

Q0: Binary Search

1. Problem Understanding

- Binary Search is an algorithm used to efficiently search for a value in a sorted array or in any monotonic search space.
 - It works by repeatedly dividing the search interval into half.
 - It applies to:
 - Searching for an element
 - Finding first/last occurrence
 - Finding boundaries
 - Solving questions where answer lies in a range
 - Questions where the condition behaves monotonically (true/false pattern)
 - Binary search is NOT just for arrays – it's for searching on answers (Binary Search on Answer).
-

2. Constraints

- Array must be sorted (or monotonic condition must exist).
- Monotonic means the values move only in one direction (only increasing or only decreasing). Nothing zig-zag.
- Time limit often requires $O(\log n)$ or $O(n \log n)$ solutions.
- Values may be very large, but binary search handles it easily due to logarithmic complexity.
- **What is a Monotonic Array?**
 - An array is monotonic if it satisfies one of the following:
 - 1. $\square \in \text{Monotonic Increasing}$
 - Each element is greater than or equal to the previous.
 - Example:
 - 1 2 2 3 4 7 9
 - Here:
 - $\text{arr}[i] \leq \text{arr}[i+1]$
 - 2. $\square \in \text{Monotonic Decreasing}$
 - Each element is less than or equal to the previous.
 - Example:
 - 9 7 7 4 3 1
 - Here:
 - $\text{arr}[i] \geq \text{arr}[i+1]$
 - 3. $\square \in \text{Why Binary Search Needs a Monotonic Condition?}$
 - Binary search works only if we can discard half of the search space every time.
 - That is possible only if:
 - Values consistently increase OR

- Values consistently decrease
- So we know which direction contains the target.
- If array is like:
 - 3 5 2 8 1
 - Not sorted
 - Not monotonic
 - Values go up and down â†’ cannot apply binary search

3. Edge Cases

- These are the edge cases that break 80% of beginners' code:
 - $mid = (l + r) / 2$ â†’ may overflow; use
 - $mid = l + (r - l) / 2$.
 - Infinite loop due to wrong update (like using $l = mid$ instead of $l = mid + 1$).
 - Using wrong comparison ($<$ vs $<=$).
 - Arrays with duplicate elements (must choose left/right boundary carefully).
 - l surpassing r (stop condition).
 - Off-by-one errors in returning the boundary.
 - When doing binary search on answers â†’ ensure the predicate is monotonic.

4. Examples

ðŸ§ª Examples (Simple)

Searching 7 in [1,3,4,7,9]
Sorted â†’ can apply binary search.

Finding floor square root of 40 â†’ answer lies in integer range [0..40].

Finding minimum capacity to ship packages â†’ monotonic condition exists.

5. Approaches

Approach 1: Classic Binary Search on Sorted Array

Idea:

- Search for exact element.

Java Code:

```
public static int binarySearch(int[] arr, int target) {
    int l = 0, r = arr.length - 1;
    while (l <= r) {
        int mid = l + (r - l) / 2;
```

```

        if (arr[mid] == target) return mid;
        else if (arr[mid] < target) l = mid + 1;
        else r = mid - 1;
    }
    return -1;
}

```

Intuition Behind the Approach:

- Each comparison eliminates half the array \hat{a} ' logarithmic speed.

Complexity (Time & Space):

- $\hat{a} \pm i$ Time Complexity
 - $O(\log n)$ because we cut search space by half every iteration.
- $\hat{a} \pm i$ Space Complexity
 - $O(1)$ iterative.

Approach 2: Lower Bound (first index $\hat{a} \geq$ target)

Idea:

- Find the first position where element is \geq target.

Java Code:

```

public static int lowerBound(int[] arr, int target) {
    int l = 0, r = arr.length - 1;
    int ans = arr.length;
    while (l <= r) {
        int mid = l + (r - l) / 2;
        if (arr[mid] >= target) {
            ans = mid;
            r = mid - 1;
        } else {
            l = mid + 1;
        }
    }
    return ans;
}

```

Approach 3: Upper Bound (first index $>$ target)

Java Code:

```

public static int upperBound(int[] arr, int target) {
    int l = 0, r = arr.length - 1;
    int ans = arr.length;
    while (l <= r) {

```

```

        int mid = l + (r - l) / 2;
        if (arr[mid] > target) {
            ans = mid;
            r = mid - 1;
        } else {
            l = mid + 1;
        }
    }
    return ans;
}

```

Approach 4: Search Insert Position

Java Code:

```

public static int searchInsert(int[] arr, int target) {
    int l = 0, r = arr.length - 1;
    int ans = arr.length;
    while (l <= r) {
        int mid = l + (r - l) / 2;
        if (arr[mid] >= target) {
            ans = mid;
            r = mid - 1;
        } else {
            l = mid + 1;
        }
    }
    return ans;
}

```

Approach 5: Binary Search on Answer (VERY IMPORTANT)

Idea:

- Used in many DSA problems:
 - "Allocate minimum pages"
 - "Aggressive cows"
 - "Koko eating bananas"
 - "Painters partition"
 - "Minimum capacity to ship packages"
 - "Find smallest divisor"
 - "Minimize maximum distance to gas station"
 - "etc."
- "Idea"
 - Guess an answer → check if it's valid → move left or right.
 - The key is the predicate must be monotonic:
 - If x works → all values > x might work
- Or vice versa

Java Code:

```
public static int searchAnswer(int low, int high) {
    int ans = -1;
    while (low <= high) {
        int mid = low + (high - low) / 2;
        if (isPossible(mid)) {
            ans = mid;        // mid is valid â†’ try left
            high = mid - 1;
        } else {
            low = mid + 1;    // mid invalid â†’ go right
        }
    }
    return ans;
}
```

Approach 6: Binary Search on Real Numbers

Idea:

- Used for:
 - Square root with precision
 - Koko eating bananas (if using double)
 - Gas station placement
- Key difference:
 - Loop until precision achieved instead of $l \leq r$.

