

# Optimization of $O(n^2)$ search: Two Pointers Technic & Sliding Window

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# Summary

- 1 First approach: inverting a string
- 2 Sorted linear structures
  - 1 Pointers at the start: removing duplicates
    - 1 Naive solution in  $O(n^2)$
    - 2 Two pointers in  $O(n)$
  - 2 Pointers at both extremums: Targeted sums
- 3 Unsorted linear structures: Maximizing an area
- 4 Subset of static size: Largest  $k$ -subsequence
- 5 Subset of random size: Minimum subsequence
- 6 Conclusion

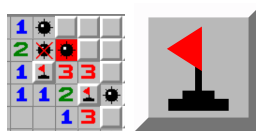


# What is a pointer ?

## Definition

Generally: A variable containing the memory adress of another variable.

Here: A numerical variable  $p$  containing the index of another variable  $S$  in a data structure  $D$ . In other words,  $D[p] == S$ . We say that  $p$  points to  $S$ , even though it's more of an iterator.



**Figure:** A pointer can be imagined like a red flag in the minesweeper game: it's a tool that helps keeping track of a special position.



# Reversing a string

## Problem A

Given a string  $S$  such that  $1 < |S| < 1000$ , write a code that returns its reversal.



# Reversing a string

## 1st approach

- 1 define an empty string T
- 2 add characters of S in T starting from its end
- 3 return T

## 2nd approach (Two pointers)

- 1 define 2 pointers  $i = 0$  and  $j = |S| - 1$
- 2 while  $i < j$ , swap  $S_i$  and  $S_j$  then increase  $i$  and decrease  $j$
- 3 return S

# Reversing a string

Example :  $S = \text{"ABCDEFGG"}$

G	B	C	D	E	F	A
↑↑ i						↑↑ j

Table: iteration 1

G	F	C	D	E	B	A
	↑↑ i				↑↑ j	

Table: iteration 2



# Reversing a string

```
def inverse(T):
    S = list(T)
    i = 0
    j = len(S) - 1
    while i < j:
        S[i], S[j] = S[j], S[i]
        i += 1
        j -= 1
    return X

s = input()
x = reverse(s)
for e in x:
    print(z, end=" ")
```



# Removing duplicates

## Problem B

Given a sorted list  $A$  of  $1 < n < 10^9$  sorted integers, write a code that returns the same list without duplicates.





# Removing duplicates

## Brute force

- 1 create an empty list  $M$
- 2 in 2 neested loops, for each element  $x$  in  $A$ , check if  $x \in M$ 
  - if  $x \in M$ , continue
  - else, append  $x$  to  $M$  and continue
- 3 return  $M$

time complexity :  $O(n^2)$



# Removing duplicates

## Two pointers technic

- 1 create an empty list  $M$  and initialize 2 pointers  $p = 0$  &  $q = 1$
- 2 in 1 loop, while  $q < n$ 
  - if  $A[p] \neq A[q]$ 
    - 1 append  $A[p]$  to  $M$
    - 2  $p = q$
  - $q++$
- 3 append  $A[q - 1]$  (or  $A[p]$ ) to  $M$

