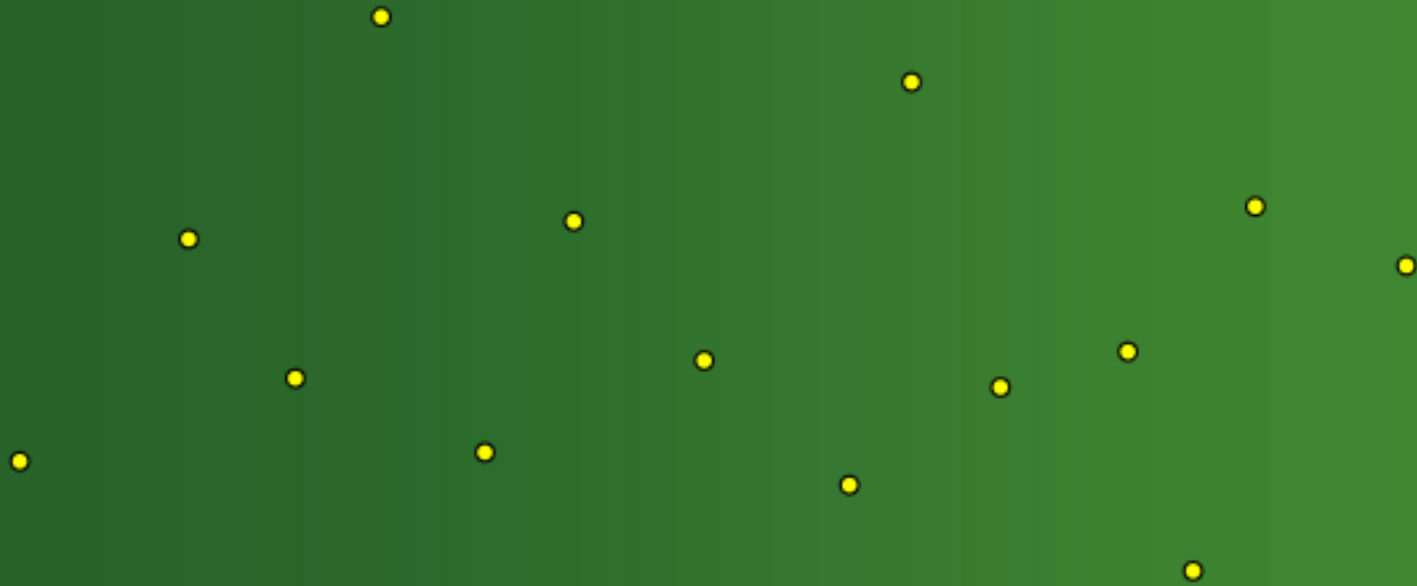
Closest-Pair Algorithm

Given: A set of points in 2-D



Closest-Pair Algorithm

Step 1: Sort the points in one D



Closest-Pair Algorithm

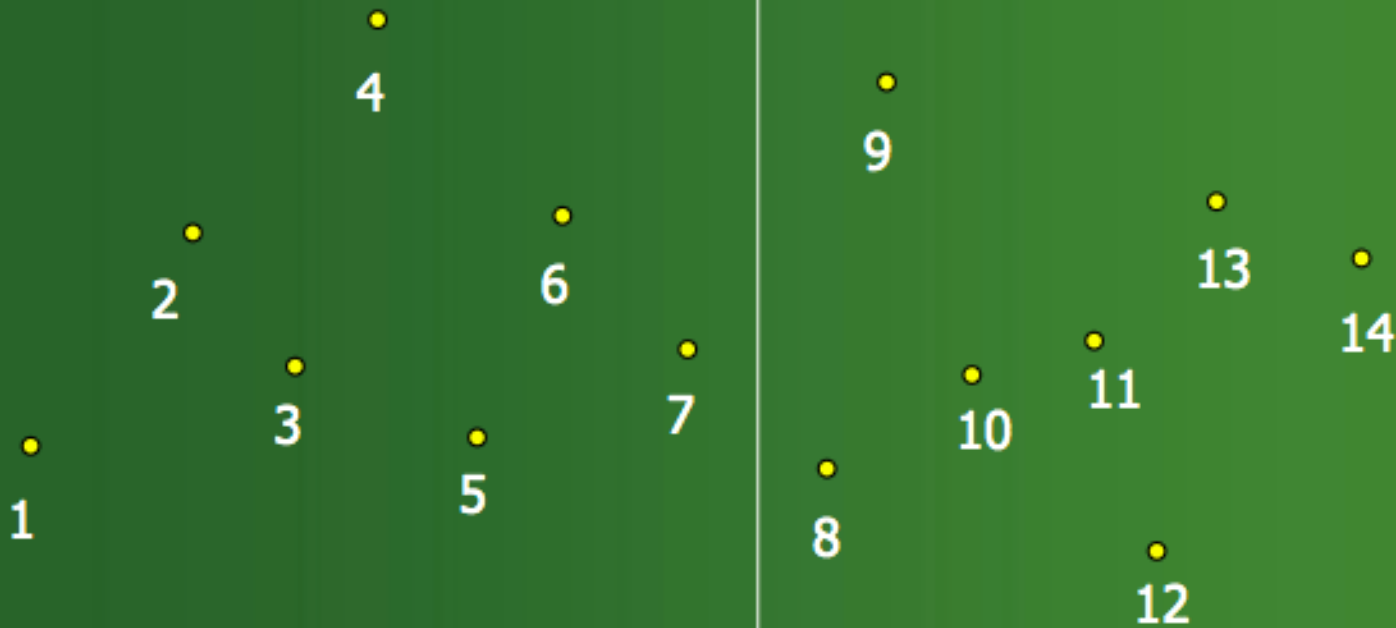
Step 2: Split the points, i.e.,
Draw a line at the mid-point between 7 and 8



Closest-Pair Algorithm

Advantage: Normally, we'd have to compare each of the 14 points with every other point.

$$(n-1)n/2 = 13*14/2 = \mathbf{91 \text{ comparisons}}$$



