



Lebanese University  
Faculty of Science  
Computer Science BS Degree

# Advanced Algorithms

## I3341

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# Course Chapters

as per the textbook



1. Introduction
2. Data Structures and Libraries
3. Problem Solving Paradigms
4. Graph
5. Mathematics
6. String Processing
7. Computational Geometry

# Problem Solving Paradigms

## Chapter 3

# Chapter Outline

1. Complete Search = Brute Force
  - a. Iterative Complete Search
  - b. Recursive Complete Search
  - c. Tips
2. Divide and Conquer
  - a. Interesting Usages of Binary Search
3. Greedy
  - a. Examples
4. Dynamic Programming
  - a. DP Illustration
  - b. Classical Examples
  - c. Non-Classical Examples

# Overview and Motivation

# Let us focus on Divide & Conquer

- ❑ 1. Find largest & smallest element
  - $\rightarrow O(n)$  Complete Search
- ❑ 2. Find kth smallest element
  - find smallest and replace with big number
  - repeat k times
  - if  $k = n/2 \rightarrow O(n \times n/2) \sim O(n^2)$
  - better sort then choose kth  $\rightarrow O(n \log n) \rightarrow$  Divide and Conquer
  - better  $O(n) \rightarrow$  also Divide and Conquer
- ❑ 3. Find largest gap  $g / x, y \in A \ \& \ g = |x - y|$ 
  - consider every pair  $\rightarrow O(n^2)$
  - largest – smallest  $\rightarrow O(n) \rightarrow$  Greedy
- ❑ 4. Find longest increasing subsequence
  - try all  $O(2^n)$  possible subsequences
  - not feasible for all  $n \leq 10K$
  - $\rightarrow O(n^2)$  DP
  - $\rightarrow O(n \log n)$  Greedy

# Divide & Conquer

# Introduction

- Based on recursion.

- How it works ?

- Recursively break down a problem into two or more sub problems of the same type, until they become simple enough to be solved directly.
- The solutions to the sub problems are then combined to give a solution to the original problem.



# Strategy?

- The D & C strategy solves a problem by:
  1. **Divide**: Breaking the problem into sub problems that are themselves smaller instances of the same type of problem.
  2. **Recursion**: Recursively solving these sub problems.
  3. **Conquer**: Appropriately combining their answers.

# Does Divide and Conquer Always Work?

❑ NO !!

❑ For all problems it is not possible to find the subproblems which are the same size and  $D$  &  $C$  is not a choice for all problems.

# Visualization

- ❑ Assume that  $n$  is the size of the original problem.
- ❑ Divide the problem into  $b$  sub problems with each of size  $n/b$  (for some constant  $b$ ).
- ❑ Solve the sub problems recursively and combine their solutions to get the solution for the original problem.

# Visualization

```
DivideandConquer(P){  
    if(small(P))  
        // P is very small so that a solution is obvious  
        return solution(P);  
  
    Divide P into b subproblems: P1, P2, ... Pb  
  
    return (  
        combine(  
            DivideandConquer(P1),  
            DivideandConquer(P2),  
            ...,  
            DivideandConquer(Pb)  
        )  
    )  
}
```

# Examples

- ❑ We have already solved many problems based on *D & C* strategy: like [Binary Search](#), [Merge Sort](#), [Quick Sort](#), etc....
- ❑ Looking for a name in a phone book: We have a phone book with names in alphabetical order. Given a name, how do we find whether that name is there in the phone book or not?
- ❑ Finding our car in a parking lot.
- ❑ ...

# Advantages of Divide and Conquer

- ❑ **Solving difficult problems:** *D & C* is a powerful method for solving difficult problems: Tower of Hanoi problem.
- ❑ **Parallelism:** Since *D & C* allows us to solve the subproblems independently.

# Disadvantages of Divide and Conquer

- ❑ Recursion is slow.
  - Overhead of the repeated subproblem calls.
  - Stack for storing the calls.
  
- ❑ Sometimes more complicated than an iterative approach.
  - Example: add  $n$  numbers, a simple loop to add them up in sequence is much easier than a *D & C* approach that breaks the set of numbers into two halves, adds them recursively, and then adds the sums.

# Divide and Conquer Applications

- ❑ Binary Search
- ❑ Merge Sort and Quick Sort
- ❑ Median Finding
- ❑ Min and Max Finding
- ❑ Matrix Multiplication
- ❑ Closest Pair problem



# Binary Search: The Ordinary Usage

- ❑ As the size of search space is halved (in a binary fashion) after each check, the complexity of this algorithm is  $O(\log n)$ .
- ❑ There are built-in library routines for this algorithm, e.g.
  - the C++ STL `algorithm::lower_bound` / `algorithm::upper_bound`
  - the Java `Collections.binarySearch`

# Binary Search: The Ordinary Usage

```
#include <iostream>      // std::cout
#include <algorithm>     // std::lower_bound,
                        // std::upper_bound, std::sort
#include <vector>        // std::vector
int main () {
    int myints[] = {10,20,30,30,20,10,10,20};
    std::vector<int> v(myints,myints+8);
    std::sort (v.begin(), v.end());
    std::vector<int>::iterator low,up;
    low=std::lower_bound (v.begin(), v.end(), 20);
    up= std::upper_bound (v.begin(), v.end(), 20);

    std::cout << "lower_bound at position "
               << (low- v.begin()) << '\n';
    std::cout << "upper_bound at position "
               << (up - v.begin()) << '\n';
    return 0;
}
```

Output:

```
lower_bound at position 3
upper_bound at position 6
```

```
import java.util.List;
import java.util.ArrayList;
import java.util.Collections;
public class GFG
{
    public static void main(String[] args)
    {
        List al = new ArrayList();
        al.add(10);
        al.add(20);
        al.add(1);
        al.add(2);
        al.add(3);
        Collections.sort(al);

        int index = Collections.binarySearch(al, 10);
        System.out.println(index);

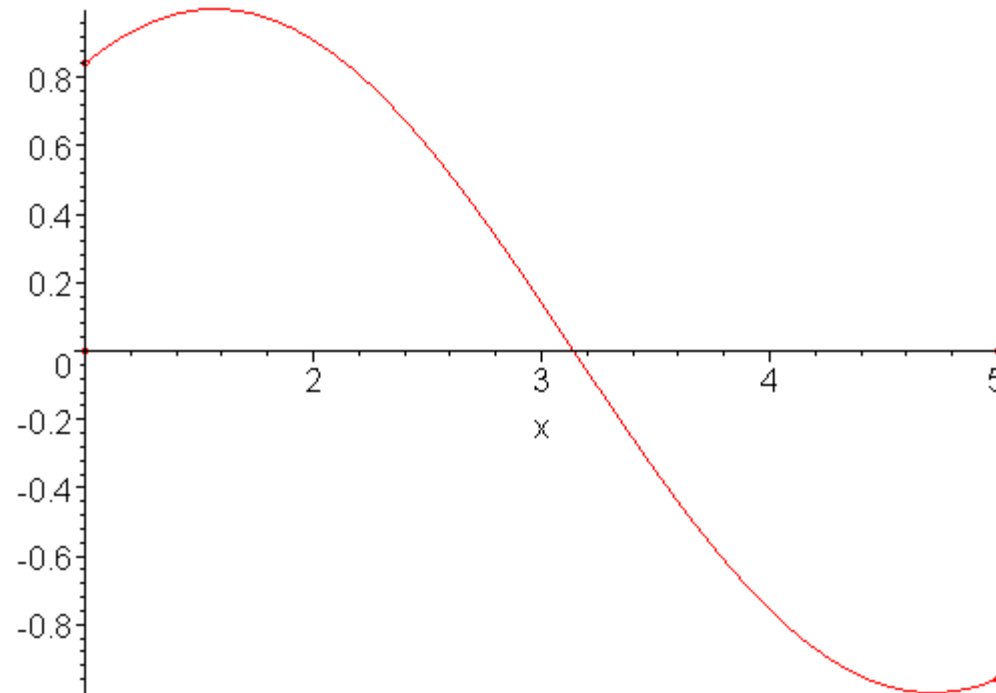
        // 13 is not present. 13 would have been inserted
        // at position 4. So the function returns (-4-1)
        // which is -5.
        index = Collections.binarySearch(al, 13);
        System.out.println(index);
    }
}
```

Output:

```
3
-5
```

# Bisection Method

- Find the root of a function that may be difficult to compute directly.



