

# Countinsort

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## Abstract

*When it comes to sorting elements - in this case numerical values - most of the time we lack a lot of information to reach a good overall performance. When we know the maximum key value in our list of elements as well as the number of elements we can employ an implementation of Countinsort to get them sorted. It basically counts the occurrences of each key value and exploits these counts as pointers to their final position. Deeper analysis shows that in general the runtime is **linear** upon the number of elements but we can easily produce edge-cases where the overall runtime explodes.*

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## 1 Problem and Motivation

Sorting elements topologically is a problem that we are facing in many different situations. A popular example would be sorting contestants in a leaderboard. Depending on the application we want to - and sometimes even must - aim for the best runtime complexity. We do not want to wait another season to determine the winner of a football-match, do we? Since we have to touch every element at least once we can already tell that we have  $\Omega(n)$  as our lower boundary in sorting. One way to reach that runtime is an implementation of **Countingsort**.

## 2 Solution

### 2.1 Overview

In order for **Countingsort** to work we assume that our input only consists of numbers; more specifically positive integer values. The reason for this comes clear in **Phase 2**. In total the algorithm consists of three phases.

### 2.2 Phase 1 (*Counting-phase*)

This phase is pretty straight forward. We simply go over all our elements and count their occurrence. For that we introduce a counting array **Counts** which employs a length of **k** where **k** is the maximum value a key employed in our input. **Side note:** Even if unknown finding the maximum key only takes at most  $n \in O(n)$  steps.

**Input:**  $A = [a_1, a_2, \dots, a_n]$ ,  $n$  = Number of elements in **A**

**Output:**  $A' = [a'_1, a'_2, \dots, a'_n]$  with  $\forall a'_i, a'_j; 1 \leq i < j \leq n : a'_i \leq a'_j$

**function** COUNTINGSORT(**A**, **n**)

$k = \text{Maximum}(A)$

**Counts** = New Array of size **k**

$\triangleright$  Filled with zeroes

**for**  $i = 1 \dots n$  **do**

$value = A[i]$

**Counts**[ $value$ ] += 1

**end for**

    [...]

$\triangleright$  End of Counting-Phase

**end function**

### 2.3 Phase 2 (*Pointing Phase*)

Now that we know how many times each key exists in our array we need to determine the final position of each key. The idea here is going to be that our **Counts** array already holds the respective keys in a sorted order. The only thing missing is the respective offsets in the sorted array.

Let  $k_1, k_2, \dots, k_n = \mathbf{k}$  be the possible keys going by **Counts** then **Counts**( $k_i$ ) represents the remaining numbers of elements of value  $k_i$  with  $1 \leq i \leq n$ . After **Phase 1** these represent the exact counts for each key. For simplicity let's assume that every key  $k_i$  exists at least once in our input. The lowest key  $k_1$  obviously sits at the very beginning of the final array. The first element representing the second lowest key  $k_2$  sits at the next position after the last element of  $k_1$ . This pattern applies for every pair of key  $k_i, k_j$  with  $k_i < k_j$ .

To retrieve a list of final positions after **Phase 1** we need to produce...

$$\text{Counts}(k_i) = \sum_{i=1}^{i-1} \text{Counts}(i)$$

The fact that we potentially have no occurrences for some of these keys does nothing to the overall idea of this pointers adjustment because they will have

an initial value of zero and therefore do nothing in the sums portraied above. Recursively this whole setup boils down to...

$$\mathbf{Counts}(i) = \mathbf{Counts}(i) + \mathbf{Counts}(i-1)$$

... beginning from the lowest key going through to the highest key.

```

function COUNTINGSORT(A, n)
    [...]                                     ▷ End of Counting-Phase

    for  $i = 2 \dots k$  do
         $\mathbf{Counts}[i] += \mathbf{Counts}[i - 1]$ 
    end for
    [...]                                     ▷ End of Pointing-Phase
end function
```

## 2.4 Phase 3 (*Sorting Phase*)

The final part of the algorithms is copying the elements to the respective positions that we now hold in our **Counts** array. For that part we go over our input elements another time and for each element we retrieve their final position from **Counts** . Finally we copy the element over to the retrieved postion in a array we call **Result** of length **n**. This yeilds the following pattern...

$$\begin{aligned} \mathbf{Result}[\mathbf{Counts}(x)] &= x \\ \mathbf{Counts}(x) &= \mathbf{Counts}(x) - 1 \end{aligned}$$

```

function COUNTINGSORT(A, n)
    [...]                                     ▷ End of Pointing-Phase
     $A' = \text{New Array of size } n$ 

    for  $i = n \dots 1$  do
         $value = A[i]$ 
         $position = \mathbf{Counts}[value]$ 
         $A'[position] = value$ 
         $\mathbf{Counts}[value] -= 1$ 
    end for
    return Test
end function
```

## 2.5 Pseudocode(s)

**Input:**  $A = [a_1, a_2, \dots, a_n]$ ,  $n$  = Number of elements in  $A$

**Output:**  $A' = [a'_1, a'_2, \dots, a'_n]$  with  $\forall a'_i, a'_j; 1 \leq i < j \leq n : a'_i \leq a'_j$

**function** COUNTINGSORT( $A, n$ )

$k = \text{Maximum}(A)$

$Counts = \text{New Array of size } k$

$\triangleright$  Filled with zeroes

**for**  $i = 1 \dots n$  **do**

$value = A[i]$

$Counts[value] += 1$

**end for**

**for**  $i = 2 \dots k$  **do**

$Counts[i] += Counts[i - 1]$

**end for**

$A' = \text{New Array of size } n$

**for**  $i = n \dots 1$  **do**

$value = A[i]$

$position = Counts[value]$

$A'[position] = value$

$Counts[value] -= 1$

**end for**

**return**  $A'$

**end function**

### 3 Implementations

Python:

```
1 # Finding a maximum among n elements in any
2 # language should not be worth any commenting really
3 def find_maximum(a):
4     maximum = a[0]
5     counter = 0
6     for x in a:
7         if x > maximum:
8             maximum = x
9             counter += 1
10    return counter, maximum
11 # Same applies for finding the minimum
12 def find_minimum(a):
13     minimum = a[0]
14     counter = 0
15     for x in a:
16         if x < minimum:
17             minimum = x
18             counter += 1
19    return counter, minimum
20
21
22 def advanced_counting_sort(a, high_key=None):
23     # Phase 0: Normalizing for negative elements
24     num_elements, low_key = find_minimum(a)
25     for i in range(0, num_elements):
26         a[i] -= low_key
27
28     if not high_key:
29         _, high_key = find_maximum(a)
30
31     high_key -= low_key
32
33     # Phase 1: Counting occurrences of each key
34     counts = [0 for i in range(0, high_key + 1)]
35     for i in range(0, num_elements):
36         counts[a[i]] += 1
37
38     # Phase 2: Adjusting result array pointers
39     for i in range(1, high_key + 1):
40         counts[i] += counts[i - 1] # Counts-array becomes a
41         # pointer array
42
43     # Phase 3: Backwardly filling up the result
44     # array with the input elements
45     result = [None for x in a]
46     for i in range(num_elements - 1, -1, -1):
47         element = a[i]
48         a[i] += low_key # Reverting normalization
49         result_pointer = counts[element]
50         result[result_pointer - 1] = element + low_key
51         counts[element] -= 1
52
53     return num_elements, result
```



*This code represents an implementation in Python. It is called 'advanced Countingsort' because it can already deal with negative values. The basic concept however is the same since we only need to normalize the values of our input data to zero which is a linear projection that persists topology.*