Approach 7: Binary Search on Rotated Sorted Array

Idea:

- “Search in rotated sorted array
- “Find minimum in rotated sorted array
- Concept:
- Left half or right half is always sorted â†’ determine where target lies.

Approach 8: Binary Search with Duplicates

Idea:

- First occurrence
- Last occurrence
- Count occurrences using:
 - $\text{last} - \text{first} + 1$
- Requires adjusting boundaries carefully.

6. Variants / Follow-Ups

- Floor
- Ceil
- Square root
- Cube root
- kth missing positive
- Peak element
- Allocate minimum pages
- Maximum average subarray (variant)
- Aggressive cows
- Minimum time to complete tasks
- Koko eating bananas
- Search in mountain array

7. Tips & Observations

- Always check monotonicity before applying binary search.
- For duplicates â†’ use lower/upper bound logic.
- For rotated arrays â†’ identify sorted half.
- For BS on answer â†’ define the range properly.
- Watch for infinite loops by using $mid = l + (r - l) / 2$.
- Always test with edge cases: single element, all equal, target not found, target smaller/larger than all.
- **Complexity**
 - Time Complexity (Final)
 - All array-based BS â†’ $O(\log n)$
 - BS on answer â†’ $O(\log(\text{range}))$
 - Space Complexity (Final)
 - $O(1)$ for iterative.

Q135: Find Minimum in Rotated Sorted Array

1. Problem Understanding

- You are given a sorted array that has been rotated between 1 and n times.
- Examples:
 - Original: [1,2,3,4,5]
 - Rotated: [3,4,5,1,2]
- Your task:

- Find the minimum element in the rotated array.
 - Must run in $O(\log n)$ â†’ Binary Search.
-

2. Constraints

- $1 \leq n \leq 5000$
 - Numbers can be negative.
 - All elements are unique
 - Must use binary search â†’ no linear scan.
-

3. Edge Cases

- Array not rotated â†’ minimum at index 0.
 - Example: [1,2,3,4]
 - Rotation by n times also gives original array.
 - Minimum might be at the boundaries.
 - Example: [2,3,4,5,1]
 - Very small arrays: size 1, size 2.
 - Already sorted means return `nums[0]`.
-

4. Examples

Example 1

Input:

[3,4,5,1,2]

Output:

1

Example 2

Input:

[4,5,6,7,0,1,2]

Output:

0

5. Approaches

Approach 1: Brute Force (Linear Scan)

Idea:

- The minimum of any array is simply the smallest element.
- Directly scan the array.

Intuition Behind the Approach:

- A sorted rotated array still contains a minimum somewhere.
- You inspect all elements to find it.

Complexity (Time & Space):

- Time Complexity
 - $O(n)$
 - Because scanning all elements.
- Space Complexity
 - $O(1)$
 - Only one variable used.

Approach 2: Slightly Better (Check Adjacent Break Point)

Idea:

- In a rotated array, the minimum occurs at the break point where:
- $\text{nums}[i] < \text{nums}[i-1]$
- Scan and check only such break.

Java Code:

```
public static int findMin(int[] nums) {
    int n = nums.length;
    for (int i = 1; i < n; i++) {
        if (nums[i] < nums[i - 1]) return nums[i];
    }
    return nums[0];
}
```

Intuition Behind the Approach:

- The array is sorted except one pivot point.
- Minimum sits right after the point where order breaks.

Complexity (Time & Space):

- Time Complexity
 - $O(n)$ worst case
 - If array not rotated.
- Space Complexity
 - $O(1)$

- $O(1)$

Approach 3: Binary Search " $O(\log n)$

Idea:

- Use binary search to find the pivot where rotation happened.
- Observations:
 - If $\text{nums}[\text{mid}] > \text{nums}[\text{right}]$, the minimum is to the right.
 - Else " minimum is to the left or at mid.

Steps:

- Set $l = 0, r = n-1$.
- While $l < r$:
 - Compute mid.
 - If $\text{nums}[\text{mid}] > \text{nums}[\text{r}]$ " search right half.
 - Else " search left half (including mid).
- At the end $l = \text{index of minimum}$.

Java Code:

```
public static int findMin(int[] nums) {
    int l = 0, r = nums.length - 1;

    while (l < r) {
        int mid = l + (r - l) / 2;

        if (nums[mid] > nums[r]) {
            l = mid + 1; // min is to the right
        } else {
            r = mid; // min is at mid or left side
        }
    }

    return nums[l]; // l is the index of minimum
}
```

" Intuition Behind the Approach:

- When $\text{nums}[\text{mid}] > \text{nums}[\text{r}]$, right half is unsorted " min lies there.
- When $\text{nums}[\text{mid}] \leq \text{nums}[\text{r}]$, right half is sorted " min cannot be in sorted region except possibly mid.
- Binary search narrows search space towards the pivot point.
- This is a classic Binary Search on answer space, not on presence.