# Bisection Method - Example

- ❑ You buy a car with loan and now want to pay the loan in monthly installments of  $d$  dollars for  $m$  months.
- ❑ Suppose the value of the car is originally  $v$  dollars and the bank charges an interest rate of  $i\%$  for any unpaid loan at the end of each month. What is the amount of money  $d$  that you must pay per month (to 2 digits after the decimal point)?

# Bisection Method - Example

- Suppose  $d = 576.19$ ,  $m = 2$ ,  $v = 1000$ , and  $i = 10\%$ .
  - After one month, your debt becomes  $1000 \times (1.1) - 576.19 = 523.81$ .
  - After two months, your debt becomes  $523.81 \times (1.1) - 576.19 \approx 0$ .

Now let's reverse the process:

- If we are only given  $m = 2$ ,  $v = 1000$ , and  $i = 10\%$ , how would we determine that  $d = 576.19$ ?

In other words, find the root  $d$  such that the debt payment function  $f(d, m, v, i) \approx 0$ .

# Bisection Method - Example

- Pick a reasonable range  $[a..b]$  as starting points.

**For the bisection method to work, we must ensure that the function values of the two extreme points in the initial Real range  $[a..b]$ , i.e.  $f(a)$  and  $f(b)$  have opposite signs**

- Fix  $d$  within the range  $[a..b]$

*where  $a = 0.01$  as we have to pay at least one cent and  $b = (1+i\%) \times v$  as the earliest we can complete the payment is  $m = 1$  if we pay exactly  $(1 + i\%) \times v$  dollars after one month. In this example,  $b = (1+0.1) \times 1000 = 1100.00$  dollars.*

# Bisection Method

a	b	$d = \frac{a + b}{2}$	Status: $f(d, m, v, i)$	action
0.01	1100.00	550.005	Undershoot by 54.9895	↑ d
550.005	1100.00	825.0025	Overshoot by 522.50525	↓ d
550.005	825.0025	687.50375	Overshoot by 233.757875	↓ d
550.005	687.50375	618.754375	Overshoot by 89.384187	↓ d
550.005	618.754375	584.379688	Overshoot by 17.197344	↓ d
550.005	584.379688	567.192344	Undershoot by 18.896078	↑ d
567.192344	584.379688	575.786016	Undershoot by 0.849366	↑ d
...	...	....	After few iterations	...
...	...	576.190476	stop; error is now less than $\epsilon$	answer = 576.19

$$(1.1) * 1000 - 550.005 = 549.995$$

$$(1.1) * 549.995 - 550.005 = 54.9895$$

# Bisection Method

- ❑ It requires  $O\left(\log_2\left(\frac{b-a}{\varepsilon}\right)\right)$  iterations to get an answer
- ❑ In this example, bisection method only takes  $\log_2\left(\frac{1099.99}{\varepsilon}\right)$  tries.
  - Using a small  $\varepsilon = 1e-9$ , this yields only  $\approx 40$  iterations.
  - Even if we use a smaller  $\varepsilon = 1e-15$ , we will still only need  $\approx 60$  tries.
  - Notice that the number of tries is *small*.
  - The bisection method is much more efficient compared to exhaustively evaluating each possible value of  $d = [0.01..1100.00]/\varepsilon$ .
- ❑ Note: The bisection method can be written with a loop that tries the values of  $d \approx 40$  to 60 times using 'binary search' technique.



# Bisection Method

```
double lo = 0.01, hi = 1100.00, mid ;

for (int i = 0; i < 50; i++) { // log2 ((1099.99 - 0.0) / 1e-9) ≈ 40
    mid = (lo + hi) / 2.0; // looping 50 times should be precise enough
    if (f(mid)>0) low= mid;
    else hi = mid;
}
return mid;
```

# Segment Tree Range Minimum Query (RMQ) problem

# Segment Tree

## Range Minimum Query (RMQ) problem

- ❑ Data structure which can efficiently answer *dynamic* range queries.
- ❑ One such range query is the problem of finding the index of the minimum element in an array within range  $[i..j]$ .
- ❑ This is more commonly known as the Range Minimum Query (RMQ) problem.

# Example

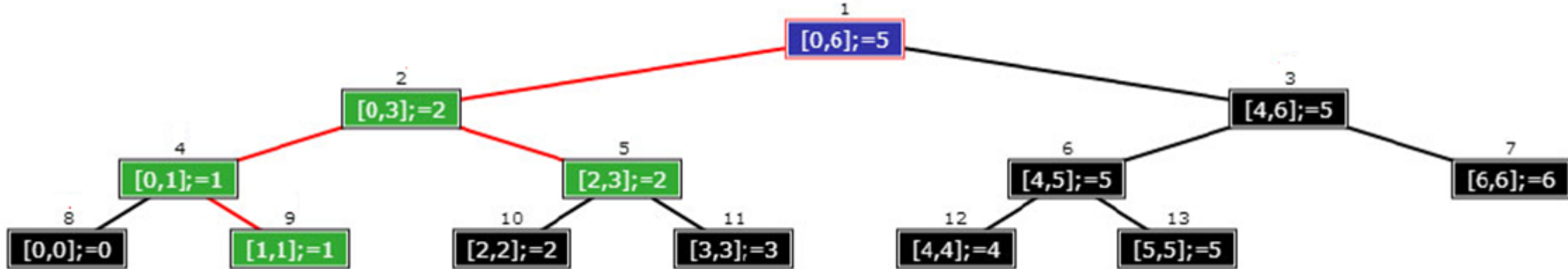
Array	Values	18	17	13	19	15	11	20
A	Indices	0	1	2	3	4	5	6

- Given an array  $A$  of size  $n = 7$ ,
- $\text{RMQ}(1, 3) = 2$ , as the index 2 contains the minimum element among  $A[1]$ ,  $A[2]$ , and  $A[3]$ .
- $\text{RMQ}(3, 4) = 4$ ,  $\text{RMQ}(0, 0) = 0$ ,  $\text{RMQ}(0, 1) = 1$ , and  $\text{RMQ}(0, 6) = 5$ .
- There are several ways to implement the RMQ. One trivial algorithm is to simply iterate the array from index  $i$  to  $j$  and report the index with the minimum value, but this will run in  $O(n)$  time per query. When  $n$  is large and there are many queries, such an algorithm may be infeasible.

# Implementation

- Segment tree like binary heap using static implementation.
  - We call the array  $st$ .
  - Index 1 (skipping index 0) is the root and the left and right children of index  $p$  are index  $2 \times p$  and  $(2 \times p) + 1$  respectively.
  - The value of  $st[p]$  is the RMQ value of the segment associated with index  $p$ .

# Implementation



- ❑ The root of segment tree represents segment  $[0, n-1]$ .
- ❑ For each segment  $[L, R]$  stored in index  $p$  where  $L \neq R$ , the segment will be split into  $[L, (L+R)/2]$  and  $[(L+R)/2+1, R]$  in a left and right vertices.
- ❑ The left sub-segment and right sub-segment will be stored in index  $2 \times p$  and  $(2 \times p) + 1$  respectively.
- ❑ When  $L = R$ , it is clear that  $st[p] = L$  (or  $R$ ). Otherwise, we will recursively build the segment tree, comparing the minimum value of the left and the right sub-segments and updating the  $st[p]$  of the segment.

