

MCD4710

Introduction to algorithms
and programming

Lecture 12

Decrease and Conquer

COMMONWEALTH OF AUSTRALIA

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Objectives

Objectives of this lecture are to:

1. Know to search efficiently in ordered sequence (**Binary Search**)
2. Understand design paradigm **decrease-and-conquer** and recognise situations with **logarithmic complexity**
3. Demonstrate the time complexity of **Euclid's algorithm**

This covers learning outcomes:

- 2 – choose and implement appropriate **problem solving strategies**
- 5 – determine the **computational cost** and limitations of algorithms

Overview

1. The Ordered Search Problem
2. Binary Search
3. Revisiting Euclid's Algorithm

Search in Ordered Sequence

Gertrudis Atkinson 0463935372

Kiley Basinger 0411484152

Romana Brose 0418721183

Shayne Brotherton 0436242684

Calandra Clifton 0479753034

Roy Dupuis 0445778949

Leticia Fukushima 0436756947

Cherlyn Gayles 0483503919

...



Problem: find (the position of) a name in a phone book.

Search in Ordered Sequence

ato.gov.au 180.149.195.3
cancer.org.au 52.187.229.23
facebook.com 31.13.71.36
google.com 172.217.12.142
monash.edu 43.245.43.30
newscientist.com 45.60.19.101
news.com.au 23.221.48.198
wikipedia.org 208.80.154.224
...



Problem: find URL in DNS records.

Searching Algorithms

- Given a list of data, find the index of a particular value or return that the value is not present
- Linear search
 - start at first item
 - is it the one I am looking for?
 - if not, go to next item
 - repeat until target is found or all items checked
- If the items are not sorted or cannot be sorted, this approach is necessary

Sequential Search

- Given a target value.
- Find the first item in a list, L , which has the value target.
- If target is found then return the index of the item.
- If target not found then return -1.

Linear Search

- Ex. Linear search for 33.

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

Linear Search

- Ex. Linear search for 33.

	13	14	25	33	43	51	53	64	72	84	93	95	96	97
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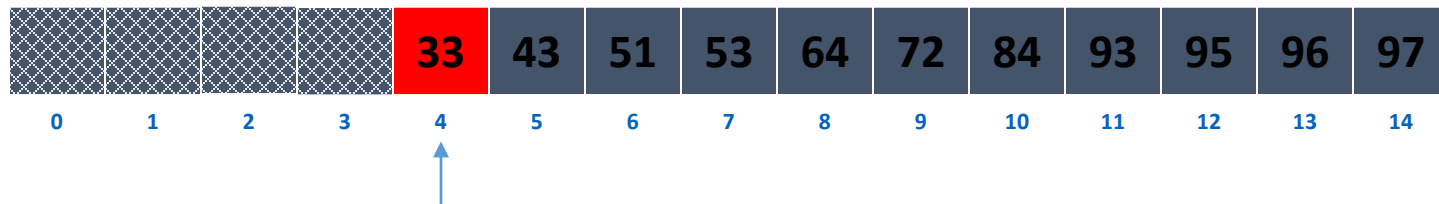
Linear Search

- Ex. Linear search for 33.

				33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

Linear Search

- Ex. Linear search for 33.



Return index: 4

Linear Search – Best Case


- The best case would be if we are looking for an item and find it on our first compare.
- In this example, that means our target would be 6.

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

Linear Search – Best Case

- Ex. Linear search for 6.
- We find the target immediately

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14



Return index: 0

Linear Search – Worst Case

- The worst case would be if we are looking for an item that's not in the list.
- In this example, if we are looking for the item 8 we would have to search through the entire list.

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

Linear Search – Worst Case

- Ex. Linear search for 8 (item not in list).

	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

Linear Search – Worst Case

- Ex. Linear search for 8 (item not in list).

		14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

Linear Search – Worst Case

- Ex. Linear search for 8 (item not in list).

			25	33	43	51	53	64	72	84	93	95	96	97
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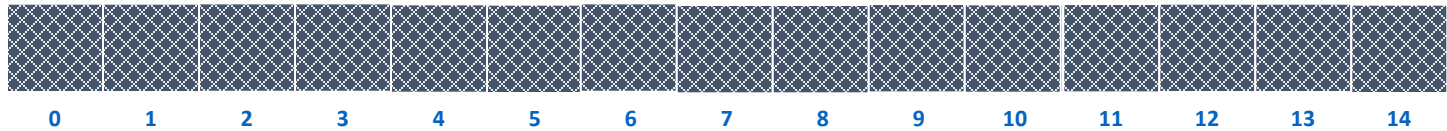
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- Ex. Linear search for 8 (item not in list).

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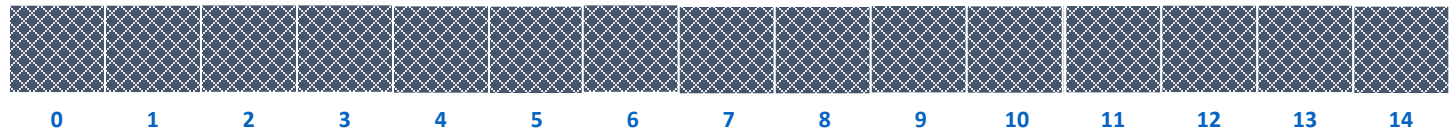
Linear Search – Worst Case

- Ex. Linear search for 8 (item not in list).



Linear Search – Worst Case

- Ex. Linear search for 8 (item not in list).
- This means that for a list of size n , it would take us n compares to find out that the target is not in the list.