Complexity (Time & Space):

- " Time Complexity
 - $O(\log n)$

- Because each iteration halves the search space.
 - Space Complexity
 - $O(1)$
 - Only two pointers.
-

6. Justification / Proof of Optimality

- The array is partially sorted.
 - Binary search is the perfect fit when:
 - There is a monotonic pattern.
 - There is a pivot.
 - We can eliminate half of the array on each decision.
 - The condition $nums[mid] > nums[r]$ is the key that tells us which side the pivot lies on.
-

7. Variants / Follow-Ups

- Find index of minimum instead of value.
 - Find number of times rotated → index of min is the rotation count.
 - Rotated array with duplicates → trickier binary search.
 - Search element in rotated sorted array.
-

8. Tips & Observations

- Always compare with the rightmost element for easier logic.
 - If whole array is already sorted → return $nums[0]$.
 - If duplicates exist:
 - Need to shrink boundaries carefully.
 - Rotated sorted array problems ALWAYS revolve around:
 - Detecting pivot
 - Searching on monotonic segments
-

Q136: Square root of a number

1. Problem Understanding

- You are given a number x .
 - You must return:
 - If x is a perfect square → its exact square root
 - Otherwise → $\text{floor}(\sqrt{x})$
 - You must not use the built-in `sqrt()` function.
 - Time complexity must be $O(\log N)$, which hints binary search.
-

2. Constraints

- $1 \leq x \leq 10^7$
 - Input is a single integer
 - Output must be an integer (floor value)
-

3. Edge Cases

- $x = 1$ → answer = 1
 - Very small values: $x = 2, 3$ → answer = 1
 - Large inputs up to 10^7
 - Perfect squares: 4, 9, 16, 25 → return exact root
 - Overflow of $mid * mid$ → use long to prevent overflow
-

4. Examples

Example 1

Input: 5

Output: 2

Example 2

Input: 36

Output: 6

Example 3

Input: 2

Output: 1

5. Approaches

Approach 1: Brute Force Approach (Linear Search)

Idea:

- Try every number from 1 to x and check the last i such that $i*i \leq x$.

Steps:

- Loop from $i=1$ to $i*i \leq x$.
- Track the last valid number.
- Return it.

Java Code:

```
public static int mySqrt(int x) {  
    int ans = 0;
```

```

    for (long i = 1; i * i <= x; i++) {
        ans = (int) i;
    }
    return ans;
}

```

Intuition Behind the Approach:

- Directly simulate what square root means.
- Simple but slow.

Complexity (Time & Space):

- Time Complexity
 - $O(\sqrt{x})$
 - Because loop runs until $i*i \leq x$.
- Space Complexity
 - $O(1)$
 - Only counters used.

Approach 2: Using Math Properties (Decreasing Search Range)

Idea:

- Instead of checking all numbers from 1 to x , we can optimize by stopping early once $i*i$ crosses x .
- (Slightly better than brute but still worst-case \sqrt{x})

Steps:

- Same as brute but break early.

Java Code:

```

public static int mySqrt(int x) {
    long i = 1;
    while (i * i <= x) i++;
    return (int)(i - 1);
}

```

Intuition Behind the Approach:

- Stop as soon as you've gone past the root.
- Still linear in \sqrt{x} but less work than brute.

Complexity (Time & Space):

- Time Complexity
 - $O(\sqrt{x})$
- Space Complexity
 - $O(1)$

Approach 3: Binary Search (Required $O(\log N)$)

Idea:

- The real square root lies between 1 and x.
- Binary search this space for the greatest mid such that:
- $\text{mid} * \text{mid} \leq x$

Steps:

- Let low = 1, high = x
- Compute mid
- If $\text{mid} * \text{mid} \leq x$, store it and move right (search for bigger)
- Else move left
- Return last stored answer

Java Code:

```
public static int mySqrt(int x) {  
    if (x == 0) return 0;  
  
    long s = 1;  
    long e = x;  
    long ans = 1;  
  
    while (s <= e) {  
        long mid = (s + e) / 2;  
  
        if (mid * mid <= x) {  
            ans = mid;        // mid is valid  
            s = mid + 1;      // try for bigger  
        } else {  
            e = mid - 1;      // mid too big  
        }  
    }  
    return (int) ans;  
}
```

Intuition Behind the Approach:

- Square root grows slowly, so binary search reduces the search space by half every step.
- We want the largest number whose square is $\leq x$ therefore move right on valid mid.
- This ensures correctness for both perfect and non-perfect squares.

Complexity (Time & Space):

- Time Complexity
 - $O(\log N)$
 - Because binary search repeatedly halves the range.
- Space Complexity

- $O(1)$
- Only variables used.

6. Justification / Proof of Optimality

- Binary search is optimal because the search range is monotonic ($i \cdot i$ increases as i increases).
- Unlike brute force ($\hat{\sim} N$ steps), binary search does it extremely fast in $\log N$ steps.
- Works well for values up to 10^7 easily within constraints.