# Implementation

```
typedef vector<int> vi;
class SegmentTree {
private: vi st, A;
vector<int> vi;
int n;
int left (int p) { return p << 1; }
int right(int p) { return (p << 1) + 1; }
void build(int p, int L, int R) {
    if (L == R) st[p] = L;
    else {
        build(left(p) , L, (L + R) / 2);
        build(right(p), (L + R) / 2 + 1, R);
        int p1 = st[left(p)], p2 = st[right(p)];
        st[p] = (A[p1] <= A[p2]) ? p1 : p2;
    }
}
```

```
class SegmentTree {
private int[] st, A;
private int n;
private int left (int p) { return p << 1; }
private int right(int p) { return (p << 1) + 1; }

private void build(int p, int L, int R) {
    if (L == R) st[p] = L;
    else {
        build(left(p) , L, (L + R) / 2);
        build(right(p), (L + R) / 2 + 1, R);
        int p1 = st[left(p)], p2 = st[right(p)];
        st[p] = (A[p1] <= A[p2]) ? p1 : p2;
    }
}
```

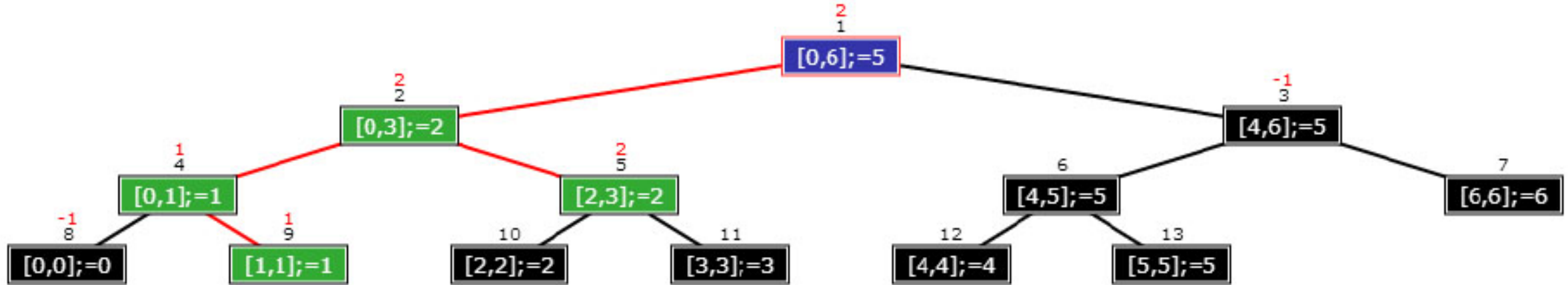
# Implementation

- ❑ With the segment tree ready, answering an RMQ can be done in  $O(\log n)$ .
- ❑ The answer for  $\text{RMQ}(i, i)$  is trivial—simply return  $i$  itself.
- ❑ However, for the general case  $\text{RMQ}(i, j)$ , further checks are needed.

Let  $p1 = \text{RMQ}(i, (i+j)/2)$  and  $p2 = \text{RMQ}((i+j)/2+1, j)$ .  
Then  $\text{RMQ}(i, j)$  is  $p1$  if  $A[p1] \leq A[p2]$  or  $p2$  otherwise.



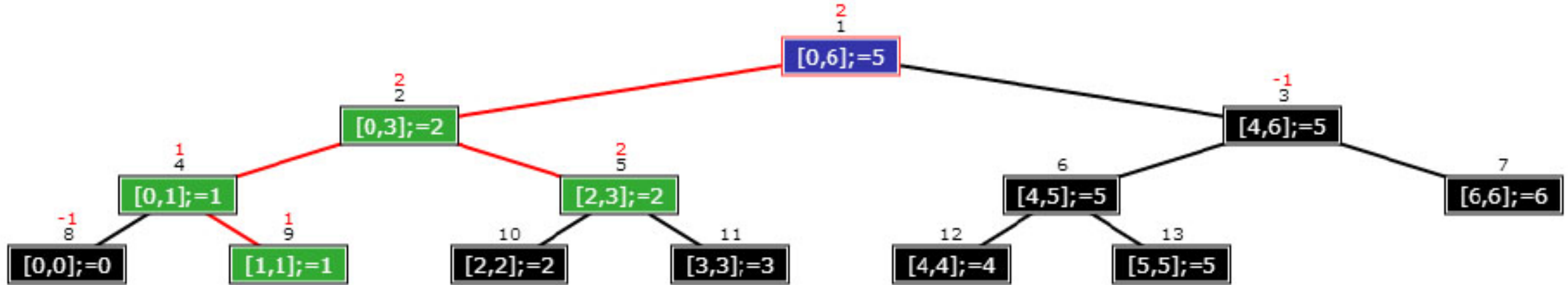
# Implementation



## □ RMQ(1, 3):

- Start from the root (index 1) which represents segment  $[0, 6]$ . We cannot use the stored minimum value of segment  $[0, 6]$ .
- From the root, we only have to go to the left subtree as the root of the right subtree represents segment  $[4, 6]$  which is outside the desired range in  $\text{RMQ}(1, 3)$ .

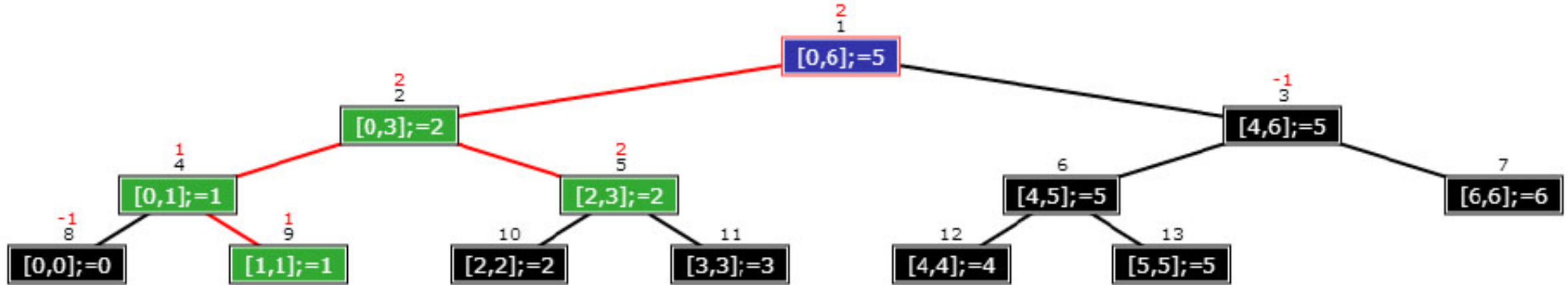
# Implementation



## □ RMQ(1, 3):

- We are now at the root of the left subtree (index 2) that represents segment  $[0, 3]$ . This segment  $[0, 3]$  is still larger than the desired  $\text{RMQ}(1, 3)$ .
- $\text{RMQ}(1, 3)$  intersects both the left sub-segment  $[0, 1]$  (index 4) and the right sub-segment  $[2, 3]$  (index 5) of segment  $[0, 3]$ , so we have to explore both subtrees (sub-segments).

