

Fenwick Tree

Binary Indexing, Least Significant One-bit

beOI Training



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Dynamic Range Sum Query

Well, can't we simply use a Segment Tree?

Yes we can, but the Fenwick Tree datastructure has some cool advantages.

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How it works

Let's define $rsq(i)$ as the sum of the elements from 1 up to i

$$rsq(i) = A[1] + A[2] + \cdots + A[i]$$

We can then answer any range sum query by computing

$$rsq(a, b) = rsq(b) - rsq(a - 1)$$

We also need a point update function $update(k, v)$ which updates element k to a new value v .

Like segment trees, Fenwick trees address these functions in $\mathcal{O}(\log n)$.

Least Significant One-bit

The **least significant 1-bit** of an integer is the rightmost 1-bit of its **binary representation**.

$$(42)_{10} = (1010\textcolor{red}{1}0)_2$$

$$(1337)_{10} = (1010011100\textcolor{red}{1})_2$$

$$(2016)_{10} = (11111\textcolor{red}{1}00000)_2$$

Let's define function $LSOne(i)$ as the least significant 1-bit isolated from i

$$LSOne(42) = 0000\textcolor{red}{1}0$$

$$LSOne(1337) = 0000000000\textcolor{red}{1}$$

$$LSOne(2016) = 0000\textcolor{red}{1}00000$$

Binary Indexing

Like in segment trees, each node is responsible for (i.e. stores the sum of) a certain range of elements.

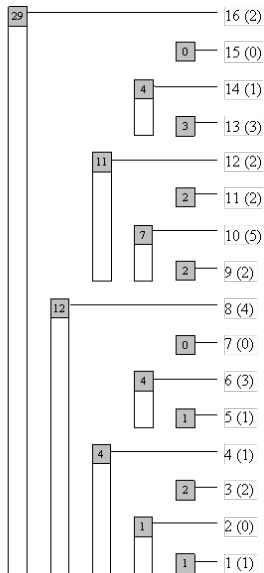
Specifically, node i is responsible for range

$$[i - LSONe(i) + 1; i]$$

This range has size $LSONe(i)$.

e.g. $LSONe(2016) = 32$, thus node 2016 will store sum of range $[1985; 2016]$ which has size 32.

Visualization



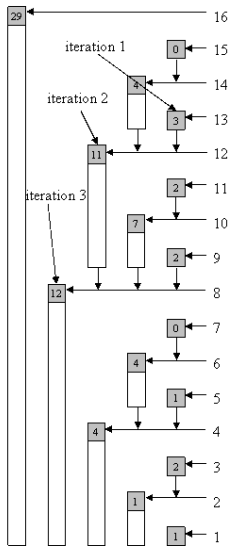
Querying

One can observe that

$$\begin{aligned}rsq(13) &= ft[1101] + ft[1100] + ft[1000] \\ &= rsq(13, 13) + rsq(9, 12) + rsq(1, 8)\end{aligned}$$

So to obtain $rsq(i)$ we need to iteratively strip off $LSOne(i)$ and sum the values stored at the nodes we come across.

Query Visualization



Query Complexity

The number of iterations is equal to the number of 1-bits in the starting index.

If the starting index n consists of b bits, we need at most b operations to find $rsq(n)$.

$$\mathcal{O}(b) = \mathcal{O}(\log n)$$

LSOne Implementation

There is a binary trick to compute the Least Significant One-bit really fast

$$LSOne(i) = (i \& (-i))$$

Query Implementation

```
int rsq(int i) {  
    int sum = 0;  
    for(; i > 0; i -= LSOne(i))  
        sum += ft[i];  
    return sum;  
}
```

Updating

When updating index i , we need to update all nodes that are responsible for index i .

Specifically, if we update index i from v_{old} to v_{new} , we need to add $v_{\text{new}} - v_{\text{old}}$ to each node responsible for index i .

Updating

One can prove that if node k is responsible for index i , then node $k + LSONe(k)$ is the smallest node $> k$ that is also responsible for i .

$$38 = (1001\textcolor{red}{1}0)_2$$

$$38 + LSONe(38) = 40 = (101000)_2$$

Updating

i is the smallest node responsible for index i .

Thus, to find all nodes responsible for i , we can simply start from i and iteratively add the Least Significant 1-bit.

$$\text{update } (10\color{red}1)_2 = 5$$

$$\text{update } (1\color{red}10)_2 = 6$$

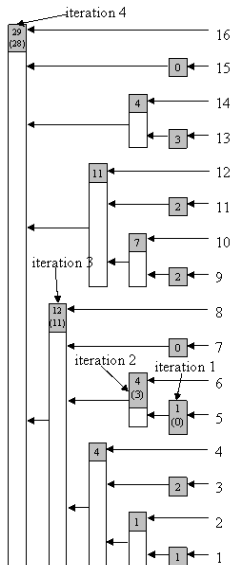
$$\text{update } (\color{red}1000)_2 = 8$$

$$\text{update } (\color{red}10000)_2 = 16$$

\vdots

When do we stop? When we arrive at a node outside our initial array.

Update Visualization



Update Complexity

When adding $LSOne(i)$, the least significant bit is shifted at least 1 place to the left.

So, in the worst case, we need to loop through all bits until we get to an index $> n$.

$$\mathcal{O}(\log n)$$

Update Implementation

```
void adjust(int k, int v) {  
    for (; k <= n; k += LSONe(k))  
        ft[k] += v;  
}
```

Building

To build the Fenwick Tree from the array A , there is no better way than updating each index separately.

$$\mathcal{O}(n \log n)$$

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Time Complexity

	Segment Tree	Fenwick Tree
Query	$\mathcal{O}(\log n)$	$\mathcal{O}(\log n)$
Update	$\mathcal{O}(\log n)$	$\mathcal{O}(\log n)$
Build	$\mathcal{O}(n)$	$\mathcal{O}(n \log n)$

In general, Fenwick Trees are slightly more efficient because of the fast bit trick and few memory accesses.

Memory Usage

Both use $\mathcal{O}(n)$ memory.

Code

Segment Tree: 40 lines of code

Fenwick Tree: 10 lines of code!

In contest, always use Fenwick if possible!

Applications

Beware: you cannot use Fenwick Trees for any function of your liking.

$$rsq(a, b) = rsq(b) - rsq(a - 1)$$

There is **no** such property for min/max, so Fenwick is not suitable for Dynamic RMQ.