Linear Search implementation

```
def linearSearch(aList, target):  
    # Returns index if target is in aList, -1 if it is not  
    for index in range(len(aList)):  
        if target == aList[index]:  
            return index  
    return -1  
  
testlist = [0, 1, 2, 18, 13, 17, 19, 32, 42, 15]  
print(linearSearch(testlist, 13))
```

Invariant : At the i^{th} iteration, $array[0..i-1]$ consists of elements that are not equal to *target*.

What is the worst case time complexity of linear search?

- A. $O(1)$
- B. $O(n)$
- C. $O(n^2)$
- D. $O(\log(n))$

1. Visit <https://flux.qa>
2. Log in using your Authcate details (not required if you're already logged in to Monash)
3. Touch the + symbol and enter the code: UF7BD9
4. Answer questions when they pop up.

We can do better than
that!

Binary Search

The list must be sorted

Decrease-and-Conquer: reduce problem to smaller subproblem

- If items are sorted then we can *decrease and conquer*
- decreasing the search space in half with each step
 - generally a good thing
- Binary Search
 - Start at middle of list
 - is that the item we are looking for?
 - If not, is it less than or greater than the item?
 - less than, move to second half of list
 - greater than, move to first half of list
 - repeat until found or sub list size = 0

• Binary Search

- Start at middle of list
- is that the item we are looking for?
- If not, is it less than or greater than the item?
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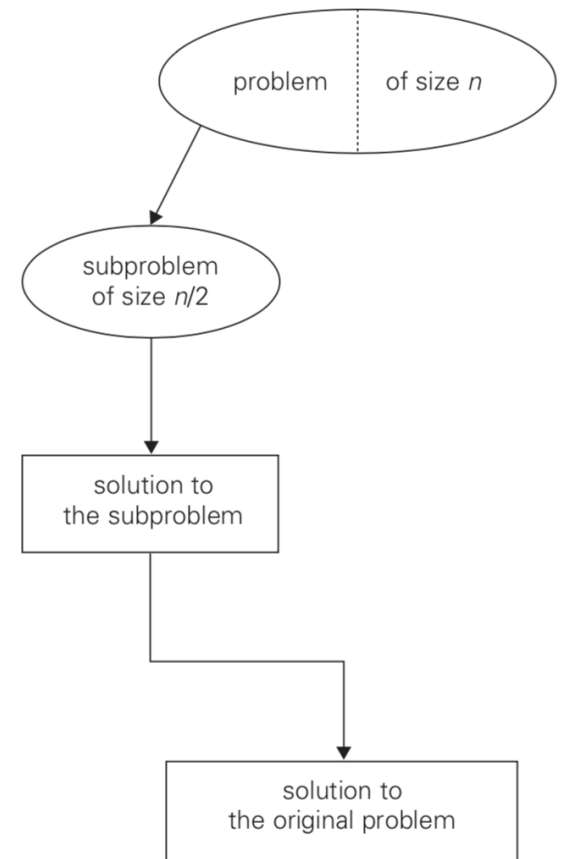


illustration: [Levitin, p. 133]

Binary Search

- Ex. Binary search for 33.

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
↑														↑
lo														hi

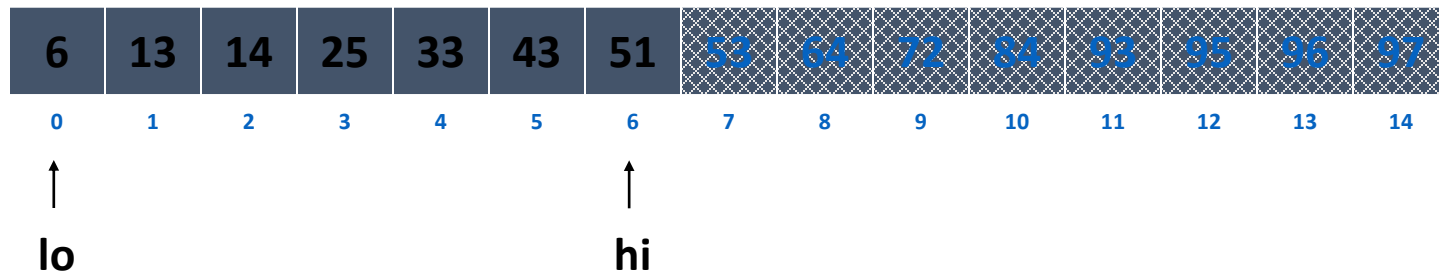
Binary Search

- Ex. Binary search for 33.



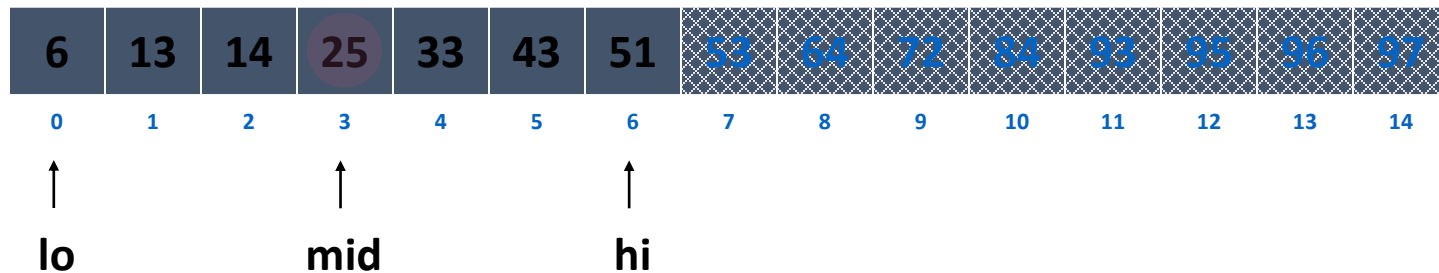
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Binary Search

