

A warm up!

Chapter 2 of the notes

# Maximum Sub-array Sum

Problem: given an array of  $n$  integers (positive and negative) find the sub-array of maximum sum.

Input: array  $A[1,n]$  of positive and negative integers

Output:  $l, r$  where  $A'[l, r]$  is the sub-array

Output: The sum value

Example:

Input    -1 5 8 -9 1 1

Output    13

# Solution 1

```
max=a[0];
for(i=0; i<n; i++)
{
    for(j=i; j<n; j++)
    {
        sum=0;
        for(k=i; k<=j; k++)
        {
            sum+=a[k];
        }
        if (sum> max) max=sum;
    }
}
```

```
max=a[0];
for(i=0; i<n; i++)
{
    for(j=i; j<n; j++)
    {
        sum=0;
        for(k=i; k<=j; k++)
        {
            sum+=a[k];
        }
        if (sum > max) max=sum;
    }
}
```

Input    -1 5 8 -9 1 1

          i    j

# Solution 1

```
max=a[0];
for(i=0; i<n; i++)
{
    for(j=i; j<n; j++)
    {
        sum=0;
        for(k=i; k<=j; k++)
        {
            sum+=a[k];
        }
        if (sum > max) max=sum;
    }
}
```

Time complexity  $O(n^3)$

# Lower Bound for Solution 1

Instruction `somma+=a[k];`  
(on sub-arrays of length  $j-i+1$ ) is executed

$$\sum_{i=0, n-1} \sum_{j=i, n-1} (j-i+1) \text{ times, or}$$

$$\sum_{i=0, n-1} \sum_{j=1, n-i} j \geq \sum_{i=0, n-1} (n-i+1)^2/2 = \sum_{j=1, n+1} j^2/2$$

$$\Omega(n^3)$$

## Solution 2

```
max=a[0];
for(i=0; i<n; i++)
{
    sum=0;
    for(j=i; j<n; j++)
    {
        sum+=a[j];
        if(sum > max) max=sum;
    }
}
```

Input -1 5 8 -9 1 1

*i*                      *j*

# Solution2

```
max=a[0];
for(i=0; i<n; i++)
{
    sum=0;
    for(j=i; j<n; j++)
    {
        sum+=a[j];
        if(sum > max) max=sum;
    }
}
```

Time complexity:  $O(n^2)$



# How to do better

Observe 2 properties of the maximum sum sub-array:

- 1) The sum of the values in each prefix of the **maximum sum sub-array** is positive, otherwise we could remove this prefix obtaining a sub-array with greater sum (**contradiction**).
- 2) The value of the element previous than the first element of the **maximum sum sub-array** is negative, otherwise could be added to the sub-array obtaining a sub-array with greater sum (**contradiction**).

-3 2 3 -1 8 1

# Solution 3

```
max = A[0]; sum = 0;
```

```
for(i=0; i<n; i++)
```

```
{
```

```
  if(sum > 0) sum+=a[i];    extend the segment
```

```
    else sum=a[i];        start a new segment
```

```
  if(sum > max) max=sum;
```

```
}
```

Time complexity:  $O(n)$

Optimal Algorithm ! Why?

# Example linear solution

	SUM	MAX
1 2 -4 1 3 2 -2 1	1	1
1 2 -4 1 3 2 -2 1	3	3
1 2 -4 1 3 2 -2 1	-1	3
1 2 -4 1 3 2 -2 1	1	3
1 2 -4 1 3 2 -2 1	4	4
1 2 -4 1 3 2 -2 1	6	6
1 2 -4 1 3 2 -2 1	4	6
1 2 -4 1 3 2 -2 1	5	6

# 2-level memory model

- $B$  = block (page) size
- $M$  = internal memory size

How to evaluate the complexity of an algorithm?

number of I/Os operations

- In this model Solution3 takes  $n/B$  I/Os operations is **optimal**.
- It is independent from the block size. Very important feature for an algorithm. **Cache-oblivious**.