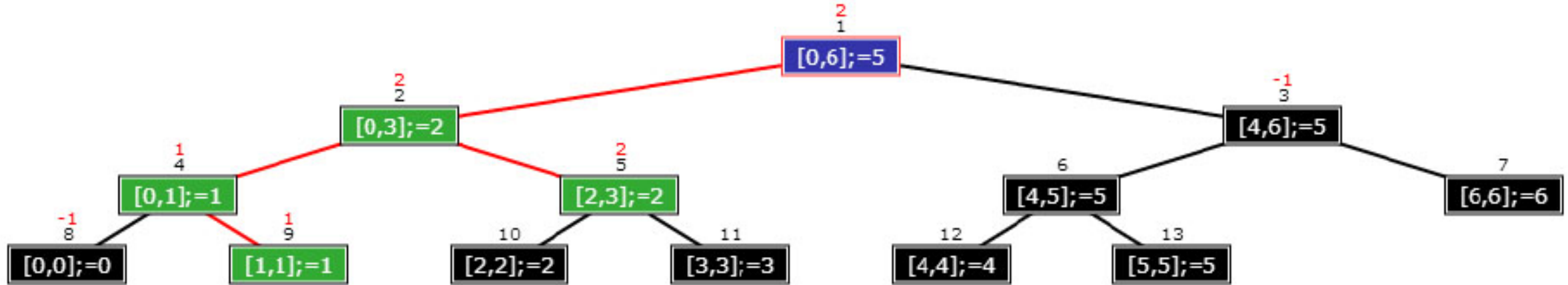
# Implementation



## □ RMQ(1, 3):

- The left segment  $[0, 1]$  (index 4) of  $[0, 3]$  (index 2) is not yet inside the  $\text{RMQ}(1, 3)$ . From segment  $[0, 1]$  (index 4), we move right to segment  $[1, 1]$  (index 9), which is now inside  $[1, 3]$ .
- At this point, we know that  $\text{RMQ}(1, 1) = \text{st}[9] = 1$  and we can return this value to the caller. The right segment  $[2, 3]$  (index 5) of  $[0, 3]$  (index 2) is inside the required  $[1, 3]$ . From the stored value inside this vertex, we know that  $\text{RMQ}(2, 3) = \text{st}[5] = 2$ . We do not need to traverse further down.

# Implementation

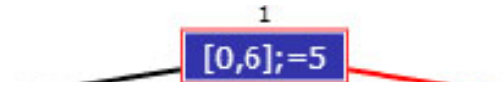


## □ RMQ(1, 3):

- Now, back in the call to segment  $[0, 3]$  (index 2), we now have  $p1 = \text{RMQ}(1, 1) = 1$  and  $p2 = \text{RMQ}(2, 3) = 2$ . Because  $A[p1] > A[p2]$  since  $A[1] = 17$  and  $A[2] = 13$ , we now have  $\text{RMQ}(1, 3) = p2 = 2$ . This is the final answer.

# Implementation

- Build the segment tree of Array  $A = \{18, 17, 13, 19, 15, 11, 20\}$  and RMQ(4, 6)



# Implementation

```
int rmq(int p, int L, int R, int i, int j) {
    if (i > R || j < L) return -1;
        // current segment outside query range
    if (L >= i && R <= j) return st[p];
        // inside query range
    // compute the min position in the left and right
    int p1 = rmq(left(p), L, (L+R) / 2, i, j);
    int p2 = rmq(right(p), (L+R) / 2 + 1, R, i, j);
    if (p1 == -1) return p2;
        // if we try to access segment outside query
    if (p2 == -1) return p1;
        // same as above
    return (A[p1] <= A[p2]) ? p1 : p2;
}
```

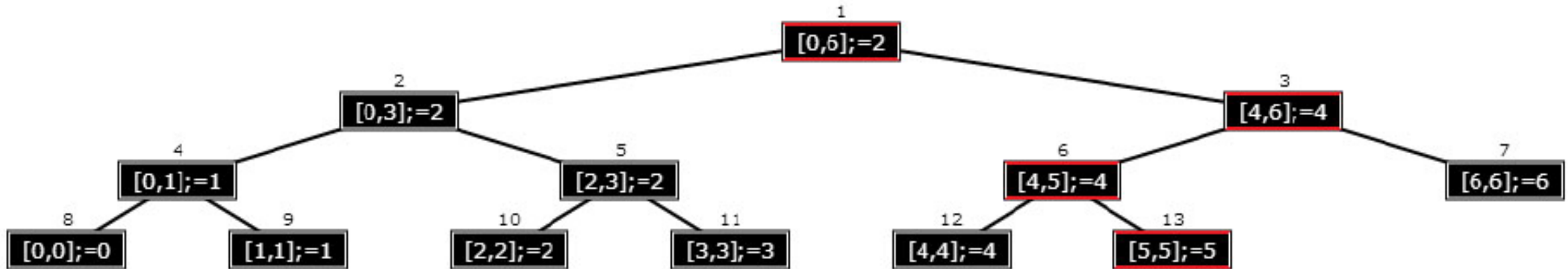
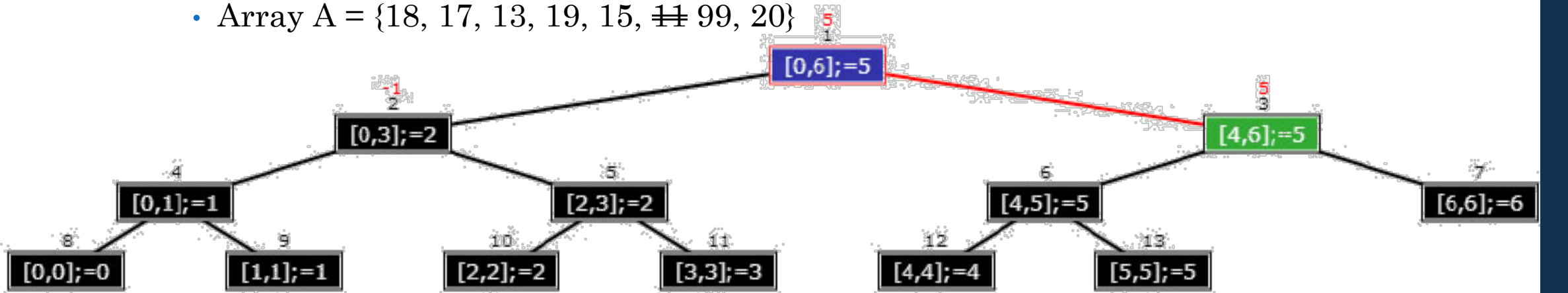
```
private int rmq(int p, int L, int R, int i, int j) {
    if (i > R || j < L) return -1;
        // current segment outside query range
    if (L >= i && R <= j) return st[p];
        // inside query range
    // compute the min position in the left and right
    int p1 = rmq(left(p), L, (L+R) / 2, i, j);
    int p2 = rmq(right(p), (L+R) / 2 + 1, R, i, j);
    if (p1 == -1) return p2;
        // if we try to access segment outside query
    if (p2 == -1) return p1;
        // same as above
    return (A[p1] <= A[p2]) ? p1 : p2; }
```

# Usage

- ❑ If the array  $A$  is static (i.e. unchanged after it is instantiated), then using a Segment Tree to solve the RMQ problem is OVERKILL as there exists a Dynamic Programming (DP) solution that requires  $O(n \log n)$  one-time pre-processing and allows for  $O(1)$  per RMQ.
  
- ❑ Segment Tree is useful if the underlying array is frequently updated (**dynamic**).