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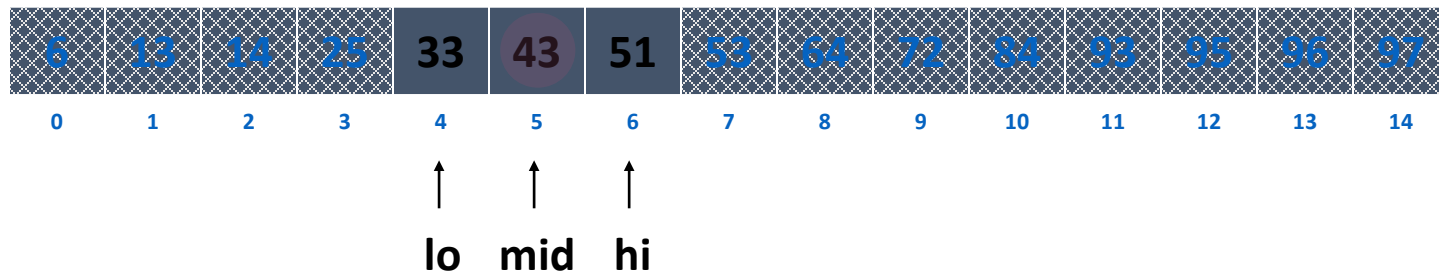
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				lo		hi								

Binary Search

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Binary Search

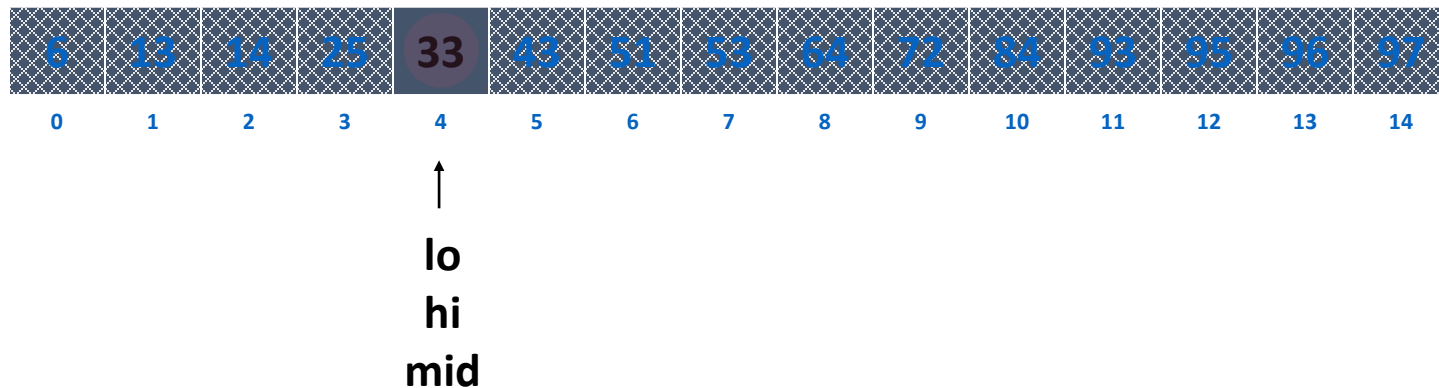
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6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

↑
lo
hi

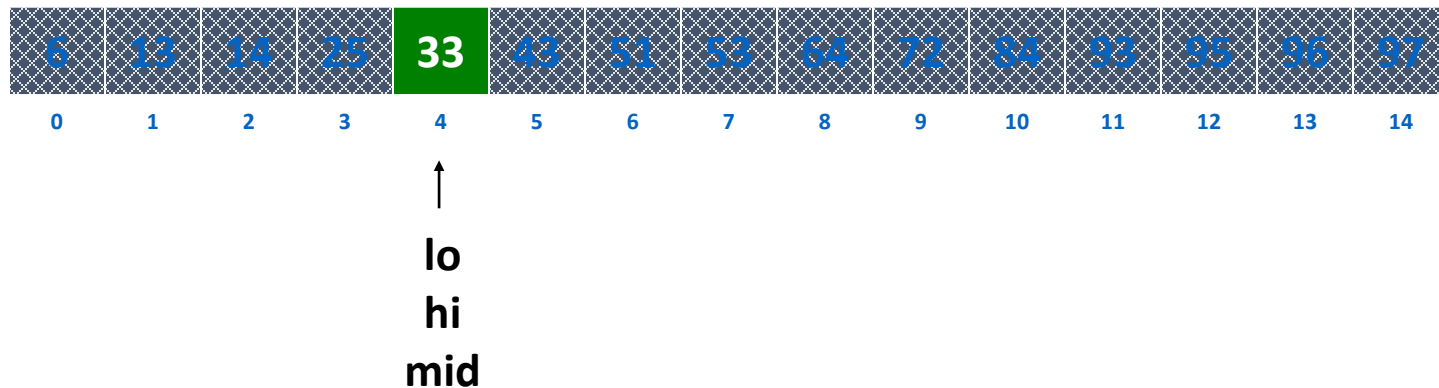
Binary Search

- Ex. Binary search for 33.



Binary Search

- Ex. Binary search for 33.