time complexity:  $O(n)$



# Removing duplicates

0	1	2	3	3	3	4	5	5	6
	↑↑	↑↑							
	p	q							

Table: A, (end of) iteration 1

0
---

Table: M, iteration 1



# Removing duplicates

0	1	2	3	3	3	4	5	5	6
		↑	↑						
		p	q						

Table: A, iteration 2

0	1
---	---

Table: M, iteration 2



# Removing duplicates

0	1	2	3	3	3	4	5	5	6
			↑	↑					
			p	q					

Table: A, iteration 3

0	1	2
---	---	---

Table: M, iteration 3



# Removing duplicates

0	1	2	3	3	3	4	5	5	6
			↑↑ p		↑↑ q				

Table: A, iteration 4

0	1	2	3	3	3	4	5	5	6
			↑↑ p			↑↑ q			

Table: A, iteration 5



# Removing duplicates

0	1	2	3	3	3	4	5	5	6
						↑↑	↑↑		
						p	q		

Table: A, iteration 6

0	1	2	3
---	---	---	---

Table: M, iteration 6



# Removing duplicates

0	1	2	3	3	3	4	5	5	6	IOR
									↑↑	↑
									p	q > 9

Table: A, last iteration

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

Table: M, last iteration





# Removing duplicates

```
def remove_duplicates(A):
    p = 0
    q = 1
    n = len(A)
    M = []
    while q < n:
        if A[p] != A[q]:
            M.append(A[p])
            p = q
        q += 1
    M.append(A[q - 1])
    return M

s = list(map(int, input().split()))
t = remove_duplicates(s)
for e in t: print(e, end=" ")
```



# Targeted sum

## Problem C

Given an array  $A$  of  $n$  sorted integers and an integer  $K > 1$ , find out a pair of integers which sum equals  $K$ . If there isn't such a pair in  $A$ , return -1.



# Targeted sum

## Brute force

Verify if  $A[i] + A[j] = K$ ,  $(0 \leq i < j \leq n - 1) \Rightarrow \frac{n(n-1)}{2}$  pairs to test. Stop when there is a satisfying pair. If all the pairs doesn't satisfy the condition, return -1.  
time complexity:  $O(n^2)$



# Targeted sum

## Two pointers technic

- 1 initialize 2 pointers  $i = 0$  and  $j = n-1$
- 2 in 1 loop, while  $i \neq j$ , evaluate  $A[i] + A[j]$ 
  - if  $A[i] + A[j] = K$ , return  $(A[i], A[j])$
  - if  $A[i] + A[j] > K$ ,  $j--$  (adding smaller number)
  - if  $A[i] + A[j] < K$ ,  $i++$  (adding bigger number)
- 3 return -1 (since no pair is satisfying)

time complexity:  $O(n)$



# Targeted sum

Example:  $K = 9$

1	2	4	7	11	12
↑↑					↑↑
p					q

Table:  $*i + *j = 13 > 9$

1	2	4	7	11	12
↑↑				↑↑	
p				q	

Table:  $*i + *j = 12 > 9$



# Targeted sum

Example:  $K = 9$

1	2	4	7	11	12
↑↑			↑↑		
p			q		

Table:  $*i + *j = 8 < 9$

1	2	4	7	11	12
	↑↑		↑↑		
	p		q		

Table:  $*i + *j = 9$



# Targeted sum

```
def findSum(A, k):
    i = 0
    n = len(A)
    j = n - 1
    while i != j:
        if A[i] + A[j] == k : return (A[i],A[j])
        elif A[i] + A[j] > k : j -= 1
        else : i += 1
    return -1

k = int(input())
A = list(map(int , input().split ()))
print (findSum(A,k))
```



# Important notice

If the order of the data isn't important in the problem, we can first sort it in  $O(n\log(n))$  then apply the two pointers technic in  $O(n)$ , which is  $O(n\log(n))$  overall. This is still better than  $O(n^2)$ .

Examples : problems B and C if the order wasn't given.





# Maximizing an area

## Problem D

Consider an array  $A$  of  $n$  integers representing the heights of  $n$  vertical sticks, in the same arrangement as in the array. Assuming that the distance between each stick is one unit, what is the maximum amount of water one could hold between two sticks (disregarding any sticks in between)?