Sub-Problem 1

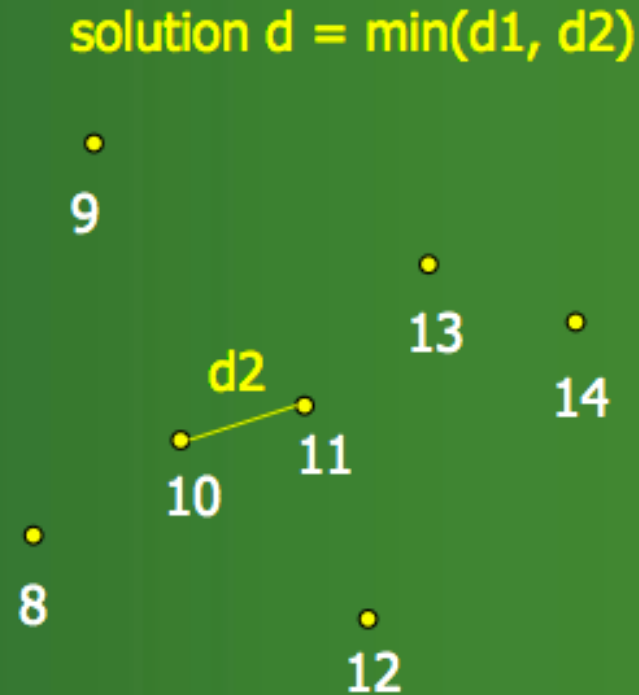
Sub-Problem 2

Closest-Pair Algorithm

Advantage: Now, we have two sub-problems of half the size. Thus, we have to do $6 \cdot 7/2$ comparisons twice, which is 42 comparisons



Sub-Problem 1



Sub-Problem 2

Closest-Pair Algorithm

Advantage: With just one split we cut the number of comparisons in half. Obviously, we gain an even greater advantage if we split the sub-problems.