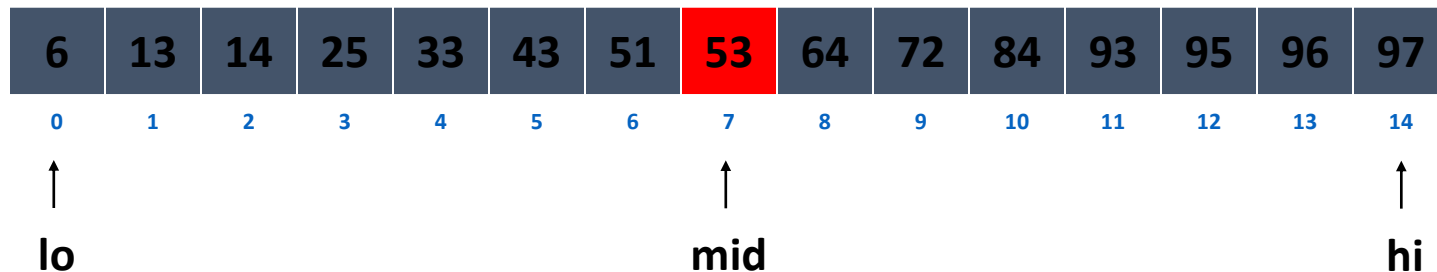
Binary Search – Best Case

- Just like for linear search, the first item we compare is our target.
- In this case, 53.

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
↑														↑
lo														hi

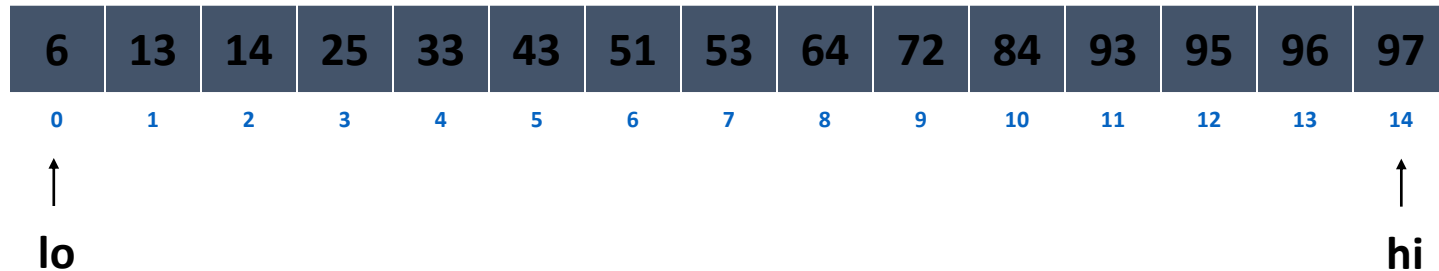
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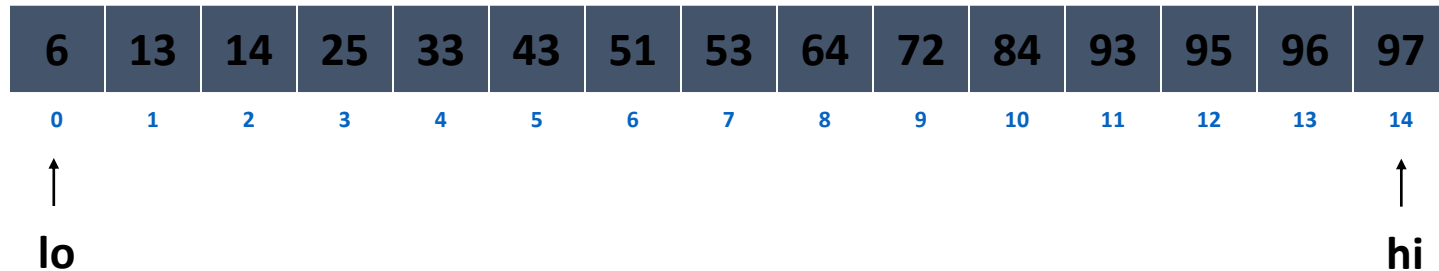
Binary Search – Worst Case

- Just like for linear search, the worst case would be if the item we are looking for is not in the list (e.g. 12)



Binary Search – Worst Case

- Unlike linear search, in binary search we would not have to search through the entire list to find out that our target is not in the list.



Binary Search – Worst Case

- Ex. Binary search for 34.

6	13	14	25	33	43	51	53	64	72	84	93	95	96	97
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
↑														↑
lo														hi

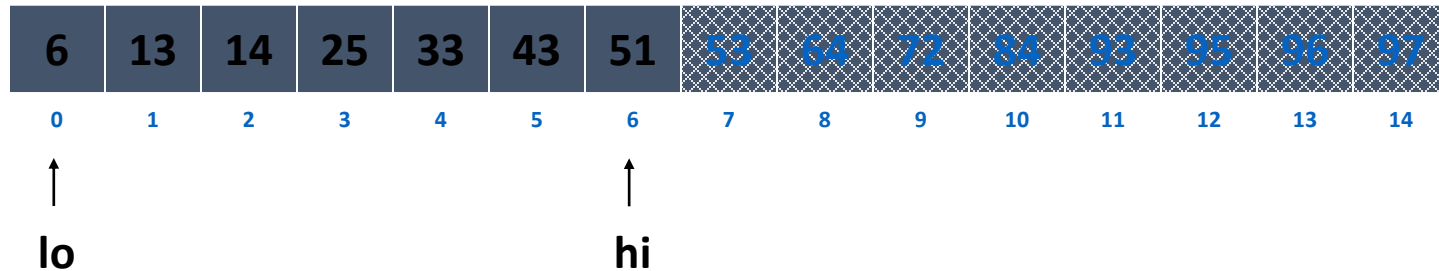
Binary Search – Worst Case

- Ex. Binary search for 34.