7. Variants / Follow-Ups

- Calculate ceil of $\hat{\sim} x$
- Return true/false if x is a perfect square
- Return $\hat{\sim} x$ without using multiplication (Newton's Method)
- $\hat{\sim} x$ for long or big integers

8. Tips & Observations

- Always use long for $\text{mid} * \text{mid}$ to avoid integer overflow.
- When searching for largest valid, use:
 - if $(\text{mid} * \text{mid} \leq x)$ $\hat{+}$ move right
- For a perfect square, binary search will naturally find it.
- Classic question in binary search category $\hat{=}$ must be mastered.

Q137: Search A 2D Matrix

1. Problem Understanding

- You are given a matrix where:
 - Each row is sorted from left $\hat{+}$ right.
 - The first element of each row is greater than the last element of previous row.
 - This means the entire matrix behaves like one sorted array.
- Goal:
 - Search if x exists.
 - Return true / false.

2. Constraints

- $1 \leq m, n \leq 1000$
 - $m * n \hat{\leq} 10^6$ $\hat{+}$ binary search is perfect.
 - Values range from -10^4 to 10^4 .
-

3. Edge Cases

- x smaller than smallest element → false
 - x larger than largest element → false
 - 1×1 matrix
 - Single row or single column
 - Large matrix → must use $O(\log(m*n))$
-

4. Examples

```
Example 1
Matrix:
1  3  5  7
10 11 16 20
23 30 34 60
x = 10 → true
```

```
Example 2
x = 12 → false
```

5. Approaches

Approach 1: Brute Force (Check Every Element)

Idea:

- Scan all rows and all columns until element is found.

Java Code:

```
public static boolean searchMatrix(int[][] mat, int x) {
    int m = mat.length, n = mat[0].length;

    for (int i = 0; i < m; i++) {
        for (int j = 0; j < n; j++) {
            if (mat[i][j] == x) return true;
        }
    }
    return false;
}
```

Complexity (Time & Space):

- Time Complexity
 - $O(m*n)$ → checks every cell
 - Why: No skipping possible.

- $\frac{3}{4}$ Space Complexity
 - $O(1)$ "no extra space"

Approach 2: Row Selection + Binary Search (Better)

Idea:

- Identify which row x could be in:
 - Check first/last elements of each row.
- Binary search that row.

Java Code:

```
public static boolean searchMatrix(int[][] mat, int x) {
    int m = mat.length, n = mat[0].length;

    for (int i = 0; i < m; i++) {
        if (x >= mat[i][0] && x <= mat[i][n-1]) {
            int l = 0, r = n-1;
            while (l <= r) {
                int mid = l + (r - l) / 2;
                if (mat[i][mid] == x) return true;
                else if (mat[i][mid] < x) l = mid + 1;
                else r = mid - 1;
            }
            return false;
        }
    }
    return false;
}
```

Intuition Behind the Approach:

- Each row is sorted, so we can binary search inside the correct row.

Complexity (Time & Space):

- \pm Time Complexity
 - Worst-case: $O(m + \log n)$
 - Why: Scan rows then binary search inside one row.
- $\frac{3}{4}$ Space Complexity
 - $O(1)$

Approach 3: Treat as Sorted 1D Array (Optimal " $O(\log(m*n))$)

Idea:

- Because:
 - Last element of row i < first element of row $i+1$
 - Matrix behaves like a single sorted array of length $m*n$.

- Map 1D index to matrix index:
 - $\text{row} = \text{mid} / n$
 - $\text{col} = \text{mid} \% n$
- Binary search on 1D indexed search space.

Java Code:

```
public static boolean searchMatrix(int[][] mat, int x) {
    int m = mat.length, n = mat[0].length;

    int l = 0, r = m * n - 1;

    while (l <= r) {
        int mid = l + (r - l) / 2;

        int row = mid / n;
        int col = mid % n;

        if (mat[row][col] == x) return true;
        else if (mat[row][col] < x) l = mid + 1;
        else r = mid - 1;
    }

    return false;
}
```

Intuition Behind the Approach:

- You're performing binary search on a flattened version of the matrix.
- Because matrix is globally sorted, this works flawlessly.

Complexity (Time & Space):

- Time Complexity
 - $O(\log(m*n))$
 - Why: Classic binary search on total number of elements.
- Space Complexity
 - $O(1)$ "constant space."

6. Justification / Proof of Optimality

- Brute force is too slow for large $m*n$.
- Row-wise selection reduces search but still touches many rows.
- Flattened binary search guarantees minimal comparisons and utilizes full sorted property "best approach."

7. Variants / Follow-Ups

- Search in a matrix where each row AND column is sorted (different problem, different approach).
 - Find the number of occurrences of x.
 - Return position instead of boolean.
-

8. Tips & Observations

- Whenever a matrix has this row-start > previous-row-end pattern $\hat{+}$ treat like 1D sorted list.
 - Binary search remains the strongest pattern if global monotonicity exists.
-

Q138: Find First and Last Position of Element in Sorted Array

1. Problem Understanding

- Given a sorted (non-decreasing) array nums, find the first and last index of a given target.
 - If target doesn't appear, return [-1, -1].
 - We must do it in $O(\log n)$, so binary search is required.
-

2. Constraints

- Array length: 0 to $1e5$
 - Values range: $-1e9$ to $1e9$
 - Array is sorted
 - Must achieve $O(\log n)$
-

3. Edge Cases

- $n = 0 \hat{+}$ return [-1, -1]
 - Target is smaller than first element
 - Target is greater than last element
 - All elements are the same as target $\hat{+}$ output [0, n-1]
 - Target appears once
 - Target appears multiple times
-

4. Examples

Example 1

Input

6 8

5 7 7 8 8 10

Output

3 4

Example 2

Input

6 6

5 7 7 8 8 10

Output

-1 -1

5. Approaches

Approach 1: Brute Force (Linear Scan)

Idea:

- Iterate whole array once to find first occurrence and last occurrence.

