

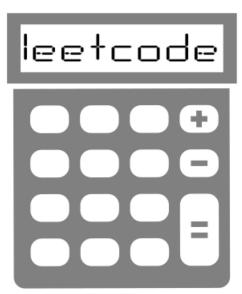
## Approach 1: Pocket Calculator Algorithm

Before going to the serious stuff, let's first have some fun and implement the algorithm used by the pocket calculators

Usually a pocket calculator computes well exponential functions and natural logarithms by having logarithm tables hardcoded or by the other means. Hence the idea is to reduce the square root computation to these two algorithms as well

$$\sqrt{x} = e^{\frac{1}{2}\log x}$$

That's some sort of cheat because of non-elementary function usage but it's how that actually works in a real life.



### Implementation

```
Java Python

1 class Solution {
    public int mySqrt(int x) {
        if (x < 2) return x;
        int left = (int)Math.pow(Math.E, 0.5 * Math.log(x));
        int right = left + 1;
        return (long)right * right > x ? left : right;
        }
    }
}
```

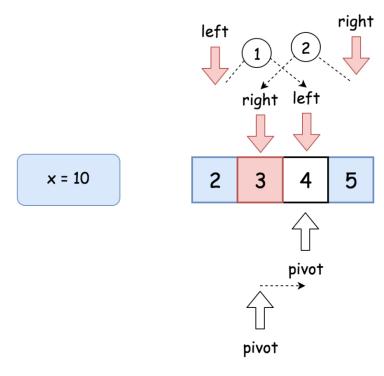
# **Complexity Analysis**

- Time complexity : \(\mu(\pm\).
- Space complexity :  $\mathcal{O}(1)$ .

# Approach 2: Binary Search

#### Intuition

Let's go back to the interview context. For  $x \ge 2$  the square root is always smaller than x/2 and larger than 0:0 < a < x/2. Since a is an integer, the problem goes down to the iteration over the sorted set of integer numbers. Here the binary search enters the scene.



#### Algorithm

- If x < 2, return x.
- $\bullet~$  Set the left boundary to 2, and the right boundary to x / 2.
- While left <= right:
  - $\circ~$  Take num = (left + right) / 2 as a guess. Compute num \* num and compare it with x:
    - If num \* num > x, move the right boundary right = pivot -1
    - Else, if num \* num < x, move the left boundary left = pivot + 1
    - Otherwise num \* num == x, the integer square root is here, let's return it
- Return right

# Implementation

```
Java Python

| class Solution {
| public int mysqrt(int x) {
| if (x < 2) return x;
| 4 |
| long num;
| 6 | int pivot, left = 2, right = x / 2;
| while (left < right) {
| pivot = left + (right - left) / 2;
| num = (long)pivot * pivot;
| if (num > x) right = pivot - 1;
| else if (num < x) left = pivot + 1;
| else return pivot;
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```

# Complexity Analysis

• Time complexity :  $\mathcal{O}(\log N)$ .

Let's compute time complexity with the help of master theorem  $T(N) = aT\left(\frac{N}{b}\right) + \Theta(N^d)$ . The equation represents dividing the problem up into a subproblems of size  $\frac{N}{b}$  in  $\Theta(N^d)$  time. Here at step there is only one subproblem a=1, its size is a half of the initial problem b=2, and all this happens in a constant time d=0. That means that  $\log_b a = d$  and hence were dealing with case 2 that results in  $\mathcal{O}(n^{\log_b a} \log^{d+1} N) = \mathcal{O}(\log N)$  time complexity.

• Space complexity :  $\mathcal{O}(1)$ .

### Approach 3: Recursion + Bit Shifts

### Intuitio

Let's use recursion. Bases cases are  $\sqrt{x}=x$  for x<2. Now the idea is to decrease x recursively at each step to go down to the base cases.

```
How to go down?
```

For example, let's notice that  $\sqrt{x}=2 imes\sqrt{\frac{x}{4}}$ , and hence square root could be computed recursively as

```
\operatorname{mySqrt}(x) = 2 \times \operatorname{mySqrt}\left(\tfrac{x}{4}\right)
```

One could already stop here, but let's use left and right shifts, which are quite fast manipulations with bits

```
x << y
       that means x \times 2^y
x >> y that means
```

 $\frac{x}{2^y}$ 

```
\mathrm{mySqrt}(x) = \mathrm{mySqrt}(x >> 2) << 1
```

That means one could rewrite the recursion above as

in order to fasten up the computations.

#### Implementation