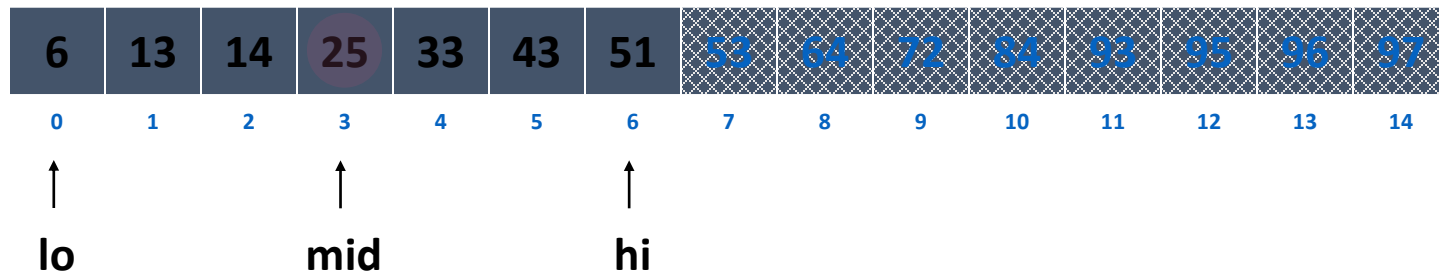
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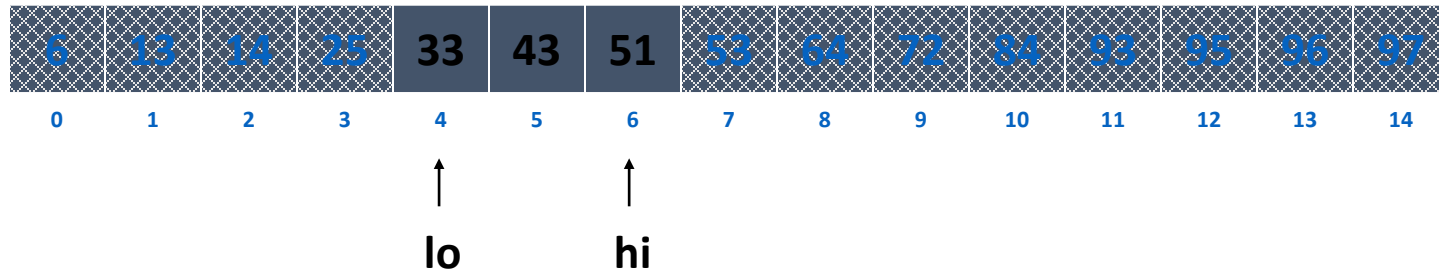
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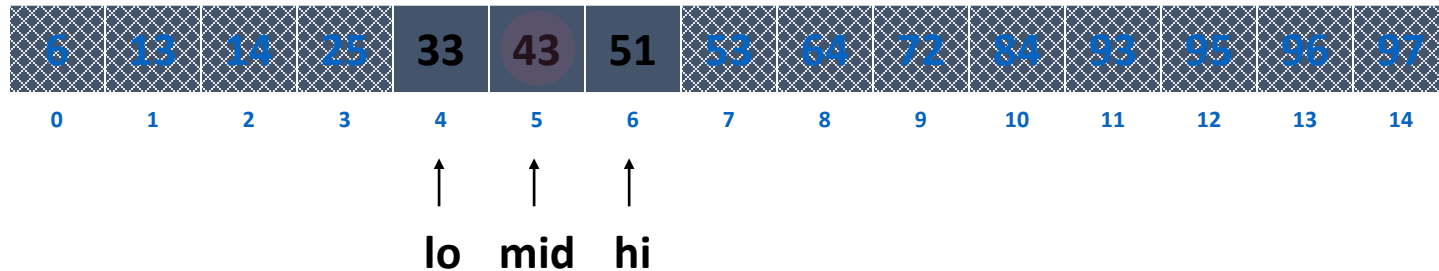
Binary Search – Worst Case