# Implementation

- Array A = {18, 17, 13, 19, 15, 99, 20}

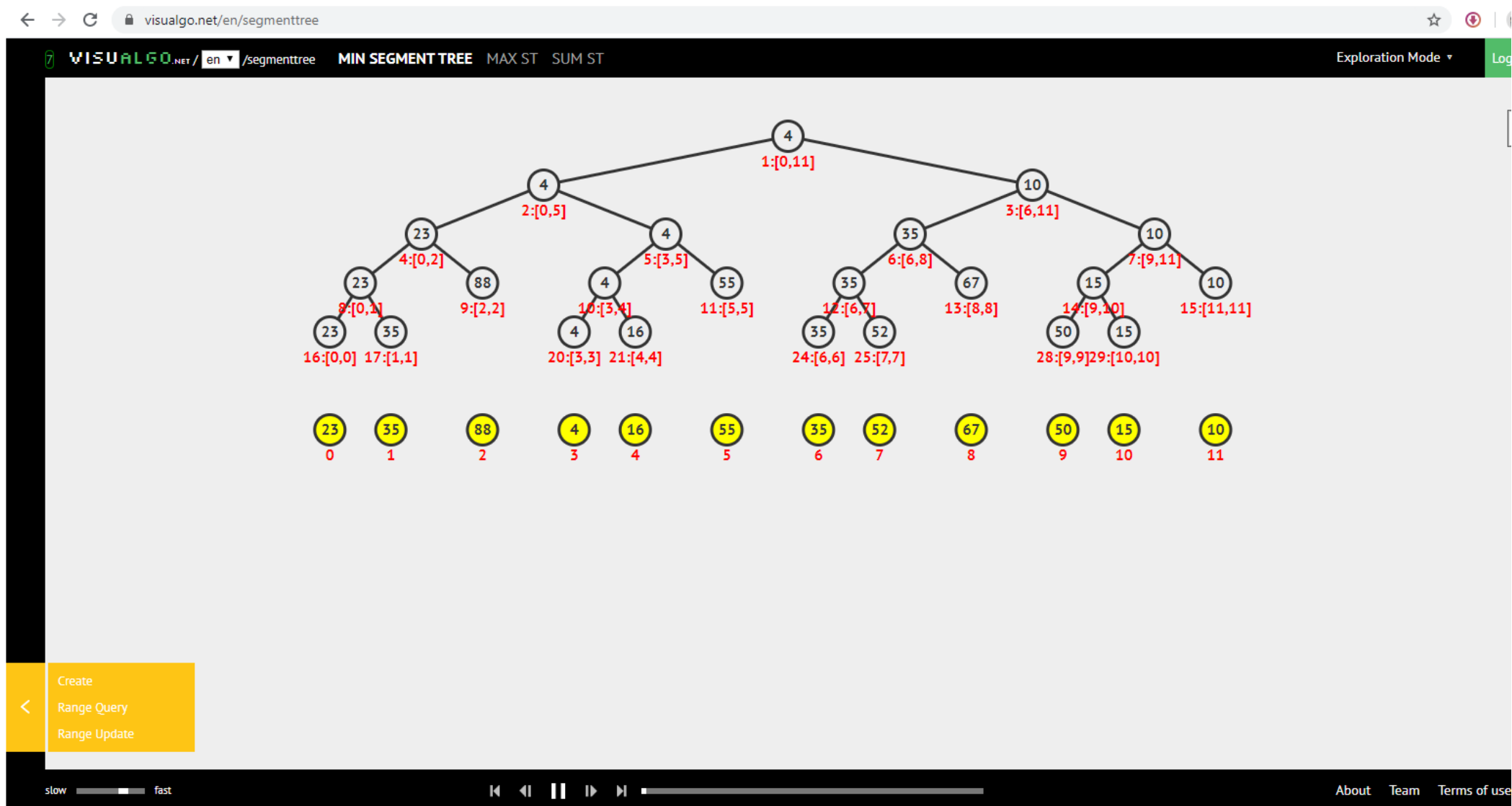




# Implementation

- ❑ We just need to update the vertices along the leaf to root path in  $O(\log n)$ .
- ❑ For comparison, the DP solution requires another  $O(n \log n)$  pre-processing to update the structure and is ineffective for such dynamic updates.

# <https://visualgo.net/en/segmenttree>



# Binary Indexed (Fenwick) Tree

# Binary Indexed (Fenwick) Tree

□ The Fenwick Tree is a useful data structure for implementing *dynamic cumulative frequency tables*.

## □ Example

- we have test scores of  $m = 11$  students  
 $f = \{2, 4, 5, 5, 6, 6, 6, 7, 7, 8, 9\}$  integer values ranging from  $[1..10]$ .

In the following:

- The frequency of each individual test score  $\in [1..10]$
- The cumulative frequency of test scores ranging from  $[1..i]$  denoted by  $cf[i]$ —that is, the sum of the frequencies of test scores 1, 2, ...,  $i$ .

# Binary Indexed (Fenwick) Tree

□ {2,4,5,5,6,6,6,7,7,8,9}

Index/ Score	Frequency f	Cumulative Frequency cf	Short Comment
0	-	-	Index 0 is ignored (as the sentinel value).
1	0	0	$cf[1] = f[1] = 0.$
2	1	1	$cf[2] = f[1] + f[2] = 0 + 1 = 1.$
3	0	1	$cf[3] = f[1] + f[2] + f[3] = 0 + 1 + 0 = 1.$
4	1	2	$cf[4] = cf[3] + f[4] = 1 + 1 = 2.$
5	2	4	$cf[5] = cf[4] + f[5] = 2 + 2 = 4.$
6	3	7	$cf[6] = cf[5] + f[6] = 4 + 3 = 7.$
7	2	9	$cf[7] = cf[6] + f[7] = 7 + 2 = 9.$
8	1	10	$cf[8] = cf[7] + f[8] = 9 + 1 = 10.$
9	1	11	$cf[9] = cf[8] + f[9] = 10 + 1 = 11.$
10	0	11	$cf[10] = cf[9] + f[10] = 11 + 0 = 11.$

# Binary Indexed (Fenwick) Tree

- The cumulative frequency table can also be used as a solution to the Range Sum Query (RSQ) problem. It stores  $RSQ(1, i) \forall i \in [1..n]$  where  $n$  is the largest integer index/score.
  - In the example, we have  $n = 10$ ,  $RSQ(1, 1) = 0$ ,  $RSQ(1, 2) = 1, \dots$ ,  $RSQ(1, 6) = 7, \dots$ ,  $RSQ(1, 8) = 10, \dots$ , and  $RSQ(1, 10) = 11$ .
  - when  $i \neq 1$ ,  
 $RSQ(i, j) = RSQ(1, j) - RSQ(1, i - 1)$ .  
Example:  
 $RSQ(4, 6) = RSQ(1, 6) - RSQ(1, 3) = 7 - 1 = 6$ .

Index/ Score	Frequency f	Cumulative Frequency cf
0	-	-
1	0	0
2	1	1
3	0	1
4	1	2
5	2	4
6	3	7
7	2	9
8	1	10
9	1	11
10	0	11

# Binary Indexed (Fenwick) Tree

- ❑ Fenwick Tree operations are also extremely efficient as they use fast bit manipulation techniques.
- ❑ We will use the function `LSOne(i)` (which is actually  $(i \& (-i))$ ).
- ❑ The operation  $(i \& (-i))$  produces the first Least Significant One-bit in  $i$ .

# LSOne(i)

□ To get the value of the least significant bit that is on (first from the right), use  $T = (i \ \& \ (-i))$ .