# Maximizing an area

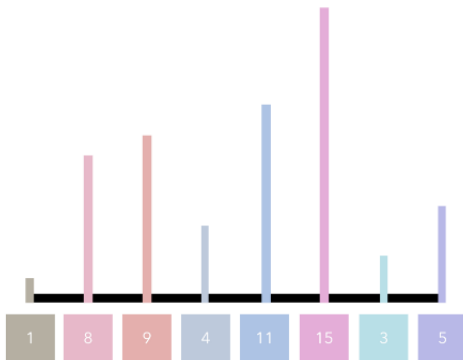


Figure: Illustration



# Maximizing an area

## Brute force

Calculate the legal area between each 2 sticks  $\Rightarrow \frac{n(n-1)}{2}$  area to evaluate, while keeping track of the maximum.

time complexity:  $O(n^2)$



# Maximizing an area

## Two pointers technic

- 1 initialize 2 pointers  $i=0$  and  $j=n-1$
- 2 initialize a variable  $max = 0$
- 3 while  $i < j$  :
  - 1 evaluate the area  $S$  between sticks  $A_i$  and  $A_j$
  - 2 if  $S > max$ , update  $max$
  - 3 if  $A_i \geq A_j$ ,  $j--$
  - 4 else,  $i++$
- 4 return  $max$

time complexity:  $O(n)$

# Maximizing an area

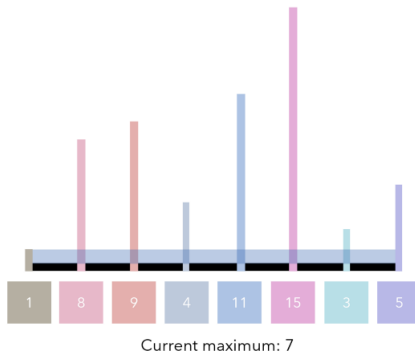


Figure: first position, all other situations can be obtained from it



# Maximizing an area

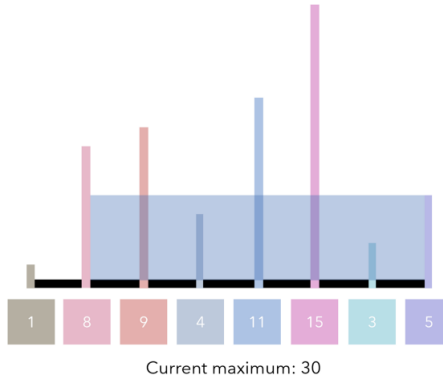


Figure: the best option is to change the shortest stick



# Maximizing an area

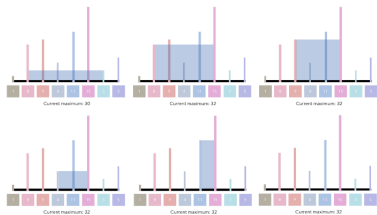
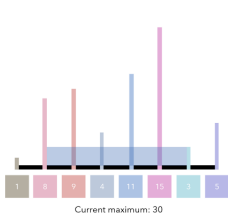


Figure: and so on..



# Maximizing an area

```
def maxArea(A):
    i = 0
    n = len(A)
    j = n - 1
    mx = 0
    while i < j:
        if A[i] + A[j] > mx : mx = A[i] + A[j]
        if A[i] >= A[j]:: j -= 1
        else : i += 1
    return mx
A = list(map(int,input().split()))
print(maxArea(A))
```





# Largest $k$ -subsequence

## Problem E

Given an array  $A$  of  $n \leq 10^9$  non-sorted integers and an integer  $k < n$ , return the largest sum of  $k$  consecutive integers.



# Largest $k$ -subsequence

## Brute force

in 2 nested loops, evaluate the sum of each  $k$  consecutive elements  
 $\Rightarrow (n - k)n$  operation to perform.  
 time complexity:  $O(n^2)$



# Largest $k$ -subsequence

## Sliding window algorithm

- 1 evaluate  $\sum_{i=0}^{k-1} A_i = \text{current}$
- 2 initialize  $\text{max} = \text{current}$
- 3 initialize 2 pointers  $i=0$  and  $j=k$
- 4 in 1 loop, while  $j < n$ 
  - update  $\text{current} = \text{current} + A_j - A_i$
  - if  $\text{current} > \text{max}$ , update  $\text{max} = \text{current}$
  - $j++$  and  $i++$
- 5 return  $\text{max}$