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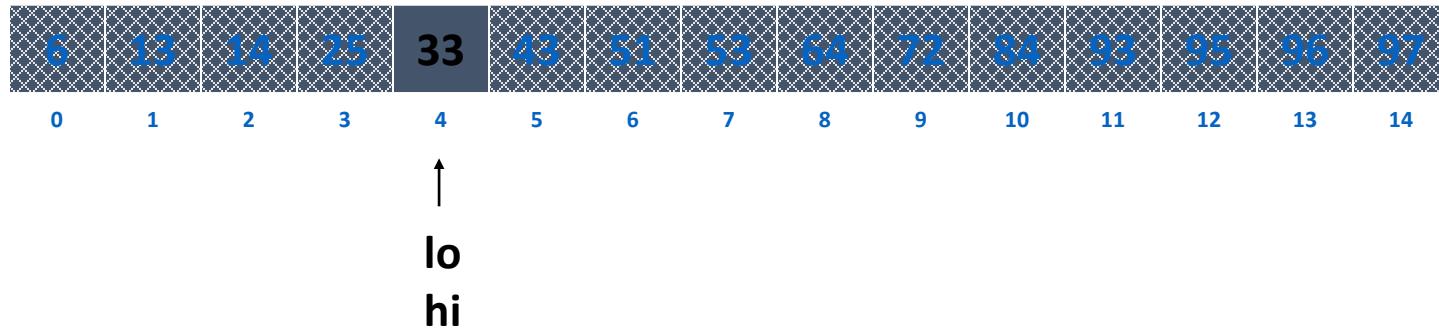
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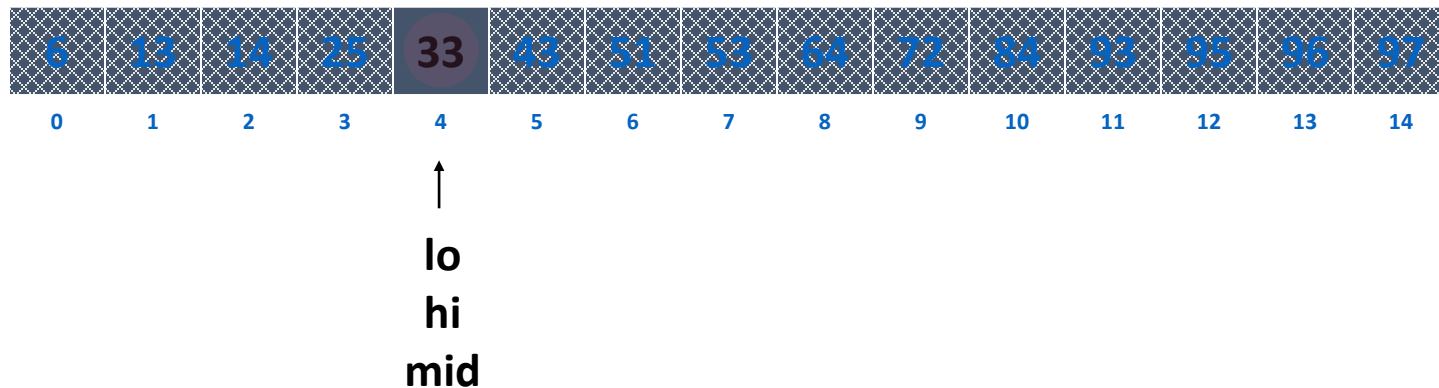
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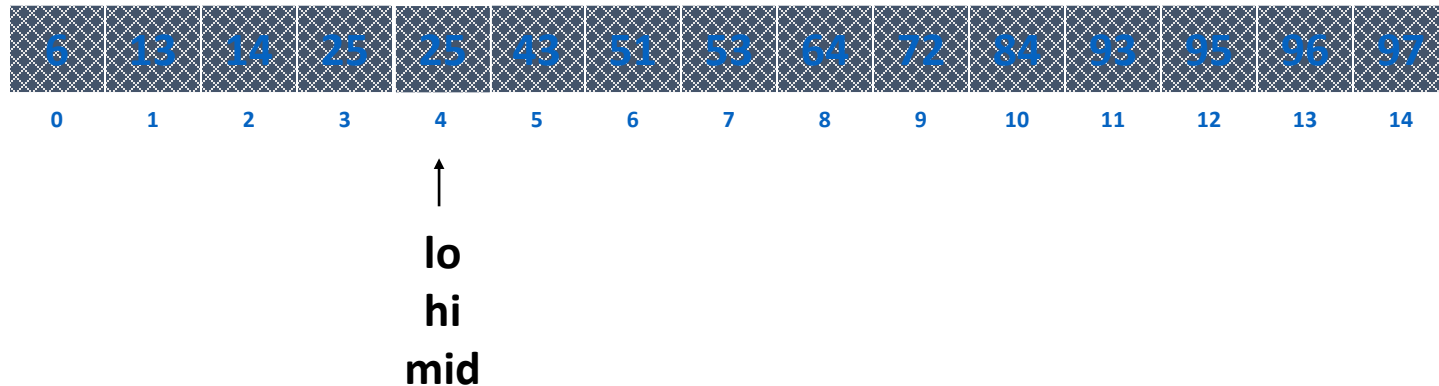
Binary Search – Worst Case

- Ex. Binary search for 34.



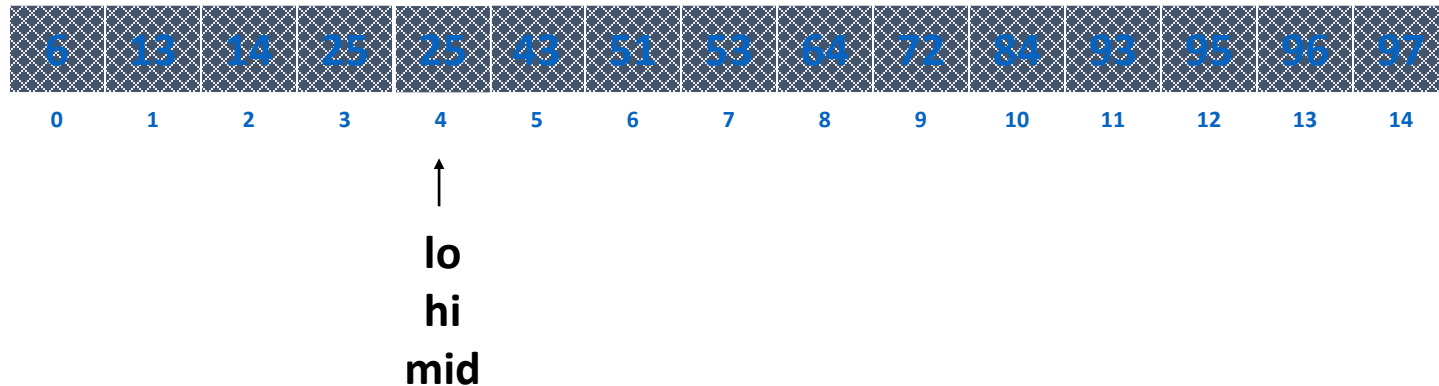
Binary Search – Worst Case

- Ex. Binary search for 34.