## □ Example

- $i = 40$  (base 10) =  $000\dots000101000$  (32 bits, base 2)
- $-i = -40$  (base 10) =  $111\dots111011000$  (two's complement)
- ----- AND
- $T = 8$  (base 10) =  $000\dots000001000$  (3rd bit from right is on)

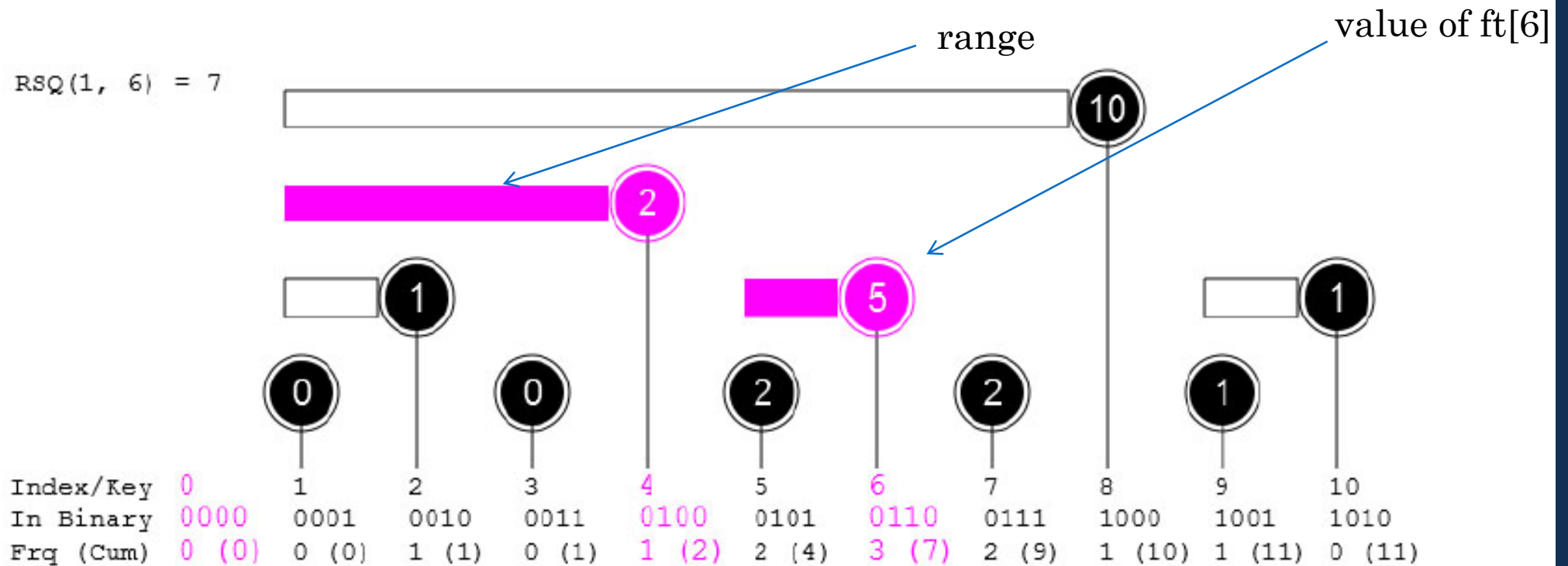


# Implementation

- ❑ The Fenwick Tree is typically implemented as an array (vector).
- ❑ The Fenwick Tree is a tree that is indexed by the *bits of its integer keys*.
- ❑ These integer keys fall within the fixed range  $[1..n]$ —skipping index 0.
- ❑ In the previous table, the scores  $[1..10]$  are the integer keys in the corresponding array with size  $n = 10$  and  $m = 11$  data points.

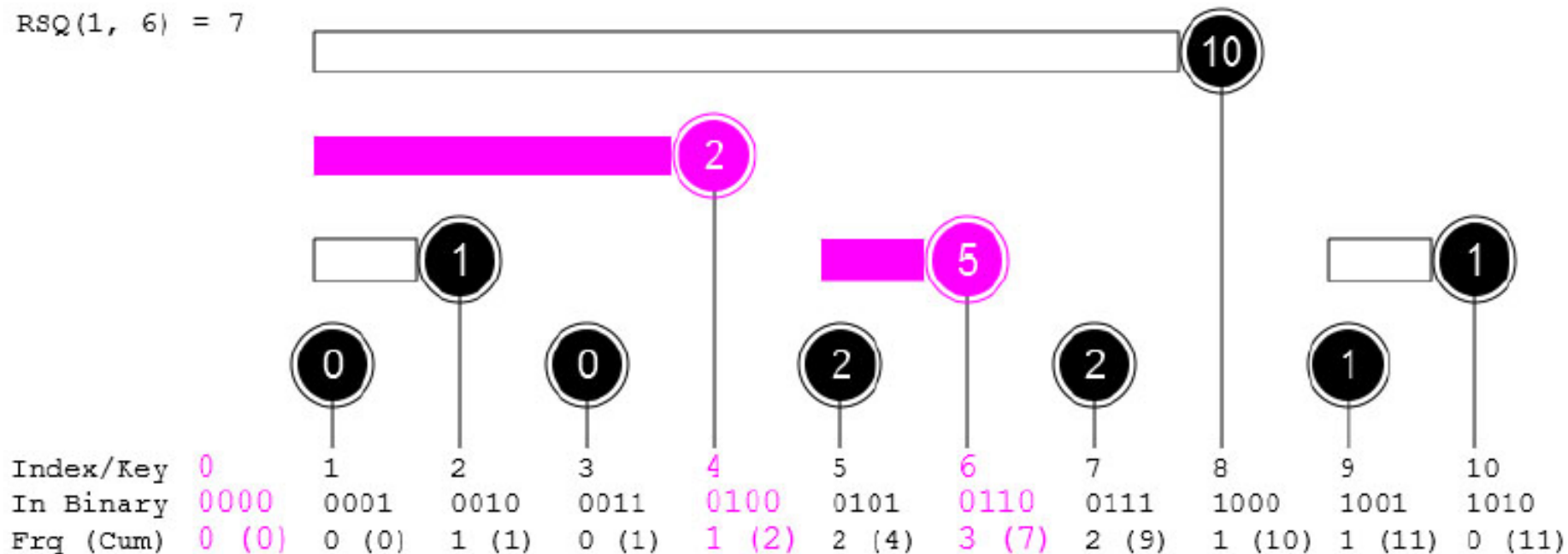
# Implementation

- Let the name of the Fenwick Tree array be  $ft$ :
  - element at index  $i$  is responsible for elements in the range  $[i - \text{LSOne}(i) + 1 .. i]$
  - $ft[i]$  stores the cf of  $\{i - \text{LSOne}(i) + 1, i - \text{LSOne}(i) + 2, i - \text{LSOne}(i) + 3, \dots, i\}$