# Largest $k$ -subsequence



Figure: a real sliding window (*thanks to Mohamed Nasser!*)



# Largest $k$ -subsequence

Example :  $k=3$

11	22	14	17	11	19	15	13
↑ i			↑ j				

Table: window  $A[0:2]$ , current = 47, max = 47

11	22	14	17	11	19	15	13
	↑ i			↑ j			

Table: window  $A[1:3]$ , current = 53, max = 53



# Largest $k$ -subsequence

Example :  $k=3$

11	22	14	17	11	19	15	13
		↑ i			↑ j		

Table: window  $A[2:4]$ , current = 42, max = 53

11	22	14	17	11	19	15	13
			↑ i			↑ j	

Table: window  $A[3:5]$ , current = 47, max = 53



# Largest $k$ -subsequence

```
def maxSubarray(A, k):
    n = len(A)
    s = sum(A[0:k])
    i = 0
    j = k
    mx = s
    while j < n :
        s -= A[i]
        s += A[j]
        if s > mx : mx = s
        j += 1
        i += 1
    return mx

k = int(input())
A = list(map(int, input().split()))
print(maxSubarray(A, k))
```



# Minimum subsequence

## Problem F

Given a string  $S$  and a string  $T$ , with  $|T| < 26$  and  $|S| \leq 10^9$ , find the shortest substring\* of  $S$  that contains all the characters of  $T$ , otherwise return an empty string.

\*a substring is a contiguous subpart of a string.





# Minimum subsequence

## Brute force

iterate in 2 nested loops on all the subsets of the string  $S$  from the smallest to the biggest  $\Rightarrow \frac{n(n-1)}{2}$  substring to generate and test.  
time complexity:  $O(n^2)$



# Minimum subsequence

## Sliding window algorithm

to resume the idea, we can imagine it as an iterative process of 2 subprocesses :

- adding more characters from  $S$  until all the characters in  $T$  are there, and keeping track of their number of occurrence
- removing characters from the start as long as we still can found the characters of  $T$  in the remaining window

and we keep track of the shortest substring satisfying the condition during this process.

time complexity:  $O(n)$

# Minimum subsequence

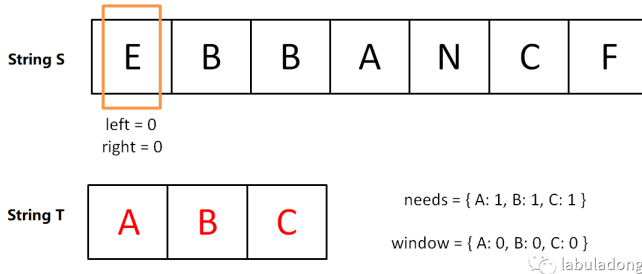
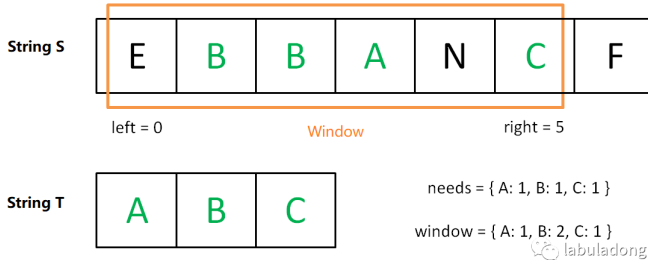


