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Introduction

Mo's algorithm is a generic idea. It applies to the following class of problems:

You are given array **Arr** of length **N** and **Q** queries. Each query is represented by two numbers **L** and **R**, and it asks you to compute some function **Func** with subarray **Arr[L..R]** as its argument.

For the sake of brevity we will denote $\text{Func}([L, R])$ as the value of **Func** on subarray **Arr[L..R]**.

If this sounds too abstract, let's look at specific example:

There is an integer array **Arr** of length **N** and **Q** queries. For each **i**, query **#i** asks you to output the sum of numbers on subarray $[L_i, R_i]$, i.e. $\text{Arr}[L_i] + \text{Arr}[L_i + 1] + \dots + \text{Arr}[R_i]$.

Here we have $\text{Func}([L, R]) = \text{Arr}[L] + \text{Arr}[L + 1] + \dots + \text{Arr}[R]$.

This does not sound so scary, does it? You've probably heard of solutions to this problem using Segment Trees or Binary Indexed Trees, or even prefix sums.

Mo's algorithm provides a way to **answer all queries in $O((N + Q) * \sqrt{N} * F)$ time** with at least $O(Q)$ additional memory. Meaning of **F** is explained below.

The algorithm is applicable if all following conditions are met:

1. **Arr** is not changed by queries;
2. All queries are known beforehand (techniques requiring this property are often called "offline algorithms");
3. If we know $\text{Func}([L, R])$, then we can compute $\text{Func}([L + 1, R])$, $\text{Func}([L - 1, R])$, $\text{Func}([L, R + 1])$ and $\text{Func}([L, R - 1])$, each in **$O(F)$** time.

Due to constraints on array immutability and queries being known, Mo's algorithm is inapplicable most of the time. Thus, knowing the method can help you on rare occasions. But. Due to property #3, the algorithm can solve problems that are unsolvable otherwise. If the problem was meant to be solved using Mo's algorithm,

then you can be 90% sure that it can not be accepted without knowing it. Since the approach is not well-known, in situations where technique is appropriate, you will easily overcome majority of competitors.

Basic overview

We have Q queries to answer. Suppose we answer them in order they are asked in the following manner:

```
for i = 0..Q-1:
    L, R = query #i
    for j = L..R:
        do some work to compute Func([L, R])
```

This can take $\Omega(N * Q)$ time. If N and Q are of order 10^5 , then this would lead to time limit exceeded.

But what if we answer queries in different order? Can we do better then?

Definition #1:

Segment $[L, R]$ is a continuous subarray $Arr[L..R]$, i.e. array formed by elements $Arr[L]$, $Arr[L + 1]$, ..., $Arr[R]$. We call L **left endpoint** and R **right endpoint** of segment $[L, R]$. We say that index i belongs to segment $[L, R]$ if $L \leq i \leq R$.

Notation

1. Throughout this tutorial " x/y " will mean "integer part of x divided by y ". For instance, $10/4 = 2$, $15/3 = 5$, $27/8 = 3$;
2. By " $\text{sqrt}(x)$ " we will mean "largest integer less or equal to square root of x ". For example, $\text{sqrt}(16) = 4$, $\text{sqrt}(39) = 6$;
3. Suppose a query asks to calculate $\text{Func}([L, R])$. We will denote this query as $[L, R]$ - the same way as the respective argument to Func ;
4. Everything is 0-indexed.

We will describe Mo's algorithm, and then prove its running time.

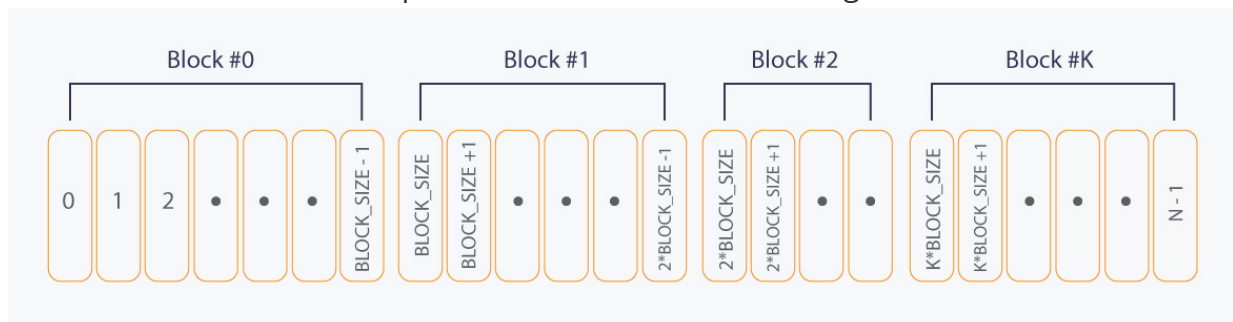
The approach works as follows:

1. Denote $\text{BLOCK_SIZE} = \text{sqrt}(N)$;
2. Rearrange all queries in a way we will call "**Mo's order**". It is defined like this: $[L_1, R_1]$ comes earlier than $[L_2, R_2]$ in Mo's order if and only if:
 - a) $L_1 / \text{BLOCK_SIZE} < L_2 / \text{BLOCK_SIZE}$
 - b) $L_1 / \text{BLOCK_SIZE} == L_2 / \text{BLOCK_SIZE} \ \&\& \ R_1 < R_2$

3. Maintain segment $[mo_left, mo_right]$ for which we know $Func([mo_left, mo_right])$. Initially, this segment is empty. We set $mo_left = 0$ and $mo_right = -1$;
 4. Answer all queries following Mo's order. Suppose the next query you want to answer is $[L, R]$. Then you perform these steps:
 - a) while mo_right is less than R , extend current segment to $[mo_left, mo_right + 1]$;
 - b) while mo_right is greater than R , cut current segment to $[mo_left, mo_right - 1]$;
 - c) while mo_left is greater than L , extend current segment to $[mo_left - 1, mo_right]$;
 - d) while mo_left is less than L , cut current segment to $[mo_left + 1, mo_right]$.
- This will take $O((|left - L| + |right - R|) * F)$ time, because we required that each extension\deletion is performed in $O(F)$ steps. After all transitions, you will have $mo_left = L$ and $mo_right = R$, which means that you have successfully computed $Func([L, R])$.

Time complexity

Let's view Arr as a union of disjoint segments of size $BLOCK_SIZE$, which we will call "blocks". Take a look at the picture for better understanding:



Let K be the index of last block. Then there are $K + 1$ blocks, because we number them from zero. Notice that K 'th block can have less than $BLOCK_SIZE$ elements.

Proposition #1:

$$K = O(\sqrt{N}).$$

Proof:

If $\sqrt{N} * \sqrt{N} = N$ (which may be false due to our definition of \sqrt{N}), then $K = \sqrt{N} - 1$, because we have $K + 1$ blocks, each of size \sqrt{N} . Otherwise, $K = \sqrt{N}$, because we need one additional block to store $N - \sqrt{N} * \sqrt{N}$ elements.

Proposition #2:

Block $\#r$ is a segment $[r * BLOCK_SIZE, \min(N - 1, (r + 1) * BLOCK_SIZE - 1)]$.

Proof (by induction):

For block $\#0$ statement is true — it is a segment $[0, BLOCK_SIZE - 1]$, containing $BLOCK_SIZE$ elements.

Suppose first $T \leq K$ blocks satisfy the above statement. Then the last of those blocks is a segment $[(T - 1) * \text{BLOCK_SIZE}, T * \text{BLOCK_SIZE} - 1]$.

Then we form the next, $T+1$ 'th block. First element of this block will have array index $T * \text{BLOCK_SIZE}$. We have at most BLOCK_SIZE elements to add to block (there may be less if $T + 1 = K + 1$). So the last index in $T+1$ 'th block will be $\min(N - 1, (T + 1) * \text{BLOCK_SIZE} - 1)$.

Corollary #1: Two indices i and j belong to same block $\#r$ if and only if

$$\lfloor i / \text{BLOCK_SIZE} \rfloor = \lfloor j / \text{BLOCK_SIZE} \rfloor = r.$$

Definition #2:

$Q_r = \{ \text{query } [L, R] \mid \lfloor L / \text{BLOCK_SIZE} \rfloor = r \}$. Informally, Q_r is a set of queries from input, whose left endpoints belong to block $\#r$. Notice that this set may be empty. We denote the size of Q_r as $|Q_r|$.

Proposition #3:

For each r , queries from Q_r lie continuously in Mo's order and they appear in it in non-decreasing order of right endpoints.

Queries from Q_r come earlier than queries from Q_{r+1} for every $r = 0..K-1$.

Proof follows from definition of Mo's order.

Corollary #2: When we are processing queries following Mo's order, we firstly process all queries from Q_0 , then all queries from Q_1 , and so on, up to Q_K .