Sub-Problem 1

$$d = \min(d1, d2)$$



Sub-Problem 2

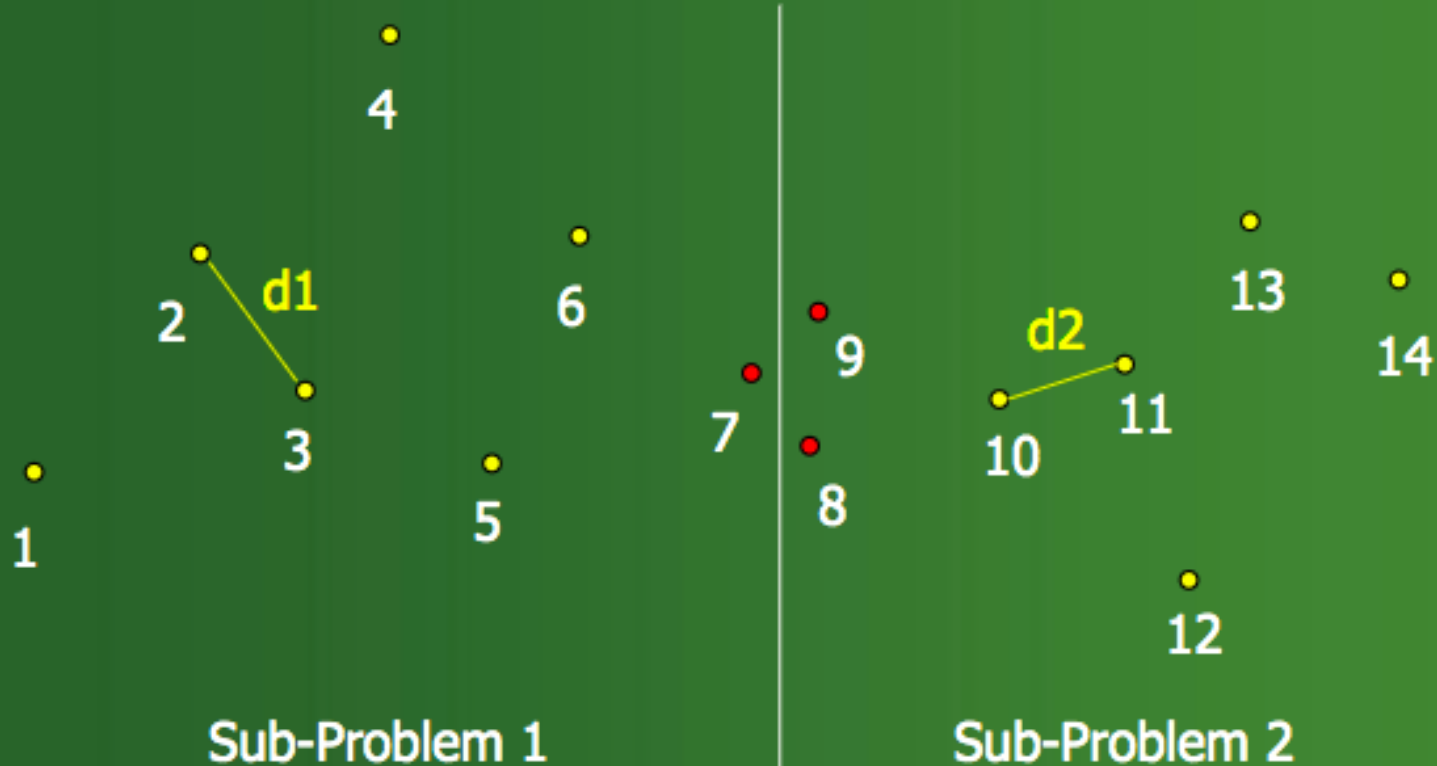
Closest-Pair Algorithm

Problem: However, what if the closest two points are each from different sub-problems?