# Another linear time algorithm

- Let  $\text{Sum}_D(x,s)$  the sum of items in positions from  $x$  to  $s$

$$\text{Sum}_D(x,s) = \text{Sum}_D(1,s) - \text{Sum}_D(1,x-1)$$

prefix sums  
until  $s$

prefix sums  
until  $x-1$

$O(n)$

New algorithm:

$$\max_s \left( \max_{b \leq s} \text{Sum}_D(b,s) \right)$$

$$\max_s \left( \max_{b \leq s} \text{Sum}_D(1,s) - \text{Sum}_D(1,b-1) \right)$$

# Another linear time algorithm

Find the positions  $\langle l, r \rangle$  of the subarray.  
 Compute prefix sums of D in array P.

D	4	-6	3	1	3	-2	3	-4	1	-9	6
P	4	-2	1	2	5	3	6	2	3	-6	0

Note that  $P[i] = P[i-1] + D[i]$     pose  $P[0] = 0$

Write the sum in terms of P:

$$\max_s \left( \max_{b \leq s} \text{Sum}_D(1,s) - \text{Sum}_D(1,b-1) \right)$$

$$\max_s \left( \max_{b \leq s} (P[s] - P[b-1]) \right)$$

Decompose max computation in **max-min** computation:

$$\max_s \left( \max_{b \leq s} (P[s] - P[b-1]) \right) = \max_s \left( (P[s] - \min_{b \leq s} P[b-1]) \right)$$

independent from b

can be precomputed in M

# Another linear time algorithm

D	4	-6	3	1	3	-2	3	-4	1	-9	6
---	---	----	---	---	---	----	---	----	---	----	---

P	4	-2	1	2	5	3	6	2	3	-6	0
---	---	----	---	---	---	---	---	---	---	----	---

$$M[i] = \min (M[i-1], P[i]) \quad M[0] = 0 \quad M[0, 10]$$

$$\min_{b \leq s} P [b-1]$$

M	0	0	-2	-2	-2	-2	-2	-2	-2	-2	-6
---	---	---	----	----	----	----	----	----	----	----	----

$$P[s] - M[s-1]$$

P'	4	-2	3	4	7	5	8	4	5	-4	6
----	---	----	---	---	---	---	---	---	---	----	---

Max value is 8 for  $s=7$ ,

Left extreme is computed as position where  $P'$  is min,  $l=2$

Return  $\langle 2, 7 \rangle$

# Another linear time algorithm

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## Algorithm 2.4 Another linear-time algorithm