Theorem #1:

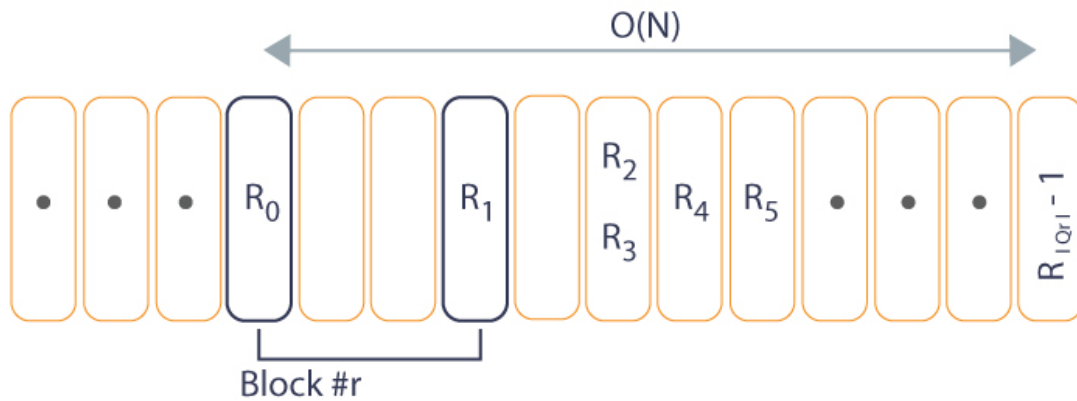
mo_right changes its value $O(N * \sqrt{N})$ times throughout the run of Mo's algorithm.

Proof:

1. Suppose we've just started processing queries from Q_r for some r , and we've already answered first (following Mo's order) query from it. Let that query be $[L, R_0]$. This means that now $\text{mo_right} = R_0$.

Let $R_0, R_1, \dots, R_{|Q_r| - 1}$ be right endpoints of queries from Q_r , in order they appear in Mo's order. From proposition #3 we know that

$$R_0 \leq R_1 \leq R_2 \leq \dots \leq R_{|Q_r| - 1}$$



From proposition #2 and definition of Q_r we know that

$$r * \text{BLOCK_SIZE} \leq L$$

Since right endpoint is not lower than left endpoint (i.e. $L \leq R$), we conclude that

$$r * \text{BLOCK_SIZE} \leq R_0$$

We have N elements in Arr , so any query has right endpoint less than N .

Therefore,

$$R_{|Q_r|-1} - 1 \leq N - 1$$

Since R 's are not decreasing, the total amount of times mo_right changes is

$$R_{|Q_r|-1} - 1 - R_0 \leq N - 1 - r * \text{BLOCK_SIZE} = O(N)$$

(we've substituted $R_{|Q_r|-1}$ with its highest possible value, and R_0 with its lowest possible value to maximize the subtraction).

There are $O(\sqrt{N})$ different values of r (proposition #1), so in total we will have $O(N * \sqrt{N})$ changes, assuming that we've already started from the first query of each Q_r .

- Now suppose we've just ended processing queries from Q_r and we must process first query from Q_{r+1} (assuming that it is not empty. If it is, then choose next non-empty set). Currently, mo_right is constrained to be:

$$r * \text{BLOCK_SIZE} \leq \text{mo_right} \leq N - 1$$

Let the first query from Q_{r+1} be $[L', R']$. We know (similarly to previous paragraph) that

$$(r + 1) * \text{BLOCK_SIZE} \leq R' \leq N - 1$$

Hence, mo_right must be changed at most

$$\max(|r * \text{BLOCK_SIZE} - (N - 1)|, N - 1 - (r + 1) * \text{BLOCK_SIZE})$$

times (we took lowest value of mo_right with highest value of R' and vice-versa). This is clearly $O(N)$ (and it does not matter whether it is $r+1$ 'th set of $r+k$ 'th for some $k > 1$).

There are $O(\sqrt{N})$ switches from r to $r + 1$ (and it's true even if we skip some empty sets), so in total we will have $O(N * \sqrt{N})$ mo_right changes to do this.

There are no more cases when mo_right changes. Overall, we have $O(N * \sqrt{N}) + O(N * \sqrt{N}) = O(N * \sqrt{N})$ changes of mo_right .

Corollary #3: All `mo_right` changes combined take $O(N * \sqrt{N} * F)$ time (because each change is done in $O(F)$ time).

Theorem #2:

`Mo_left` changes its value $O(Q * \sqrt{N})$ times throughout the run of Mo's algorithm.