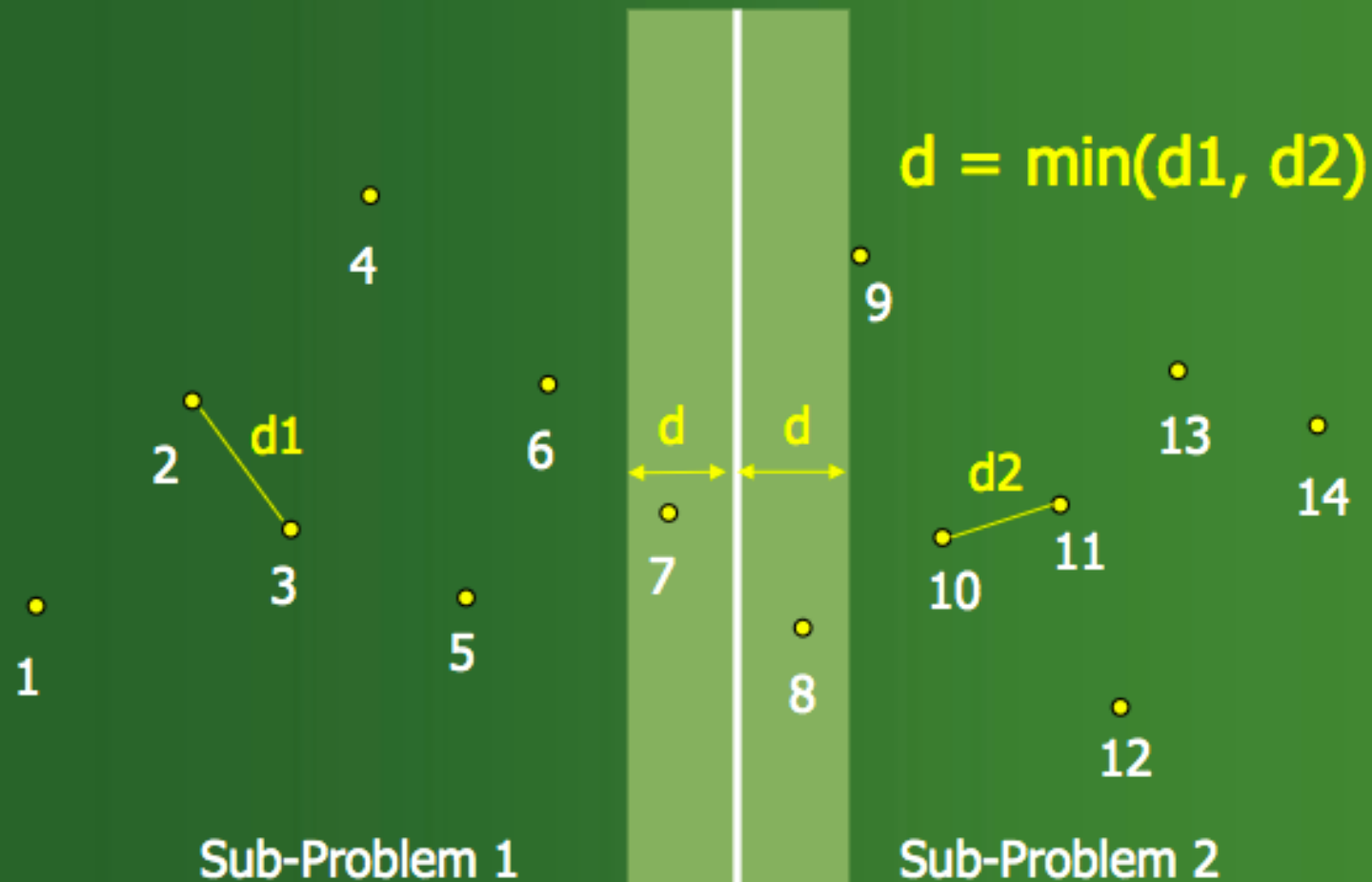
Closest-Pair Algorithm

Here is an example where we have to compare points from sub-problem 1 to the points in sub-problem 2.



Closest-Pair Algorithm

However, we only have to compare points inside the following "strip."