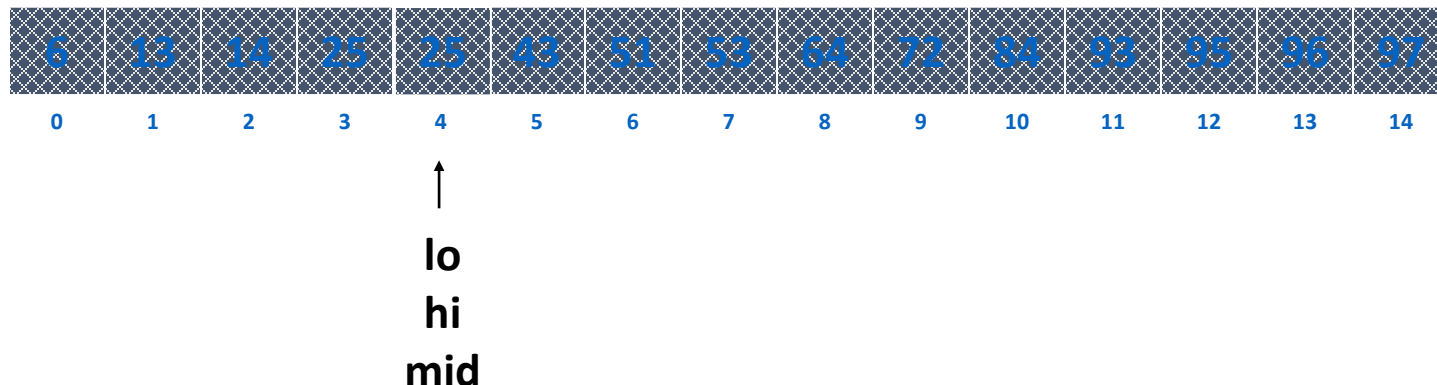
Binary Search – Worst Case

- How many comparisons would we have to make?



Binary Search – Worst Case

- How many comparisons would we have to make?
- Since we are dividing the list in half each time, how many times can we divide a list of length n by 2 before it reaches 1.



Binary Search – Worst Case

- How many times can I divide n before it equals 1?
- If I can divide n , k times before I get 1 then:

$$2^k = n$$

and

$$k = \log_2 n$$

- This means it would take me $\log_2 n$ compares to find out that the item is not in the list.

Binary Search – $O(\log_2 N)$

- We split the search space in half each time.
- If we had 100 items
 - $100 \rightarrow 50 \rightarrow 25 \rightarrow 12 \rightarrow 6 \rightarrow 3 \rightarrow 1$
- If we doubled it to 200 items, only 1 more search!
 - $200 \rightarrow 100 \rightarrow 50 \rightarrow 25 \rightarrow 12 \rightarrow 6 \rightarrow 3 \rightarrow 1$
- Each time we double the amount of items, it only takes 1 extra iteration of the search algorithm.

Binary Search

```
def binarySearch(aList, target):  
    # Returns index if target is in aList, -1 if it is not  
    # Assumes aList is sorted in ascending order  
    low = 0  
    high = len(aList)-1  
    while low <= high:  
        mid = (low + high) // 2  
        if aList[mid] == target:  
            return mid  
        elif aList[mid] > target:  
            high = mid - 1  
        else: low = mid + 1  
    return -1 # item not in list  
  
testlist = [0, 1, 2, 8, 13, 17, 19, 32, 42]  
print(binarySearch(testlist, 13))
```


What is the worst case time complexity of binary search?

- A. $O(1)$
- B. $O(n)$
- C. $O(n^2)$
- D. $O(\log(n))$

1. Visit <https://flux.qa>
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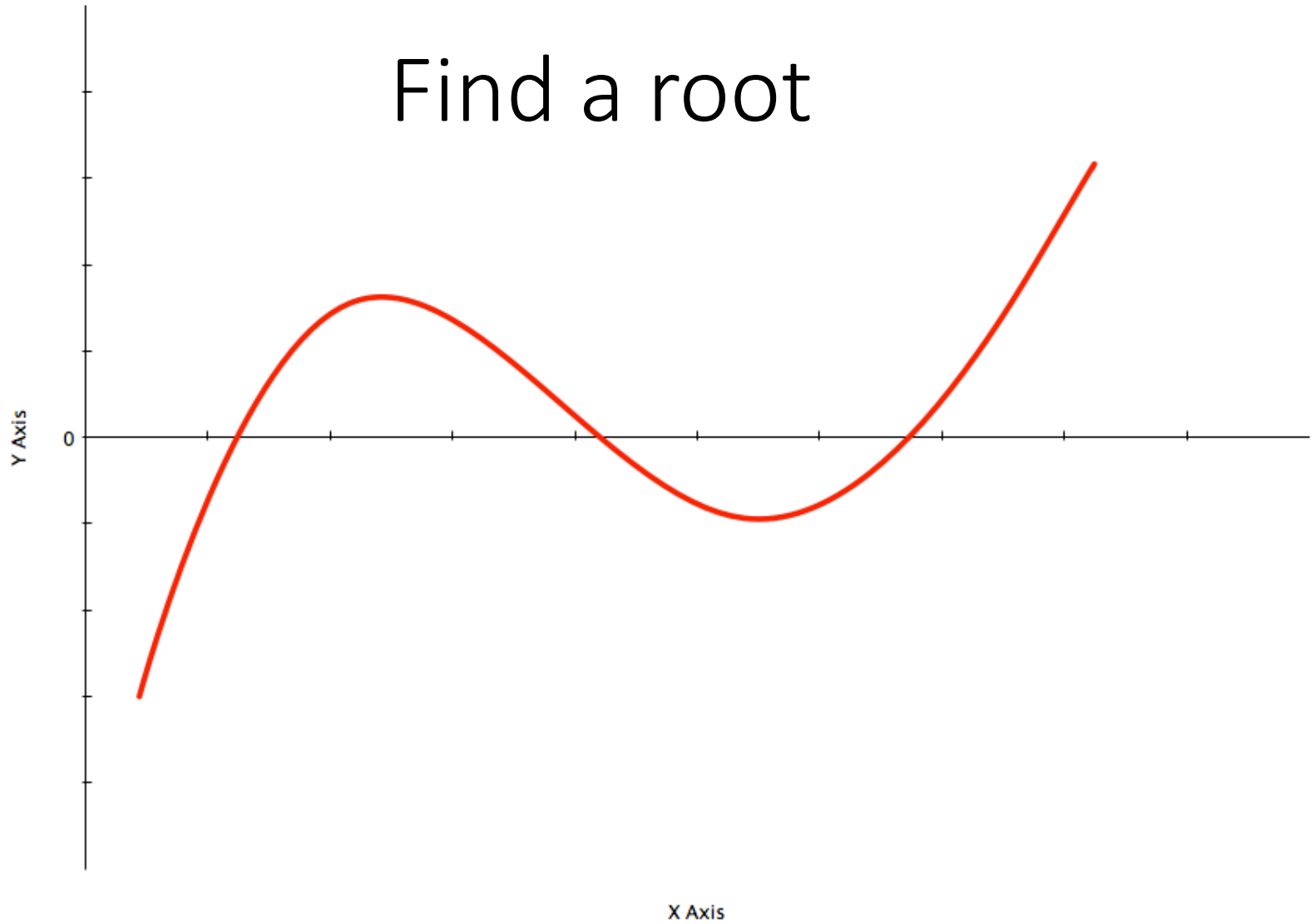
Binary Search – $O(\log N)$