Proof:

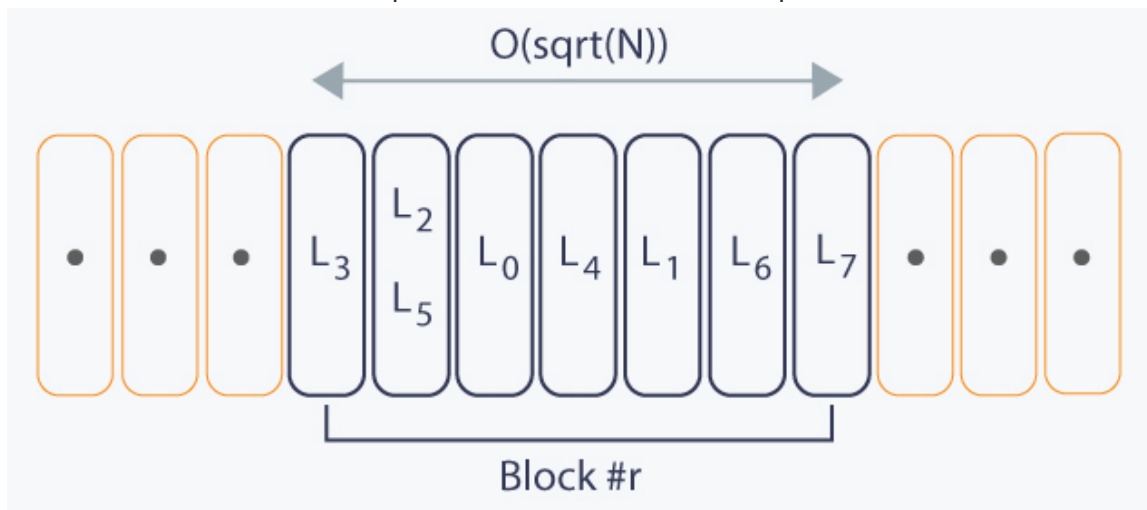
1. Suppose, as in the proof of Theorem #1, we've just started processing queries from Q_r for some r , and currently `mo_left` = L , `mo_right` = R_0 , where $[L, R_0] \in Q_r$. For every query $[L', R'] \in Q_r$ by definition #2 and proposition #2 we have:

$$r * \text{BLOCK_SIZE} \leq L' \leq (r + 1) * \text{BLOCK_SIZE} - 1$$

So, when we change `mo_left` from one query to other (both queries $\in Q_r$), we must do at most

$$(r + 1) * \text{BLOCK_SIZE} - 1 - r * \text{BLOCK_SIZE} = \text{BLOCK_SIZE} - 1 = O(\sqrt{N})$$

changes to `mo_left`. At the picture below we represented leftpoints that lie in block # r , where the subscript shows relative order of queries:



There are $|Q_r|$ queries in Q_r . That means, that we can estimate upper bound on number of changes for single r as $O(|Q_r| * N)$. Let's sum it over all r :

$$O(N * (|Q_0| + |Q_1| + \dots + |Q_K|)) = O(\sqrt{N} * Q)$$

(because each query's leftpoint belongs to exactly one block, and we have Q queries in total).

2. Now suppose we're done with queries from Q_r and want to process queries from non-empty set Q_{r+k} (for some $k > 0$). Any query $[L', R'] \in Q_{r+k}$ has

$$(r + k) * \text{BLOCK_SIZE} \leq L' \leq (r + k + 1) * \text{BLOCK_SIZE} - 1$$

Similarly, any query $[L, R] \in Q_r$ has

$$r * \text{BLOCK_SIZE} \leq L \leq (r + 1) * \text{BLOCK_SIZE} - 1$$

We can see that the maximum number of changes needed to transit from some query from Q_r to some query from Q_{r+k} is

$$(r + k + 1) * \text{BLOCK_SIZE} - 1 - r * \text{BLOCK_SIZE} = k * \text{BLOCK_SIZE} - 1$$

Now if we sum this over all r , we will get at most

$K * \text{BLOCK_SIZE} = O(\sqrt{N}) * O(\sqrt{N}) = O(N)$,
because sum of all possible k 's does not exceed K .

There are no more cases when `mo_left` changes. Overall, we have $O(\sqrt{N} * Q) + O(N) = O(\sqrt{N} * Q)$ changes of `mo_left`.

Corollary #4: All `mo_left` changes combined take $O(Q * \sqrt{N} * F)$ time (because each change is done in $O(F)$ time).

Corollary #5: Time complexity of Mo's algorithm is $O((N + Q) * \sqrt{N} * F)$.

Example problem

Let's look at an example problem (idea taken from [here](#)):

You have an array `Arr` of N numbers ranging from 0 to 99. Also you have Q queries $[L, R]$. For each such query you must print

$$V([L, R]) = \sum_{i=0..99} \text{count}(i)^2 * i$$

where `count(i)` is the number of times i occurs in `Arr[L..R]`.

Constraints are $N \leq 10^5$, $Q \leq 10^5$.

To apply Mo's algorithm, you must ensure of three properties:

1. `Arr` is not modified by queries;
2. Queries are known beforehand;
3. If you know $V([L, R])$, then you can compute $V([L + 1, R])$, $V([L - 1, R])$, $V([L, R - 1])$ and $V([L, R + 1])$, each in $O(F)$ time.