Steps:

- Loop from left to right â€ˆ find first match
- Loop from right to left â€ˆ find last match

Java Code:

```
public int[] searchRange(int[] nums, int target) {  
    int first = -1, last = -1;  
  
    for (int i = 0; i < nums.length; i++) {  
        if (nums[i] == target) {  
            first = i;  
            break;  
        }  
    }  
  
    for (int i = nums.length - 1; i >= 0; i--) {  
        if (nums[i] == target) {  
            last = i;  
            break;  
        }  
    }  
}
```

```
        return new int[]{first, last};
    }
```

Intuition Behind the Approach:

- We check the entire array because it's unspecified where target may occur.

Complexity (Time & Space):

- Time Complexity
 - $O(n)$ because we scan the array twice.
- Space Complexity
 - $O(1)$ because no extra space.

Approach 2: Two Binary Searches (Optimal)

Idea:

- Use binary search twice:
 - once to find the first occurrence
 - once to find the last occurrence
- Because array is sorted, binary search ensures $O(\log n)$.

Steps:

- Steps
- Find first position
 - Normal binary search
 - When you find `nums[mid] == target`, move left to see if earlier occurrence exists
 - So set `high = mid - 1`
- Find last position
 - When `nums[mid] == target`, move right
 - So set `low = mid + 1`
- Return `[first, last]`

Java Code:

```
public int[] searchRange(int[] nums, int target) {
    int first = findFirst(nums, target);
    int last = findLast(nums, target);
    return new int[]{first, last};
}

private int findFirst(int[] nums, int target) {
    int low = 0, high = nums.length - 1, ans = -1;
    while (low <= high) {
        int mid = low + (high - low) / 2;

        if (nums[mid] == target) {
            ans = mid;
        }
    }
    return ans;
}
```

```

        high = mid - 1; // go left
    } else if (nums[mid] < target) {
        low = mid + 1;
    } else {
        high = mid - 1;
    }
}
return ans;
}

private int findLast(int[] nums, int target) {
    int low = 0, high = nums.length - 1, ans = -1;
    while (low <= high) {
        int mid = low + (high - low) / 2;

        if (nums[mid] == target) {
            ans = mid;
            low = mid + 1; // go right
        } else if (nums[mid] < target) {
            low = mid + 1;
        } else {
            high = mid - 1;
        }
    }
    return ans;
}

```

Intuition Behind the Approach:

- Binary search normally stops when the target is found, but we modify it:
 - For first occurrence, we force the search to continue left until no earlier index contains the target.
 - For last occurrence, we force search right.
- This preserves the $O(\log n)$ speed but still finds boundaries.

Complexity (Time & Space):

- $\hat{O}(\log n)$ Time Complexity
 - Two binary searches $\hat{O}(\log n)$
 - Why?
 - Each binary search halves the array repeatedly, so two searches still $\log n + \log n = \log n$.
- $\hat{O}(1)$ Space Complexity
 - $O(1)$
 - Why?
 - Only variables used; no extra data structures.

6. Justification / Proof of Optimality

- Binary search ensures fastest possible runtime ($O(\log n)$), and by slightly modifying conditions, we can locate the exact boundaries of the target.

7. Variants / Follow-Ups

- Find first occurrence only
 - Find last occurrence only
 - Count occurrences $\hat{+}$ last - first + 1
 - Works for rotated sorted arrays with modifications
-

8. Tips & Observations

- Always prefer binary search if array is sorted and you need positions.
 - Use the mid calculation trick:
 - $\text{mid} = \text{low} + (\text{high} - \text{low}) / 2$ to avoid overflow.
 - Searching for first $\hat{+}$ move left
 - Searching for last $\hat{+}$ move right
-

Q139: Count 1 in sorted binary array

1. Problem Understanding

- You are given a binary, sorted in non-increasing order array:
 - 1 1 1 ... 1 0 0 0
 - You must find the count of 1s.
 - Because all 1s come first and all 0s later, the problem becomes finding the boundary between 1 $\hat{+}$ 0.
 - Binary search helps find this boundary in $O(\log n)$.
-

2. Constraints

- $1 \leq N \leq 10^6$
 - Array contains only 0 and 1
 - Array is sorted in non-increasing order
 - Must be efficient (binary search recommended)
-

3. Edge Cases

- All 1s $\hat{+}$ count = N
 - All 0s $\hat{+}$ count = 0
 - Only one element
 - Transition happens at index 0
 - Transition happens at index N-1
-

4. Examples

Example 1

Input

8
1 1 1 1 1 0 0 0

Output

5

Example 2

Input

4
1 1 1 1

Output

4

5. Approaches

Approach 1: Linear Scan (Brute Force)

Idea:

- Traverse from the left and count until you see the first 0.