Figure: initialization



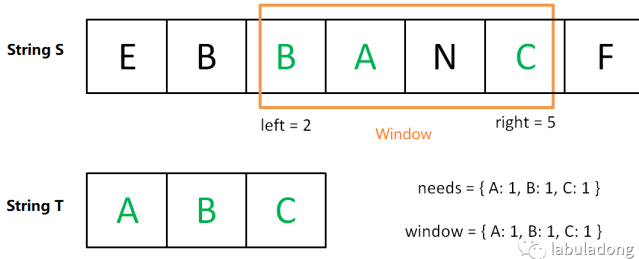
# Minimum subsequence



**Figure:** end of increasing of the right pointer. The current window is valid, not optimal.



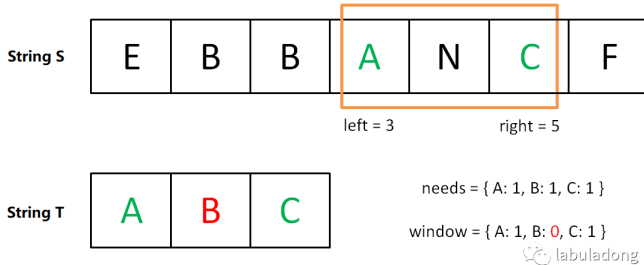
# Minimum subsequence



**Figure:** increasing of left pointer. The current window is still valid but more optimal.



# Minimum subsequence



**Figure:** increasing of left pointer. The current window is not valid. We return to increase the right pointer. The final result is "BANC".



# Minimum subsequence

- 1 initialize a hash table *need* (map/dictionary) with the characters of  $T$  as keys and their occurrences as values
- 2 initialize an empty hash table *window*
- 3 initialize 2 pointers  $i=0$  and  $j=0$
- 4 initialize 3 variables  $min = n+1$ ,  $match = 0$  and  $k = \text{len}(need)$  (number of distinct elements of  $T$ )
- 5 initialize a variable  $start = 0$



# Minimum subsequence

```

6 while j < n :
    1 if need[S[j]] != 0 then
        1 window[S[j]]++
        2 if need[S[j]]==window[S[j]] then match++
    2 j++
    3 while match == k :
        1 if j - i < min then update start = i and min = j - i
        2 if need[S[i]] != 0 then
            1 window[S[i]]--
            2 if need[S[i]]>window[S[i]] then match--
        3 i++
7 if min > n then return "", else return S[start:start+min]
    
```





# Minimum subsequence

```
def minWindow(S, T):
    i = 0
    j = 0
    n = len(S)
    start = 0
    minimum = len(S) + 1
    match = 0
    need = dict()
    for e in T:
        if need.get(e, "-1") == "-1" : need[e] = 0
        need[e] += 1
    k = len(need)
    window = dict()
```



# Minimum subsequence

```

while j < n :
    if need.get(S[j], "-1") != "-1" :
        if window.get(S[j], "-1") == "-1" : window[S[j]] = 0
        window[S[j]] += 1
        if window[S[j]] == need[S[j]] : match += 1
    j += 1
    while match == k:
        if j - i < minimum :
            start = i
            minimum = j - i
        if need.get(S[i], "-1") != "-1" :
            window[S[i]] -= 1
            if window[S[i]] < need[S[i]] : match -= 1
        i += 1
    if minimum > n : return ""
    return S[start:start+minimum]

```

```

s = input()
t = input()
print(minWindow(s, t))

```



# Conslusion

The Two Pointers Technic and the Sliding Window Algorithm are two powerful tools to deal with search in different types of linear data structures and that can help optimizing it from a polynomial or exponential time to a linear or linear-logarithmic time. This could be very important when the input size passes  $10^4$ .

The only differnce between this 2 algorithms, is that in the two pointers we only consider **the 2 elements pointed** by our pointers, but in the sliding window we consider also **the elements between them**.



## Further reading

Amortized Analysis repository on GitHub:  
[github.com/HamzaBamohammed/amortized-analysis](https://github.com/HamzaBamohammed/amortized-analysis)



# Webography

1. "Two-pointer technique". *Leetcode*. Visited in 21/02/2022.
2. Andre Ye, "Two-Pointer Technique: Solving Array Problems at Light Speed". *Medium*. Visited in 21/02/2022.
3. labuladong, "Sliding Window Algorithm". *Github*. Visited in 21/02/2022.
4. Antti Laaksonen, Competitive Programmer's Handbook.