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```
1:  $MaxSum = -\infty; b_o = 1;$ 
2:  $TmpSum = 0; MinTmpSum = 0;$ 
3: for ( $s = 1; s \leq n; s++$ ) do
4:    $TmpSum += D[s];$ 
5:   if ( $MaxSum < TmpSum - MinTmpSum$ ) then
6:      $MaxSum = TmpSum; s_o = s;$ 
7:   end if
8:   if ( $TmpSum < MinTmpSum$ ) then
9:      $MinTmpSum = TmpSum; b_o = s + 1;$ 
10:  end if
11: end for
12: return  $\langle MaxSum, b_o, s_o \rangle;$ 
```

---

The discussed algorithm takes three scans over  $D$ , in fact can be organized in a single pass and no memory.

Keep the values of  $P[s]$  or  $M[s-1]$  (max and min) in 2 variables  $Tmpsum$  and  $MinTmpSum$  scanning the array and compute formula  $P[s] - M[s-1]$  incrementally.



# Interesting variants with application to the Bioinformatics

Sub-array  Segment

Maximum-sum segment problem

DNA sequence is a string on 4 letters (*A, T, C, G*)

Problem: Identify segments reach of *C* and *G* nucleotides (biologically significant).

Is it possible to exploit our algorithm?

Input: from DNA sequences to arrays of numbers.

# Two ways

1. Assign penalty  $-p$  to A and T and a reward  $1-p$  to C and G. In this way the sum of the values of a segment of length  $l$  containing  $x$  C or G is  $x-pl$ .

$$x(1-p) - (l-x)p = x - xp - xl + xp = x - pl$$

Our linear algorithm can be used to solve this problem with this objective function.

Often biologist prefer to put limits on the length of the segments, to avoid extremely short or long segments. Now the algorithm cannot be applied!!!

Another linear solution, however, is possible.

# Small changes in the problem:: Big Jump in the complexity

**The trap is:** no limits to the segment length implies only trivial solutions of length 1.

Circumvent the single output searching:

**Problem:** Given an array  $D[1,n]$  of positive and negative integers, find the longest segment in  $D$  whose sum is largest of a fixed threshold  $t$ .

**Complement** of the previous one: Maximize the sum provided that its length is within a given range.

The structure of the algorithmic solution is the same!

The two problem can be reduced one into the other

$$\frac{\text{Sum}_D(x,y)}{y-x+1} = \sum_{k=x,y} \frac{D[k]}{y-x+1} \geq t \iff \sum_{k=x,y} (D[k] - t) \geq 0$$

Subtracting  $t$  to all elements in  $D$ , the problem of bounded density becomes equivalent to find the **longest segment with sum larger or equal to 0**.

A sum based problem is equivalent to a density based problem!

Reduction is very important technique to reuse solutions.

**Go back to the problem:**

Given an array  $D[1,n]$  of positive and negative integers, find the longest segment in  $D$  whose sum is larger than a fixed threshold  $t$ .

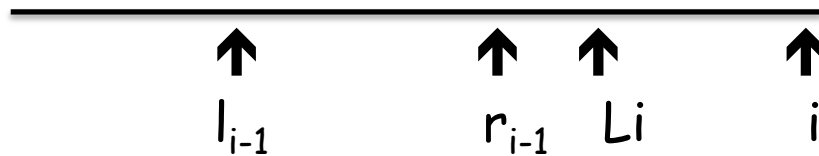
# Algorithm

Inductively let  $D[1, i-1]$  **the longest segment with sum  $\geq t$**  already computed for  $i=1, \dots, n$  with solution  $D[l_{i-1}, r_{i-1}]$ .

Compute  $D[1, i]$ : 2 possibilities

1.  $r_i < i$  the solution does not change
2.  $r_i = i$   $D[l_i, r_i]$  is longer than  $D[l_{i-1}, r_{i-1}]$ .

**Observe:**  $l_i$  occurs to the left of position  $L_i = i - (r_{i-1} - l_{i-1})$



We can discard for  $l_i$  all positions between  $L_i$  and  $i$  since they generate solutions shorter than  $D[l_{i-1}, r_{i-1}]$ .

**Reformulated problem.** Given  $D[1, n]$  of positive and negative numbers we want to find at every step the smallest index  $i$  such that

$$\text{Sum}_D [l_i, i] \geq t.$$

# Algorithm

Recall to compute the sum compute the prefix sums as before so that  $\text{Sum}_D[1, i] - \text{Sum}_D[1, li-1] = P[i] - P[li-1]$ .

We look for the smallest index  $li$  in  $[1, Li]$  such that  $P[i] - P[li-1] \geq t$ .  
Array  $P$  pre-computed in linear time and space.

But we have to find a minimum for all values of  $i$  that means  $O(n^2)$ .

Instead, we identify a set of candidate positions for iteration:  $C_{i,j}$  is **the leftmost minimum** of the sub-array  $P[1, C_{i,j} - 1]$ .  $C_{i0} = Li$ .

$P$   $Li$



$C_{i3}$

$C_{i2}$

$C_{i1}$

$C_{i0}$

Properties:  $C_{i3} < C_{i2} < C_{i1} < C_{i0} = Li$

$P[C_{i3}] > P[C_{i2}] > P[C_{i1}]$

$P[C_{ij}]$  is the leftmost minimum of prefix  $P[1, C_{ij} - 1]$  hence is smaller than any other values on its left

# Algorithm

**Fact:** At each iteration  $i$ , the largest index  $j^*$  such that  $\text{Sum}_D [C_{i,j^*+1}, i] \geq t$  if any is the longest segment we are searching for.

**Proof:** in the lecture notes.

The computation of all candidate positions takes  $O(n)$  time. See also lecture notes.

The computation of  $j^*$  is not constant, but if at each iteration  $i$  we go on of  $s_i$  steps, we also go on in the construction of a longest solution of  $s_i$  steps. We extend the solution of  $\theta(s_i)$ . The sum of all these extra costs cannot be larger than  $O(n)$ .