Steps:

- Initialize count = 0
- Loop from left to right
- If element is 1, increment
- If element is 0, break immediately
- Return count

Java Code:

```
public int countOnes(int[] arr) {  
    int count = 0;  
    for (int x : arr) {  
        if (x == 1) count++;  
        else break;  
    }  
}
```

```
    return count;
}
```

Complexity (Time & Space):

- Time Complexity
 - $O(k)$ where k = number of initial 1s
 - Worst case $O(n)$
 - Why?
 - You may traverse the entire array if all values are 1.
- Space Complexity
 - $O(1)$
 - Why?
 - No extra memory used.

Approach 2: Binary Search for First 0 (Better)

Idea:

- Count of 1s = index of the first 0.
- Example:
- [1 1 1 0 0] \rightarrow first 0 at index 3 \rightarrow count = 3.

Steps:

- Binary search to find the first index where $arr[mid] == 0$.
- If no 0 exists \rightarrow return N.
- Else return that index.

Java Code:

```
public int countOnes(int[] arr) {
    int low = 0, high = arr.length - 1;
    int firstZero = arr.length;

    while (low <= high) {
        int mid = low + (high - low) / 2;

        if (arr[mid] == 0) {
            firstZero = mid;
            high = mid - 1;
        } else {
            low = mid + 1;
        }
    }

    return firstZero;
}
```

ðŸ’ Intuition Behind the Approach:

- The first 0 divides the array into:
- [1s] [0s]
- Finding that boundary gives us the count of 1s.

Complexity (Time & Space):

- Time Complexity
 - $O(\log n)$
 - Because each step halves the search space.
- Space Complexity
 - $O(1)$
 - Only a few variables used.

Approach 3: Binary Search for Last 1 (Optimal)

Idea:

- Find last index where `arr[mid] == 1`, then `count = mid + 1`

Steps:

- Steps
- Binary search over array.
- If `arr[mid] == 1`, record index and move right to find last 1.
- If `arr[mid] == 0`, move left.
- At end â†’
 - if `last1 = -1`, return 0
 - else return `last1 + 1`

Java Code:

```
public int countOnes(int[] arr) {
    int low = 0, high = arr.length - 1;
    int lastOne = -1;

    while (low <= high) {
        int mid = low + (high - low) / 2;

        if (arr[mid] == 1) {
            lastOne = mid;
            low = mid + 1; // search right
        } else {
            high = mid - 1;
        }
    }

    return lastOne + 1;
}
```

Intuition Behind the Approach:

- The array looks like this:
 - 1 1 1 1 | 0 0 0
- Binary search targets the rightmost 1 efficiently.

Complexity (Time & Space):

- Time Complexity
 - $O(\log n)$
 - Boundary search using binary search.
 - Space Complexity
 - $O(1)$
-

6. Justification / Proof of Optimality

- Binary search works perfectly because the array has a monotonic pattern: all 1s first, then all 0s.
 - Finding the transition boundary is the fastest way to determine count.
-

7. Variants / Follow-Ups

- Find first 1
 - Find first 0 in increasing binary array
 - Count zeros
 - First and last occurrence of any target
 - Works in monotonic boolean functions
 - Find first element greater than X
-

8. Tips & Observations

- Always look for monotonicity to apply binary search.
 - For non-increasing binary arrays:
 - First 0 $\hat{=}$ count of 1s
 - Last 1 $\hat{=}$ count of 1s
 - Binary search boundary problems always need careful condition handling.
-

Q140: Floor in a Sorted Array

1. Problem Understanding

- You are given a sorted array without duplicates and a value x.
- You must find the floor of x:
 - Floor of x = largest element \leq x.
- Return its index (0-based).

- If it doesn't exist return -1.
-

2. Constraints

- $1 \leq N \leq 1e5$
 - Array sorted strictly increasing
 - $0 \leq x \leq arr[N-1]$
-

3. Edge Cases

- $x < \text{smallest element}$ return -1
 - $x > \text{largest element}$ return index of last element
 - Exact match exists return index
 - Single-element array
 - Missing element but floor exists (example: $x=5$, $arr=[1,2,8]$)
-

4. Examples

Example 1

Input

```
7 0
1 2 8 10 11 12 19
```

Output

```
-1
```

Example 2

Input

```
7 5
1 2 8 10 11 12 19
```

Output

```
1
```

Example 3

Input

```
7 10
1 2 8 10 11 12 19
```

Output

3

5. Approaches

Approach 1: Linear Scan (Brute Force)

Idea:

- Scan from left until elements exceed x.

Steps:

- Loop through array
- Track last index with $\text{arr}[i] \leq x$
- Return index

Java Code:

```
public int floorIndex(int[] arr, int x) {  
    int ans = -1;  
    for (int i = 0; i < arr.length; i++) {  
        if (arr[i] <= x) ans = i;  
        else break;  
    }  
    return ans;  
}
```

Intuition Behind the Approach:

- Since array is sorted, once you exceed x, you won't find a valid floor later.

Complexity (Time & Space):

- Time Complexity
 - $O(n)$
 - Might scan whole array.
- Space Complexity
 - $O(1)$

Approach 2: Binary Search (Optimal)

Idea:

- Use binary search to find last element $\leq x$.