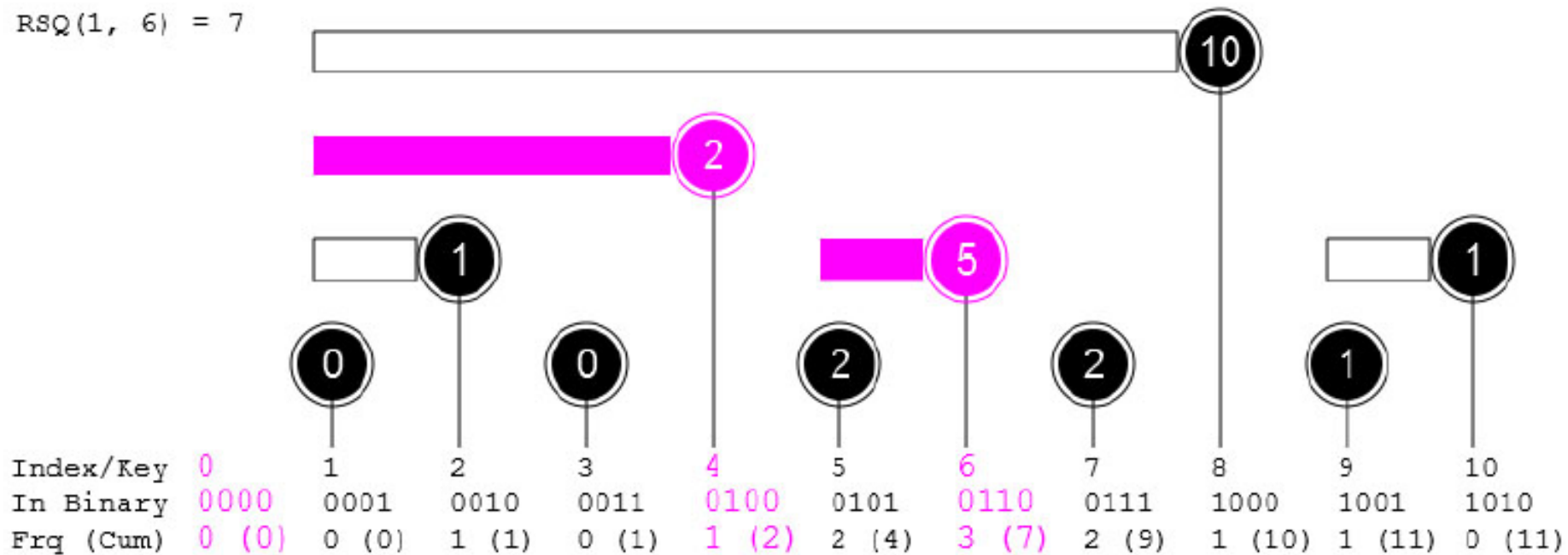
# Implementation

- ft[4] = 2 is responsible for range [4-4+1..4] = [1..4]
- ft[6] = 5 is responsible for range [6-2+1..6] = [5..6]
- ft[7] = 2 is responsible for range [7-1+1..7] = [7..7]
- ft[8] = 10 is responsible for range [8-8+1..8] = [1..8]



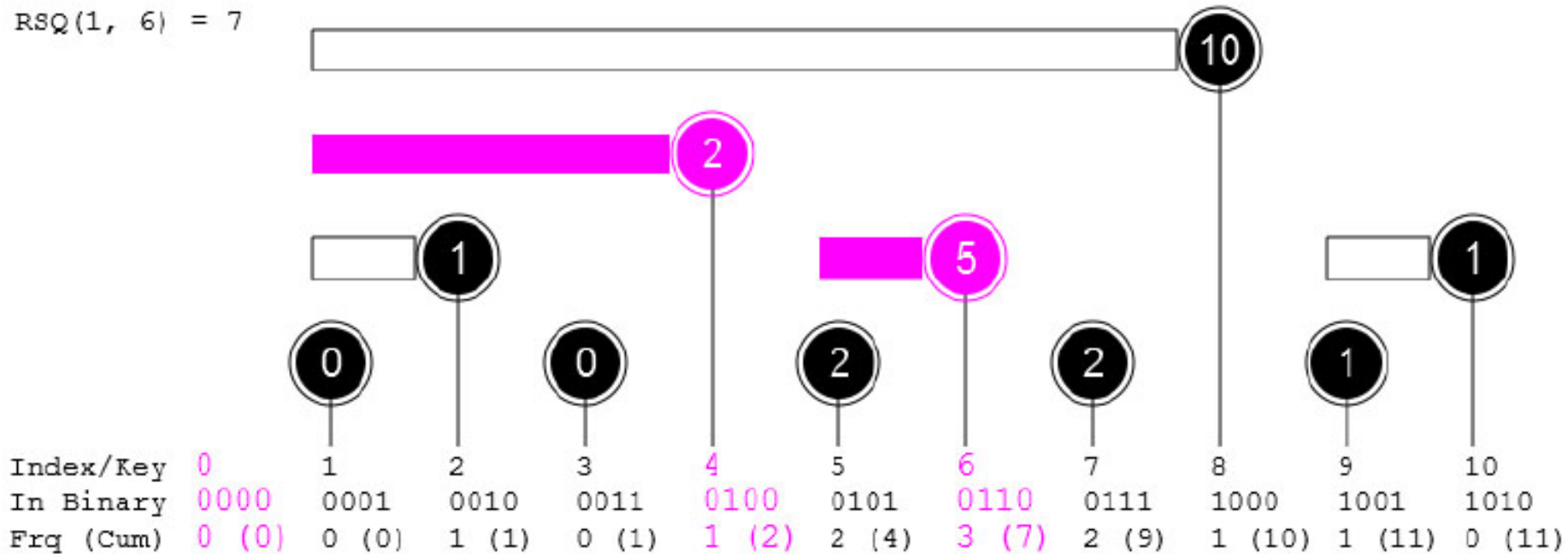
# Implementation

- To obtain the cf between  $[1..b]$ , i.e.  $rsq(b)$ , we simply add  $ft[b]$ ,  $ft[b']$ ,  $ft[b'']$ , ... until index  $b^i$  is 0.
- This sequence of indices is obtained via subtracting the Least Significant One-bit via the bit manipulation expression:  $b' = b - LSOne(b)$ .



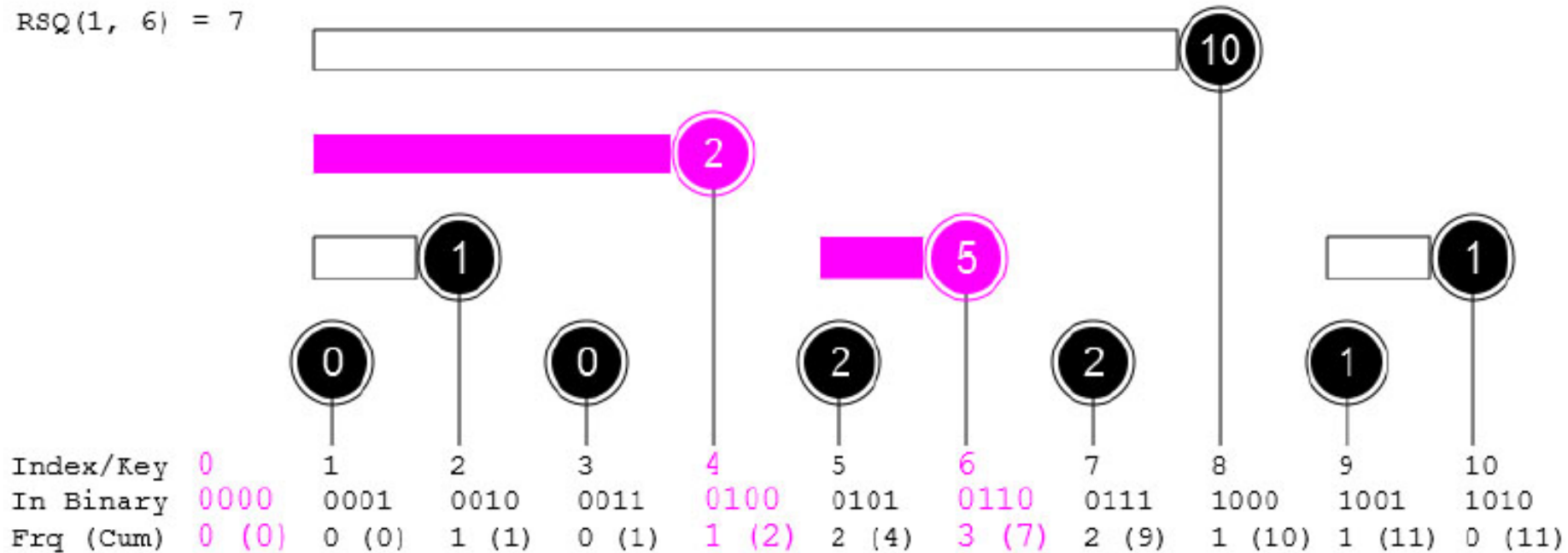
# Implementation

- $\text{rsq}(6) = \text{ft}[6] + \text{ft}[4] = 5 + 2 = 7$ . Notice that indices 4 and 6 are responsible for range  $[1..4]$  and  $[5..6]$ , respectively.
- The indices 6, 4, and 0 are related in their binary form:  $b = 6_{10} = (110)_2$  can be transformed to  $b' = 4_{10} = (100)_2$  and subsequently to  $b'' = 0_{10} = (000)_2$ .