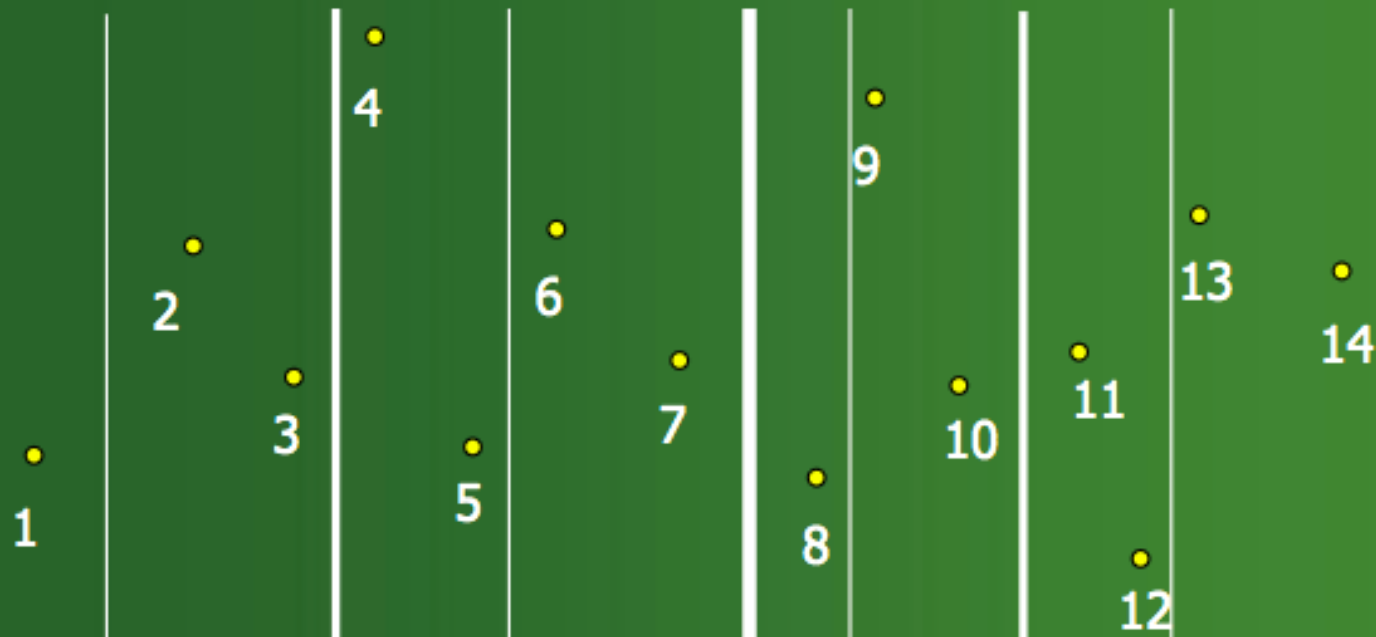
Closest-Pair Algorithm

Step 3: But, we can continue the advantage by splitting the sub-problems.