Steps:

- Initialize low=0, high=n-1, ans=-1
- While low ≤ high:
 - Compute mid
 - If arr[mid] ≤ x:
 - Record ans = mid
 - Go right → low = mid + 1
 - Else (arr[mid] > x):
 - Go left → high = mid - 1
- Return ans

Java Code:

```
public int floorIndex(int[] arr, int x) {
    int low = 0, high = arr.length - 1;
    int ans = -1;

    while (low <= high) {
        int mid = low + (high - low) / 2;

        if (arr[mid] <= x) {
            ans = mid;
            low = mid + 1;
        } else {
            high = mid - 1;
        }
    }

    return ans;
}
```

Intuition Behind the Approach:

- This is a boundary search:
- We want the rightmost value that is still ≤ x.
- Binary search lets us jump over large sections to find this efficiently.

Complexity (Time & Space):

- Time Complexity
 - $O(\log n)$
 - Binary search halves the space each step.
- Space Complexity
- $O(1)$
- Constant space.

6. Justification / Proof of Optimality

- Binary search is ideal because the array is sorted and we need to find a boundary position.
 - This ensures the best efficiency ($O(\log n)$).
-

7. Variants / Follow-Ups

- Find ceil of x (smallest $\geq x$)
 - Find first element greater than x
 - Lower bound / upper bound problems
 - Find last occurrence of a number
-

8. Tips & Observations

- Floor \rightarrow go right when $\text{arr}[\text{mid}] \leq x$
 - Ceil \rightarrow go left when $\text{arr}[\text{mid}] \geq x$
 - Boundary problems always use modified binary search
 - Always store ans before moving
-

Q141: Rotated Sorted Array Search

1. Problem Understanding

- You are given an array that was originally sorted in non-decreasing order but then rotated at some pivot. Example:
 - Original: 0 1 2 4 5 6 7
 - Rotated : 4 5 6 7 0 1 2
 - Your task:
 - Find the index of target B in this rotated sorted array.
 - If not present \rightarrow return -1.
 - No duplicates exist.
 - Goal: $O(\log n)$ \rightarrow must use binary search.
-

2. Constraints

- $1 \leq N \leq 10^6$
 - $1 \leq A[i] \leq 10^9$
 - Array was sorted then rotated
 - All elements are unique
 - Must achieve $O(\log n)$
-

3. Edge Cases

- Target is at pivot

- Array is not rotated at all (normal sorted)
 - Target smaller than all elements
 - Target larger than all elements
 - Single-element array
 - Pivot at index 0 or last index
 - Target at boundaries
-

4. Examples

Example 1

Input

```
8
4 5 6 7 0 1 2 3
4
```

Output

```
0
```

Example 2

Input

```
4
5 17 100 3
6
```

Output

```
-1
```

5. Approaches

Approach 1: Linear Scan (Brute Force)

Idea:

- Loop from 0 to N-1
- If element equals target $\hat{=}$ return index
- Else return -1

Approach 2: Find Pivot then Binary Search Both Halves (Better)

Idea:

- In a rotated sorted array, the smallest element is the pivot.
- Example:
 - 4 5 6 7 0 1 2 → pivot is at index 4.
- Steps:
 - Find pivot using binary search
 - Decide whether target is in left side or right side
 - Apply binary search normally in that half

Steps:

- Binary search to find smallest element index (pivot).
- If target lies in interval [arr[pivot], arr[n-1]] → search in right half.
- Else search in left half.

Java Code:

```
public int search(int[] arr, int target) {
    int n = arr.length;

    // Step 1: find pivot (index of smallest element)
    int low = 0, high = n - 1;
    while (low < high) {
        int mid = low + (high - low) / 2;
        if (arr[mid] > arr[high]) low = mid + 1;
        else high = mid;
    }
    int pivot = low;

    // Step 2: decide which half to search
    if (target >= arr[pivot] && target <= arr[n - 1]) {
        return binarySearch(arr, pivot, n - 1, target);
    } else {
        return binarySearch(arr, 0, pivot - 1, target);
    }
}

private int binarySearch(int[] arr, int low, int high, int target) {
    while (low <= high) {
        int mid = low + (high - low) / 2;
        if (arr[mid] == target) return mid;
        else if (arr[mid] < target) low = mid + 1;
        else high = mid - 1;
    }
    return -1;
}
```

Intuition Behind the Approach:

- A rotated sorted array is two sorted halves.
- Find pivot → pick correct half → normal binary search.

Complexity (Time & Space):

- Time Complexity
 - Pivot search: $O(\log n)$
 - Binary search in correct half: $O(\log n)$
 - Total: $O(\log n)$
- Space Complexity
 - $O(1)$

Approach 3: Single-Pass Binary Search (Optimal)

Idea:

- We modify binary search logic itself.
- At any mid, one side is always sorted:
- Either:
 - Left half is sorted
 - OR right half is sorted
- We check which half is sorted and decide where to move.

Steps:

- Standard binary search loop
- If `arr[mid] == target` return mid
- Check which half is sorted
 - If left half sorted (`arr[low] ≤ arr[mid]`):
 - Check if target lies in left sorted range
 - move right or left accordingly
 - Else right half sorted:
 - Check if target lies in right sorted range
 - move accordingly

Java Code:

```
public int search(int[] arr, int target) {
    int low = 0, high = arr.length - 1;

    while (low <= high) {
        int mid = low + (high - low) / 2;

        if (arr[mid] == target) return mid;

        // Left half sorted
        if (arr[low] <= arr[mid]) {
            if (target >= arr[low] && target < arr[mid])
                high = mid - 1;
            else
                low = mid + 1;
        }
        // Right half sorted
    }
}
```

```

        else {
            if (target > arr[mid] && target <= arr[high])
                low = mid + 1;
            else
                high = mid - 1;
        }
    }

    return -1;
}

```

Intuition Behind the Approach:

- Every mid splits the array into two halves, and one half is guaranteed to be sorted.
- By checking:
 - whether target lies inside that sorted half
 - or in the unsorted half
- we discard half of the array each time â classic binary search.
- This avoids finding pivot separately.