First two properties obviously hold. The third property depends on the time bound — $O(F)$.

Surely, we can compute $V([L + 1, R])$ from scratch in $\Omega(R - L) = \Omega(N)$ in the worst case. But looking at complexity of Mo's algorithm, you can deduce that this will surely time out, because we multiply $O(F)$ with $O((Q + N) * \sqrt{N})$.

Typically, you want $O(F)$ to be $O(1)$ or $O(\log(n))$. **Choosing a way to achieve appropriate time bound $O(F)$ is programmer's main concern when solving problems using Mo's algorithm.**

I will describe a way to achieve $O(F) = O(1)$ for this problem.

Let's maintain variable `current_answer` (initialized by zero) to store $V([\text{mo_left}, \text{mo_right}])$ and integer array `cnt` of size 100, where `cnt[x]` will be the number of times x occurs in $[\text{mo_left}, \text{mo_right}]$.

From the definition of `current_answer` we can see that

$$\text{current_answer} = V([\text{mo_left}, \text{mo_right}]) = \sum_{i=0..99} \text{cnt}'(i)^2 * i$$

where $\text{count}(x)$ is the number of times x occurs in $[\text{mo_left}, \text{mo_right}]$.

By the definition of cnt we see that $\text{count}(x) = \text{cnt}[x]$, so

$$\text{current_answer} = \sum_{i=0..99} \text{cnt}[i]^2 * i.$$

Suppose we want to change mo_right to $\text{mo_right} + 1$. It means we want to add number $p = \text{Arr}[\text{mo_right} + 1]$ into consideration. But we also want to retain current_answer and cnt 's properties.

Maintaining cnt is easy: just increase $\text{cnt}[p]$ by 1. It is a bit trickier to deal with current_answer .

We know (from the definitions) that current_answer contains summand $\text{cnt}[p]^2 * p$ before addition. Let's subtract this value from the answer. Then, after we perform addition to cnt , add again $\text{cnt}[p]^2 * p$ to the answer (make no mistake: this time it will contain updated value of $\text{cnt}[p]$). Both updates take $O(1)$ time.

All other transitions (mo_left to $\text{mo_left} + 1$, mo_left to $\text{mo_left} - 1$ and mo_right to $\text{mo_right} - 1$) can be done in the same way, so we have $O(F) = O(1)$. You can refer to the code below for clarity.

Now let's look into detail on one sample test case:

Input:

$\text{Arr} = [0, 1, 1, 0, 2, 3, 4, 1, 3, 5, 1, 5, 3, 5, 4, 0, 2, 2]$ of 18 elements

Queries (0-indexed): $[0, 8], [2, 5], [2, 11], [16, 17], [13, 14], [1, 17], [17, 17]$

The algorithm works as follows:

Firstly, set $\text{BLOCK_SIZE} = \text{sqrt}(18) = 4$. Notice that we have 5 blocks: $[0, 3], [4, 7], [8, 11], [12, 15], [16, 17]$. The last block contains less than BLOCK_SIZE elements.

Then, set $\text{mo_left} = 0$, $\text{mo_right} = -1$, $\text{current_answer} = 0$, $\text{cnt} = [0, 0, 0, 0, 0, 0]$ (I will use only first 6 elements out of 100 for the sake of simplicity).

Then sort queries. The Mo's order will be:

$[2, 5], [0, 8], [2, 11], [1, 17]$ (here ends Q_0) $[13, 14]$ (here ends Q_3) $[16, 17], [17, 17]$ (here ends Q_4).

Now, when everything is set up, we can answer queries:

1. We need to process query $[2, 5]$. Currently, our segment is $[0, -1]$. So we need to move mo_right to 5 and mo_left to 2.

Let's move mo_right first:

$\text{mo_right} = 0$, $\text{current_answer} = 0$, $\text{cnt} = [1, 0, 0, 0, 0, 0]$

$\text{mo_right} = 1$, $\text{current_answer} = 1$, $\text{cnt} = [1, 1, 0, 0, 0, 0]$

$\text{mo_right} = 2$, $\text{current_answer} = 4$, $\text{cnt} = [1, 2, 0, 0, 0, 0]$

$\text{mo_right} = 3$, $\text{current_answer} = 4$, $\text{cnt} = [2, 2, 0, 0, 0, 0]$

$\text{mo_right} = 4$, $\text{current_answer} = 6$, $\text{cnt} = [2, 2, 1, 0, 0, 0]$

$\text{mo_right} = 5$, $\text{current_answer} = 9$, $\text{cnt} = [2, 2, 1, 1, 0, 0]$

Now we must move `mo_left`:

`mo_left = 1, current_answer = 9, cnt = [1, 2, 1, 1, 0, 0]`

`mo_left = 2, current_answer = 6, cnt = [1, 1, 1, 1, 0, 0]`

Thus, the answer for query [2, 5] is 6.