```
Java Python
                                                                                                                                                                                                                                                                                                                     Copy
 class Solution {
  public int mySqrt(int x) {
  if (x < 2) return x;
}</pre>
              int left = mySqrt(x >> 2) << 1;
int right = left + 1;
return (long)right * right > x ? left : right;
 8 }
```

#### **Complexity Analysis**

• Time complexity :  $\mathcal{O}(\log N)$ .

Let's compute time complexity with the help of master theorem  $T(N) = aT\left(\frac{N}{b}\right) + \Theta(N^d)$ . The equation represents dividing the problem up into a subproblems of size  $\frac{N}{b}$  in  $\Theta(N^d)$  time. Here at step there is only one subproblem a = 1, its size is a half of the initial problem b = 2, and all this happens in a constant time d = 0. That means that  $\log_b a = d$  and hence we're dealing with case 2 that results in  $\mathcal{O}(n^{\log_b a}\log^{d+1}N) = \mathcal{O}(\log N)$  time complexity.

ullet Space complexity :  $\mathcal{O}(\log N)$  to keep the recursion stack.

# Approach 4: Newton's Method

#### Intuition

One of the best and widely used methods to compute sqrt is Newton's Method. Here we'll implement the version without the seed trimming to keep things simple. However, seed trimming is a bit of math and lot of fun, so here is a link if you'd like to dive in.

Let's keep the mathematical proofs outside of the article and just use the textbook fact that the set

```
x_{k+1} = \frac{1}{2} \left[ x_k + \frac{x}{x_k} \right]
```

converges to  $\sqrt{x}$  if  $x_0=x$ . Then the things are straightforward: define that error should be less than 1 and proceed iteratively.

### Implementation

```
Copy
Java Python
 class Solution {
public int mySqrt(int x) {
   if (x < 2) return x;
}</pre>
                 double x0 = x;

double x1 = (x0 + x / x0) / 2.0;

while (Math.abs(x0 - x1) >= 1) {

x0 = x1;

x1 = (x0 + x / x0) / 2.0;
                 return (int)x1;
```

### **Complexity Analysis**

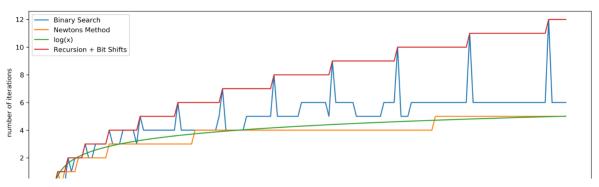
- ullet Time complexity :  $\mathcal{O}(\log N)$  since the set converges quadratically.
- Space complexity :  $\mathcal{O}(1)$ .

# Compare Approaches 2, 3 and 4

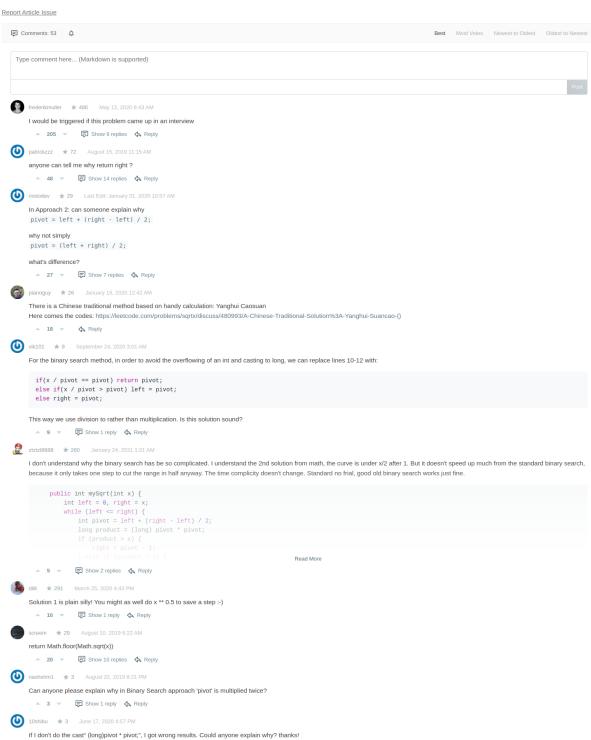
Here we have three algorithms with a time performance  $\mathcal{O}(\log N)$ , and it's a bit confusing.

```
Which one is performing less iterations?
```

Let's run tests for the range of x in order to check that. Here are the results. The best one is Newton's method.



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( 1 2 3 4 5 6 **)** 

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