Complexity (Time & Space):

- Time Complexity
 - $O(\log n)$
 - We halve the search space every iteration.
- Space Complexity
 - $O(1)$

6. Justification / Proof of Optimality

- Binary search works perfectly here because even after rotation, one part of the array remains sorted at every division.
- This property ensures we can always make the correct decision about which half to search.

7. Variants / Follow-Ups

- Search element in rotated array with duplicates
- Search min element in rotated array
- Search max element in rotated array
- Count rotations
- Find pivot index
- Find peak element

8. Tips & Observations

- At every mid, one half is sorted â the key insight
- If $arr[low] \leq arr[mid]$ â left sorted

- Else \hat{t} right sorted
 - Use ranges carefully when comparing target
 - Avoid infinite loops by updating low and high properly
-

Q142: Peak Index in a Mountain Array

1. Problem Understanding

- You are given a mountain array, meaning:
 - It strictly increases up to a peak
 - Then strictly decreases
 - Example:
 - 0 2 5 7 6 3 1 * \hat{t} * peak
 - You must find the index of the peak element.
 - Mountain array guarantees:
 - Exactly one peak
 - Peak is not at index 0 or index $n-1$
 - Must solve in $O(\log n)$
-

2. Constraints

- $3 \leq n \leq 1e5$
 - Values: $0 \leq arr[i] \leq 1e6$
 - Array is guaranteed to be a perfect mountain
 - Must solve in $O(\log n)$
-

3. Edge Cases

- (not many because mountain is guaranteed)
 - Peak at index 1 (smallest valid peak)
 - Peak at index $n-2$ (largest valid peak)
 - Large values
 - Minimum length (3 elements)
-

4. Examples

Example 1

Input

3
0 1 0

Output

1

Example 2

Input

4

0 2 1 0

Output

1

Example 3

Input

4

0 10 5 2

Output

1

5. Approaches

Approach 1: Linear Scan (Brute Force)

Idea:

- Scan until the numbers stop increasing.
- The first time you see a drop, the previous index is peak.

Steps:

- Loop from $i = 1$ to $n-1$
- If $arr[i] < arr[i-1]$
 - \hat{a} ' return $i-1$
- Because guaranteed mountain \hat{a} ' no other checks needed.

Java Code:

```
public int peakIndex(int[] arr) {  
    for (int i = 1; i < arr.length; i++) {  
        if (arr[i] < arr[i - 1]) return i - 1;  
    }  
}
```

```

    }
    return -1;
}

```

Intuition Behind the Approach:

- Peak is exactly where sequence changes from increasing to decreasing.

Complexity (Time & Space):

- Time Complexity
 - $O(n)$
- Space Complexity
 - $O(1)$

Approach 2: Binary Search (Optimal)

Idea:

- Binary search for the point where slope changes:
 - If $arr[mid] < arr[mid + 1]$ we are on the increasing slope
 - Peak is on the right
 - move low = mid + 1
 - If $arr[mid] > arr[mid + 1]$ we are on the decreasing slope
 - mid could be the peak
 - move high = mid
- Eventually low == high peak index.

Steps:

- Initialize low = 0, high = n-1
- While low < high:
 - Compute mid
 - If increasing part move right
 - If decreasing part move left
- Return low (same as high)

Java Code:

```

public int peakIndex(int[] arr) {
    int low = 0, high = arr.length - 1;

    while (low < high) {
        int mid = low + (high - low) / 2;

        if (arr[mid] < arr[mid + 1]) {
            // Increasing part: peak is on the right
            low = mid + 1;
        } else {
            // Decreasing part: peak is mid or left side

```

```

        high = mid;
    }
}

return low; // or high, both same
}

```

Intuition Behind the Approach:

- The mountain array has a monotonic behavior:
 - Increasing → Peak → Decreasing
- At any mid:
 - If $arr[mid] < arr[mid+1]$, slope is increasing → move right
 - If $arr[mid] > arr[mid+1]$, slope is decreasing → peak is left or mid
- We narrow down the peak using binary search.

Complexity (Time & Space):

- Time Complexity
 - $O(\log n)$
 - We cut the search space in half each step.
- Space Complexity
 - $O(1)$

6. Justification / Proof of Optimality

- Binary search works because the array is monotonic on each side of the peak:
 - left half strictly increasing
 - right half strictly decreasing
- This allows you to determine the direction of the peak at every step, ensuring $O(\log n)$ time.

7. Variants / Follow-Ups

- Find peak in mountain array (LeetCode 852)
- Find peak element in unsorted array (LeetCode 162)
- Find max in bitonic array
- Find first element greater than previous
- Find turning point in unimodal array

8. Tips & Observations

- Use $mid < mid+1$ to detect slope
- Always check $arr[mid] < arr[mid+1]$ FIRST
- low will always converge to the peak
- Never check neighbors outside bounds because mountain is guaranteed