2. Our next query is [0, 8]. Current segment [`mo_left`, `mo_right`] is [2, 5]. We need to move `mo_right` to 8 and `mo_left` to 0.

Again, let's move `mo_right` first:

`mo_right = 6, current_answer = 10, cnt = [1, 1, 1, 1, 1, 0]`

`mo_right = 7, current_answer = 13, cnt = [1, 2, 1, 1, 1, 0]`

`mo_right = 8, current_answer = 22, cnt = [1, 2, 1, 2, 1, 0]`

Then we move `mo_left`:

`mo_left = 1, current_answer = 27, cnt = [1, 3, 1, 2, 1, 0]`

`mo_left = 0, current_answer = 27, cnt = [2, 3, 1, 2, 1, 0]`

So, the answer for query [0, 8] is 27.

3. Next query is [2, 11]. Current segment is [0, 8]. We need to move `mo_right` to 11 and `mo_left` to 2.

`mo_right = 9, current_answer = 32, cnt = [2, 3, 1, 2, 1, 1]`

`mo_right = 10, current_answer = 39, cnt = [2, 4, 1, 2, 1, 1]`

`mo_right = 11, current_answer = 54, cnt = [2, 4, 1, 2, 1, 2]`

`mo_left = 1, current_answer = 54, cnt = [1, 4, 1, 2, 1, 2]`

`mo_left = 2, current_answer = 47, cnt = [1, 3, 1, 2, 1, 2]`

Answer for query [2, 11] is 47.

4. Next query is [1, 17]. Current segment is [2, 11]. We need to move `mo_right` to 17 and `mo_left` to 1.

`mo_right = 12, current_answer = 62, cnt = [1, 3, 1, 3, 1, 2]`

`mo_right = 13, current_answer = 87, cnt = [1, 3, 1, 3, 1, 3]`

`mo_right = 14, current_answer = 99, cnt = [1, 3, 1, 3, 2, 3]`

`mo_right = 15, current_answer = 99, cnt = [2, 3, 1, 3, 2, 3]`

`mo_right = 16, current_answer = 105, cnt = [2, 3, 2, 3, 2, 3]`

`mo_right = 17, current_answer = 115, cnt = [2, 3, 3, 3, 2, 3]`

`mo_left = 1, current_answer = 122, cnt = [2, 4, 3, 3, 2, 3]`

Answer for query [1, 17] is 122.

5. Our next goal is query [13, 14]. Notice that it starts in different block from the previous query [1, 17]. Consequently, this is the first time `mo_right` will move to the left. We need to move `mo_left` to 13 and `mo_right` to 14.

`mo_right = 16, current_answer = 112, cnt = [2, 4, 2, 3, 2, 3]`

`mo_right = 15, current_answer = 106, cnt = [2, 4, 1, 3, 2, 3]`

`mo_right = 14, current_answer = 106, cnt = [1, 4, 1, 3, 2, 3]`

`mo_left = 2, current_answer = 99, cnt = [1, 3, 1, 3, 2, 3]`

`mo_left = 3, current_answer = 94, cnt = [1, 2, 1, 3, 2, 3]`

`mo_left = 4, current_answer = 94, cnt = [0, 2, 1, 3, 2, 3]`

`mo_left = 5, current_answer = 92, cnt = [0, 2, 0, 3, 2, 3]`

mo_left = 6, current_answer = 77, cnt = [0, 2, 0, 2, 2, 3]
 mo_left = 7, current_answer = 65, cnt = [0, 2, 0, 2, 1, 3]
 mo_left = 8, current_answer = 62, cnt = [0, 1, 0, 2, 1, 3]
 mo_left = 9, current_answer = 53, cnt = [0, 1, 0, 1, 1, 3]
 mo_left = 10, current_answer = 28, cnt = [0, 1, 0, 1, 1, 2]
 mo_left = 11, current_answer = 27, cnt = [0, 0, 0, 1, 1, 2]
 mo_left = 12, current_answer = 12, cnt = [0, 0, 0, 1, 1, 1]
 mo_left = 13, current_answer = 9, cnt = [0, 0, 0, 0, 1, 1]
 Answer for query [13,14] is 9.

6. Next query is [16, 17]. Notice, however, that now we do not need to move mo_right to the left. We need to move mo_left to 16 and mo_right to 14.

mo_right = 15, current_answer = 9, cnt = [1, 0, 0, 0, 1, 1]
 mo_right = 16, current_answer = 11, cnt = [1, 0, 1, 0, 1, 1]
 mo_right = 17, current_answer = 17, cnt = [1, 0, 2, 0, 1, 1]
 mo_left = 14, current_answer = 12, cnt = [1, 0, 2, 0, 1, 0]
 mo_left = 15, current_answer = 8, cnt = [1, 0, 2, 0, 0, 0]
 mo_left = 16, current_answer = 8, cnt = [0, 0, 2, 0, 0, 0]
 Answer for [16, 17] is 8.