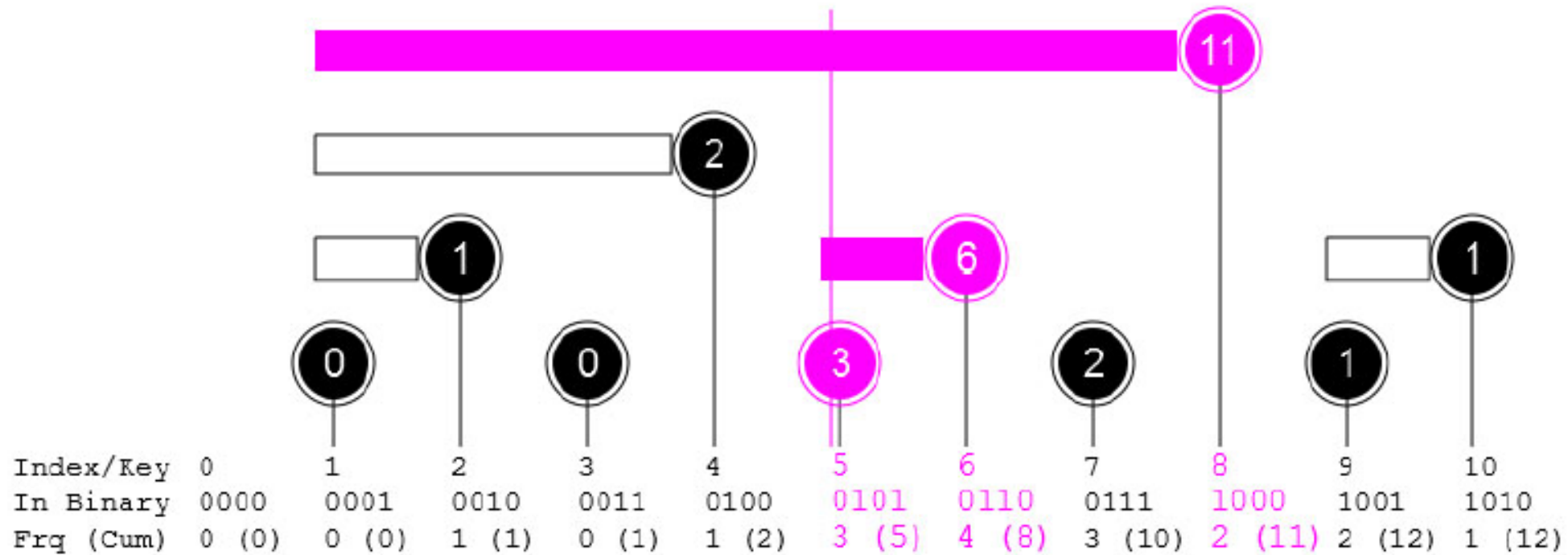
# Implementation

- cf between  $[a..b]$  where  $a \neq 1$  is simple, evaluate  $rsq(a, b) = rsq(b) - rsq(a - 1)$ .
- $rsq(4, 6)$ , we can simply return  $rsq(6) - rsq(3) = (5+2) - (0+1) = 7 - 1 = 6$ .



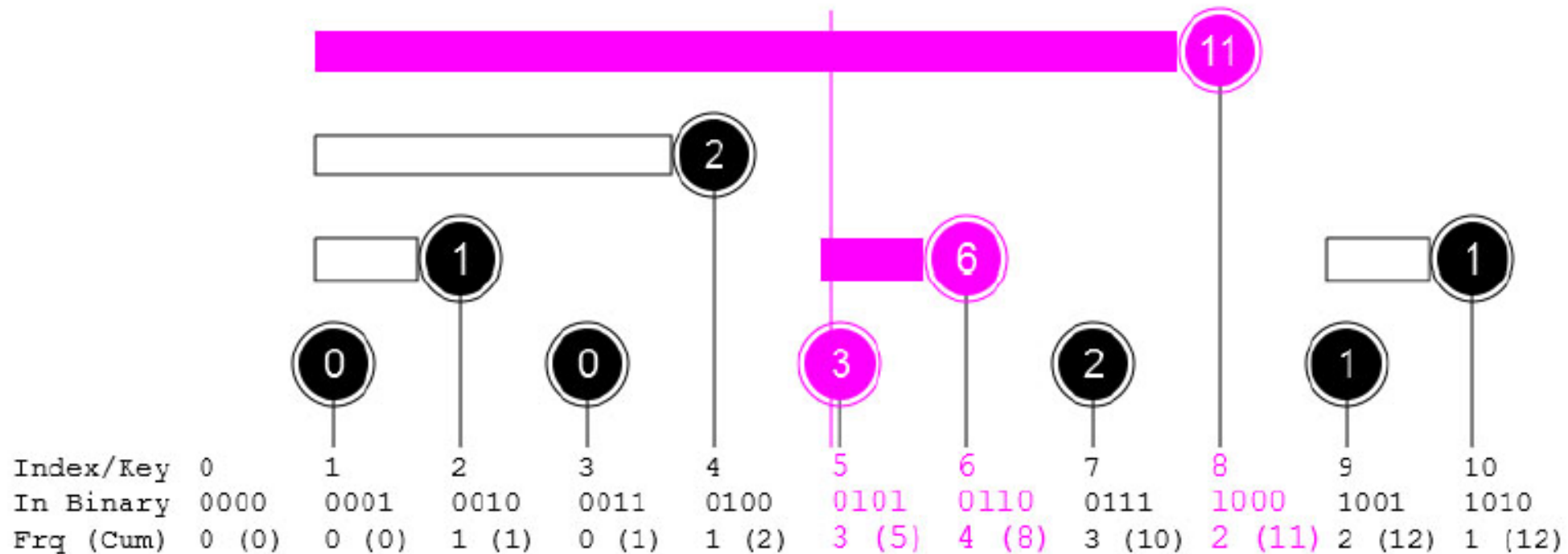
# Implementation

- When updating the value of the element at index  $k$  by adjusting its value by  $v$  (note that  $v$  can be either positive or negative), i.e. calling  $\text{adjust}(k, v)$ , we have to update  $\text{ft}[k], \text{ft}[k'], \text{ft}[k''], \dots$  until index  $k^i$  exceeds  $n$ .  $k' = k + \text{LSOne}(k)$ .



# Implementation

- adjust(5, 1) will affect (add +1 to) ft[k] at indices  $k = 5_{10} = (101)_2$ ,  $k' = (101)_2 + (001)_2 = (110)_2 = 6_{10}$ , and  $k'' = (110)_2 + (010)_2 = (1000)_2 = 8_{10}$  via the expression given above.





# Implementation

```
class FenwickTree {
    private Vector<Integer> ft;

    private int LSOne(int S) { return (S & (-S)); }

    public FenwickTree() {}

    // initialization: n + 1 zeroes, ignore index 0
    public FenwickTree(int n) {
        ft = new Vector<Integer>();
        for (int i = 0; i <= n; i++) ft.add(0);
    }

    public int rsq(int b) { // returns RSQ(1, b)
        int sum = 0; for (; b > 0; b -= LSOne(b)) sum += ft.get(b);
        return sum; }

    public int rsq(int a, int b) { // returns RSQ(a, b)
        return rsq(b) - (a == 1 ? 0 : rsq(a - 1)); }

    // adjusts value of the k-th element by v (v can be +ve/inc or -ve/dec)
    void adjust(int k, int v) { // note: n = ft.size() - 1
        for (; k < (int)ft.size(); k += LSOne(k)) ft.set(k, ft.get(k) + v); }
};
```

# <https://visualgo.net/en/fenwicktree>

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0	0	1	0	2	2	5	2	10	1	1
0	1	0	1	2	3	2	1	1	0	0

slow fast

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