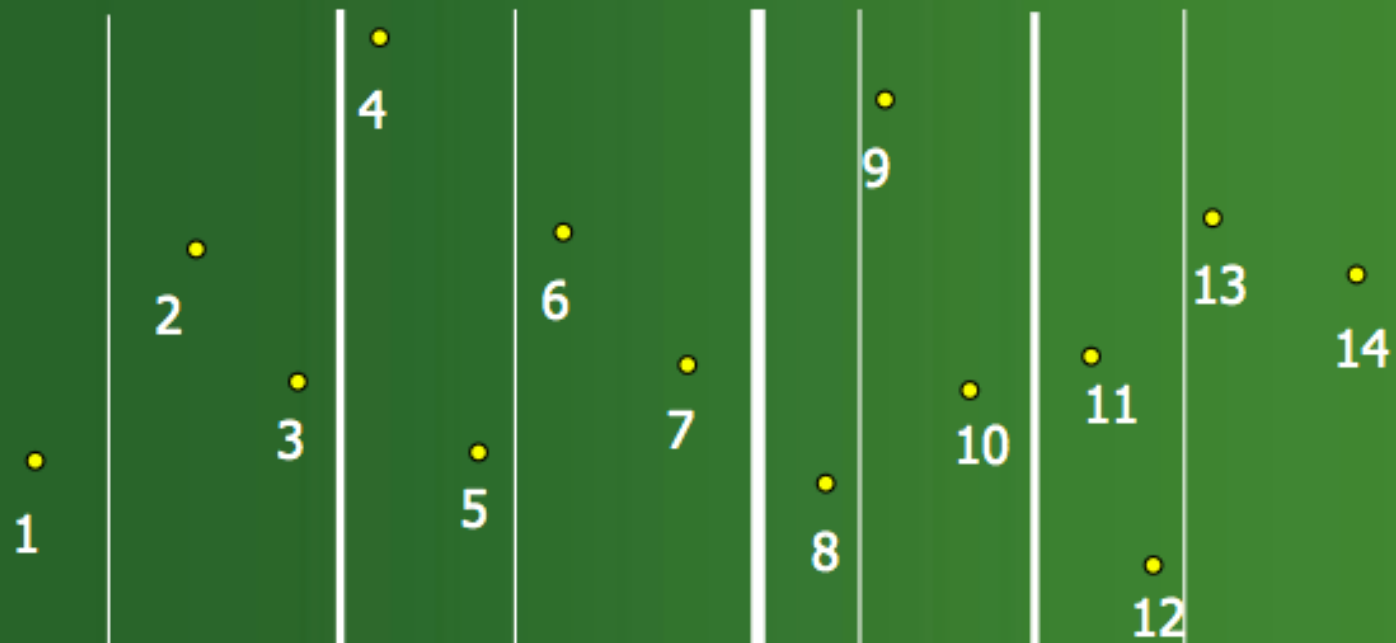
Closest-Pair Algorithm

Step 3: In fact we can continue to split until each sub-problem is trivial, i.e., takes one comparison.