7. The last query is [17, 17]. It requires us to move mo_left one unit to the right.

mo_left = 17, current_answer = 2, cnt = [0, 0, 1, 0, 0, 0]
 Answer for this query is 2.

Now the important part comes: we must **output answers** not in order we've achieved them, but in **order they were asked**.

Output:

27
 6
 47
 8
 9
 122
 2

Implementation

Here is the C++ implementation for the above problem:

```

#include <bits/stdc++.h>
using namespace std;

int N, Q;
  
```

```

// Variables, that hold current "state" of computation
long long current_answer;
long long cnt[100];

// Array to store answers (because the order we achieve them is not the same)
long long answers[100500];
int BLOCK_SIZE;
int arr[100500];

// We will represent each query as three numbers: L, R, idx. Idx is the
// the position (in original order) of this query.
pair< pair<int, int>, int> queries[100500];

// Essential part of Mo's algorithm: comparator, which we will
// use with std::sort. It is a function, which must return True
// if query x must come earlier than query y, and False otherwise.
inline bool mo_cmp(const pair< pair<int, int>, int> &x,
                  const pair< pair<int, int>, int> &y)
{
    int block_x = x.first.first / BLOCK_SIZE;
    int block_y = y.first.first / BLOCK_SIZE;
    if(block_x != block_y)
        return block_x < block_y;
    return x.first.second < y.first.second;
}

// When adding a number, we first nullify it's effect on current
// answer, then update cnt array, then account for it's effect again.
inline void add(int x)
{
    current_answer -= cnt[x] * cnt[x] * x;
    cnt[x]++;
    current_answer += cnt[x] * cnt[x] * x;
}

// Removing is much like adding.
inline void remove(int x)
{
    current_answer -= cnt[x] * cnt[x] * x;
    cnt[x]--;
    current_answer += cnt[x] * cnt[x] * x;
}

```

```
int main()
{
    cin.sync_with_stdio(false);
    cin >> N >> Q;
    BLOCK_SIZE = static_cast<int>(sqrt(N));

    // Read input array
    for(int i = 0; i < N; i++)
        cin >> arr[i];

    // Read input queries, which are 0-indexed. Store each query's
    // original position. We will use it when printing answer.
    for(int i = 0; i < Q; i++) {
        cin >> queries[i].first.first >> queries[i].first.second;
        queries[i].second = i;
    }

    // Sort queries using Mo's special comparator we defined.
    sort(queries, queries + Q, mo_cmp);

    // Set up current segment [mo_left, mo_right].
    int mo_left = 0, mo_right = -1;

    for(int i = 0; i < Q; i++) {
        // [left, right] is what query we must answer now.
        int left = queries[i].first.first;
        int right = queries[i].first.second;

        // Usual part of applying Mo's algorithm: moving mo_left
        // and mo_right.
        while(mo_right < right) {
            mo_right++;
            add(arr[mo_right]);
        }
        while(mo_right > right) {
            remove(arr[mo_right]);
            mo_right--;
        }

        while(mo_left < left) {
            remove(arr[mo_left]);
```

```

        mo_left++;
    }
    while(mo_left > left) {
        mo_left--;
        add(arr[mo_left]);
    }

    // Store the answer into required position.
    answers[queries[i].second] = current_answer;
}

// We output answers *after* we process all queries.
for(int i = 0; i < Q; i++)
    cout << answers[i] << "\n";
return 0;
}

```

Same solution without global variables (the way I like to implement it):

```

#include <bits/stdc++.h>
using std::vector;
using std::tuple;

/*
 * Take out adding\removing logic into separate class.
 * It provides functions to add and remove numbers from
 * our structure, while maintaining cnt[] and current_answer.
 */
struct Mo
{
    static constexpr int MAX_VALUE = 1005000;
    vector<long long> cnt;
    long long current_answer;

    void process(int number, int delta)
    {
        current_answer -= cnt[number] * cnt[number] * number;
        cnt[number] += delta;
        current_answer += cnt[number] * cnt[number] * number;
    }
}