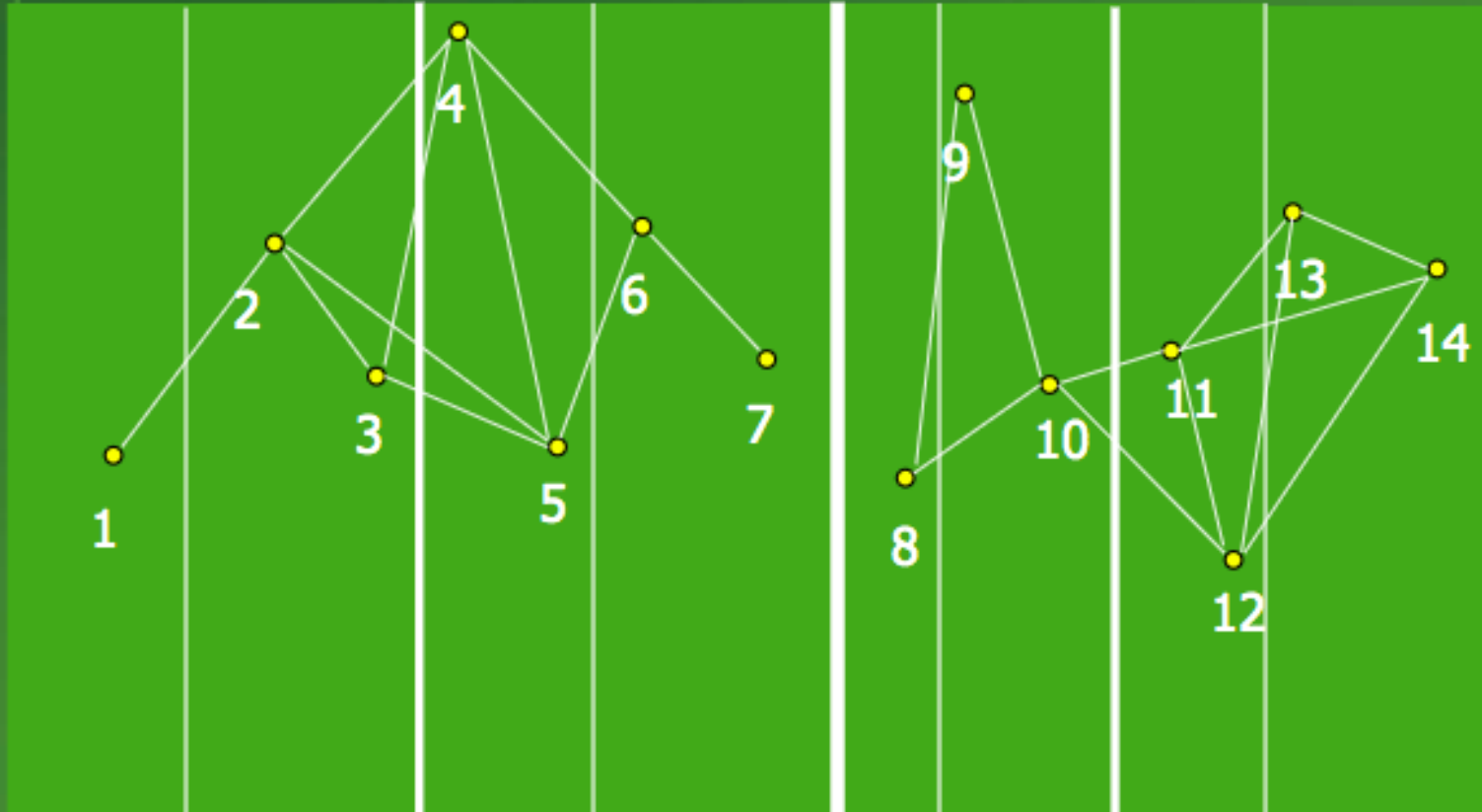
Closest-Pair Algorithm

Finally: The solution to each sub-problem is combined until the final solution is obtained



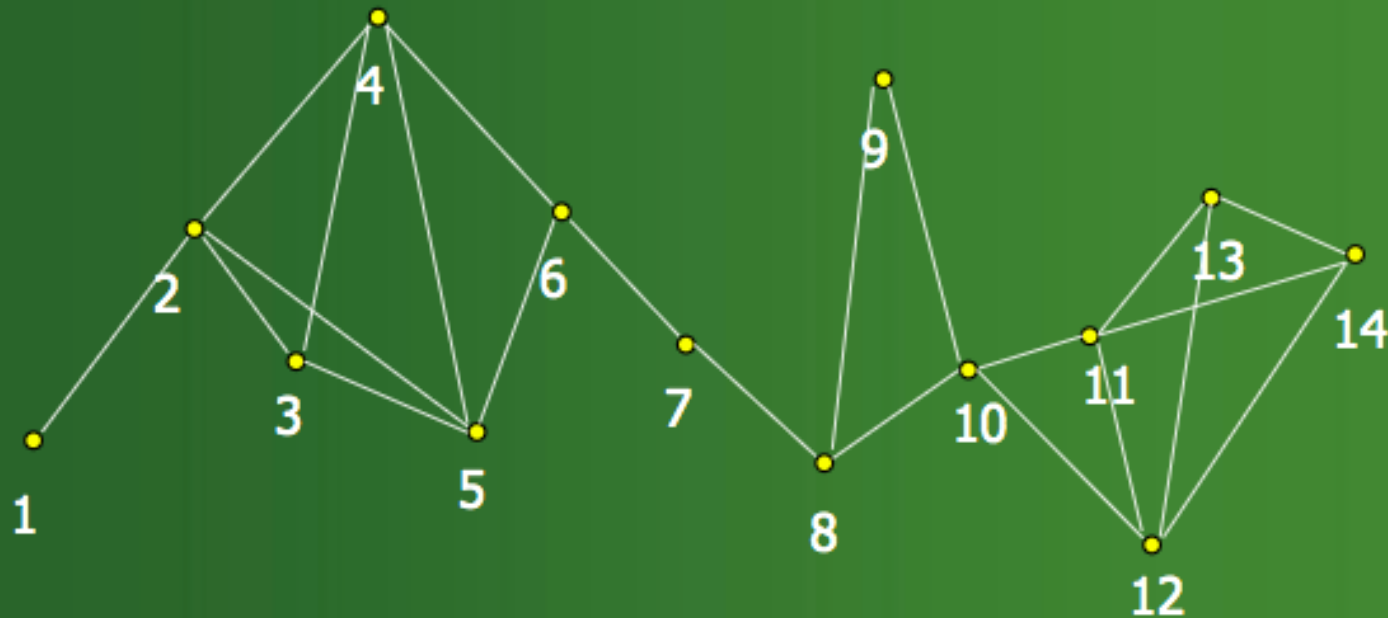
Closest-Pair Algorithm

Finally: On the last step the 'strip' will likely be very small. Thus, combining the two largest sub-problems won't require much work.



Closest-Pair Algorithm

- In this example, it takes 22 comparisons to find the closest-pair.
- The brute force algorithm would have taken 91 comparisons.
- But, the real advantage occurs when there are millions of points.





Thank
you