```

```
public:
    Mo()
    {
        cnt = vector<long long>(MAX_VALUE, 0);
        current_answer = 0;
    }

    long long get_answer() const
    {
        return current_answer;
    }

    void add(int number)
    {
        process(number, 1);
    }

    void remove(int number)
    {
        process(number, -1);
    }
};

int main()
{
    int N, Q, BLOCK_SIZE;
    std::cin.sync_with_stdio(false);
    std::cin >> N >> Q;
    BLOCK_SIZE = static_cast<int>(sqrt(N));

    // No global variables, everything inside.
    vector<int> arr(N);
    vector<long long> answers(Q);
    vector< tuple<int, int, int> > queries;
    queries.reserve(Q);

    for(int i = 0; i < N; i++)
        std::cin >> arr[i];

    for(int i = 0; i < Q; i++) {
        int le, rg;
        std::cin >> le >> rg;
```

```

        queries.emplace_back(le, rg, i);
    }

    // Comparator as a lambda-function, which catches BLOCK_SIZE
    // from the above definition.
    auto mo_cmp = [BLOCK_SIZE](const tuple<int, int, int> &x,
                                const tuple<int, int, int> &y) -> bool {
        int block_x = std::get<0>(x) / BLOCK_SIZE;
        int block_y = std::get<0>(y) / BLOCK_SIZE;
        if(block_x != block_y)
            return block_x < block_y;
        return std::get<1>(x) < std::get<1>(y);
    };

    std::sort(queries.begin(), queries.end(), mo_cmp);

    Mo solver;
    int mo_left = 0, mo_right = -1;

    // Usual Mo's algorithm stuff.
    for(const auto &q: queries) {
        int left, right, answer_idx;
        std::tie(left, right, answer_idx) = q;

        while(mo_right < right) {
            mo_right++;
            solver.add(arr[mo_right]);
        }
        while(mo_right > right) {
            solver.remove(arr[mo_right]);
            mo_right--;
        }

        while(mo_left < left) {
            solver.remove(arr[mo_left]);
            mo_left++;
        }
        while(mo_left > left) {
            mo_left--;
            solver.add(arr[mo_left]);
        }
    }

```

LIVE EVENTS

```
        answers[answer_idx] = solver.get_answer();
    }

    for(int i = 0; i < Q; i++)
        std::cout << answers[i] << "\n";
    return 0;
}
```

Practice problems

In case you want to try out implementing Mo's technique for yourself, check out these problems:

1. [Kriti and her birthday gift](#)

Difficulty: easy.

Prerequisites: string hashing.

Comment: tests for this problem are flawed (you can get **at most 33 out of 100** for this problem), because both setter's and tester's implementations from the editorial have the same bug in hashing function. Can you see what it is?

Reference implementation: [here](#).

2. [SUBSTRINGS COUNT](#)

Difficulty: easy.

Prerequisites: string hashing.

Comment: almost the same as previous problem, but this one asks slightly different question and has well-formed tests.

Reference implementation: [here](#).

3. [Sherlock and inversions](#)

Difficulty: medium.

Prerequisites: segment tree\binary indexed tree, coordinate compression.

Comment: nice and clean problem on Mo's algorithm. Might seem difficult for people not familiar with prerequisites.

Reference implementation: [here](#).

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Arjit Srivastava 2 months ago

Thank you for this amazing post, Mike! :)

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Viet Nguyenkhanh 2 months ago

thanks you very much, the notes very interesting

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Bob Hob 2 months ago

How do you create such beautiful diagrams? :^o

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Mike Koltsov ⚡ Author 2 months ago

HackerEarth staff did this for me

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Anand Hariharan 2 months ago

nice post Mike !

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Aman Goel 16 days ago

Amazing!

I got to learn Fenwick tree as well because of the last problem

Thanks a lot for such a nice explanation of the algorithm

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Arun Prasad 2 days ago

Very Good Article.

What is the difference between Mo's Algorithm and Square Root decomposition ?

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Mike Koltsov ⚡ Author 2 days ago

My opinion on this matter: Mo's algorithm is one of the tricks of sqrt decomposition. Some people think that Mo's algorithm can not be distinguished from sqrt decomposition, but I think Mo's algorithm is very interesting itself.

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AUTHOR



Mike Koltsov

Student laboratory of Mail...

Saint Petersburg, Russia

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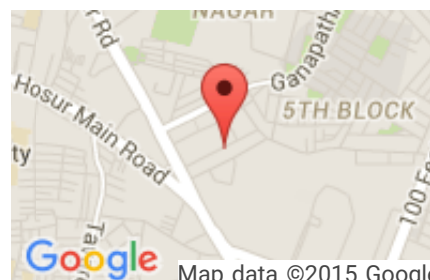
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Map data ©2015 Google
IIIrd Floor, Salarpuria Business Center,
4th B Cross Road, 5th A Block,
Koramangala Industrial Layout,

Bangalore, Karnataka 560095, India.

✉ contact@hackerearth.com

☎ +91-80-4155-4695

☎ +1-650-461-4192



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