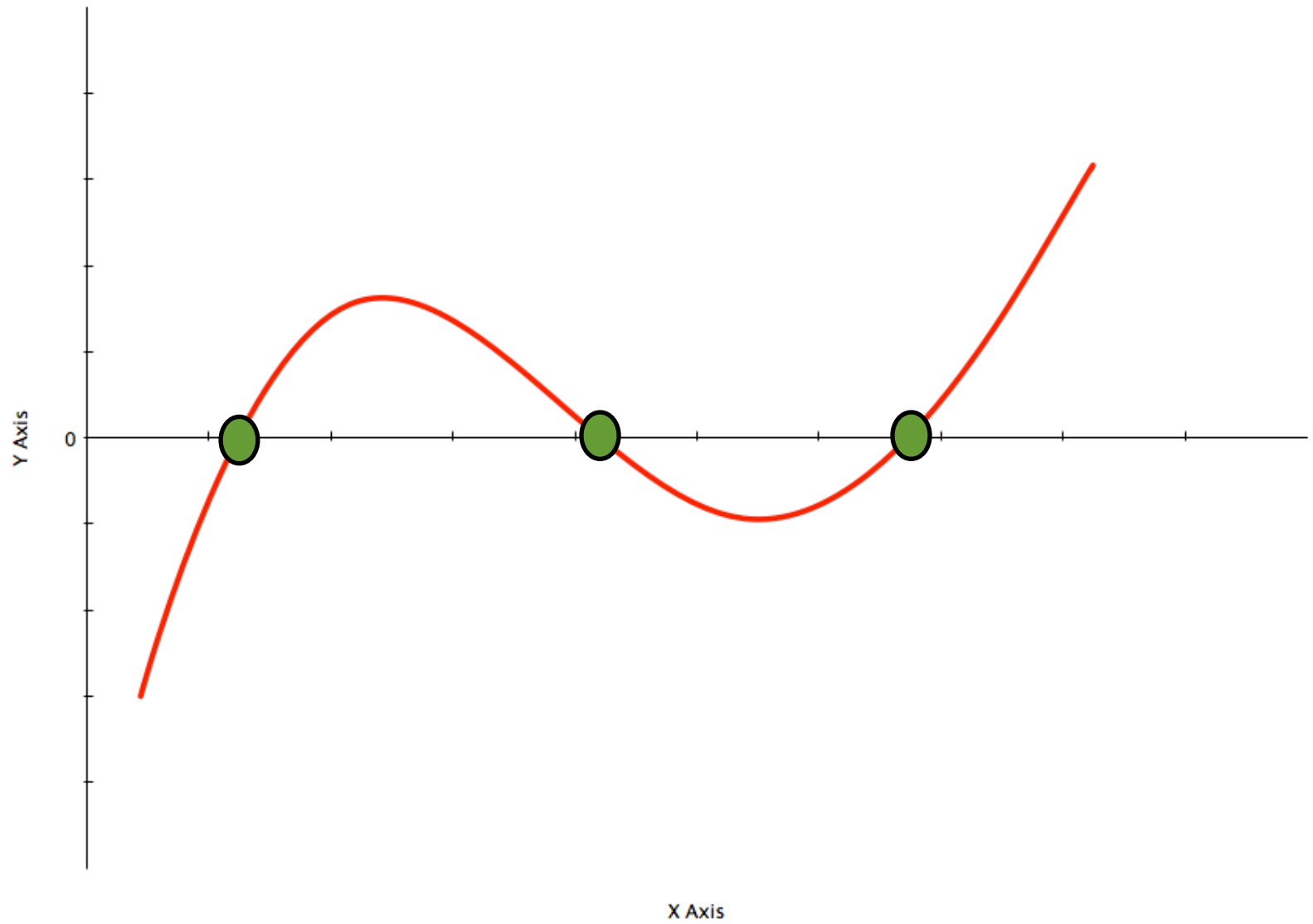
- 1 items = 1 search
- 2 items = 2 searches
- 4 items = 3 searches
- ...
- 1024 items = 10 searches
- 2048 items = 11 searches
- ...
- 1 million items \sim 20 searches
- 1 billion items \sim 30 searches

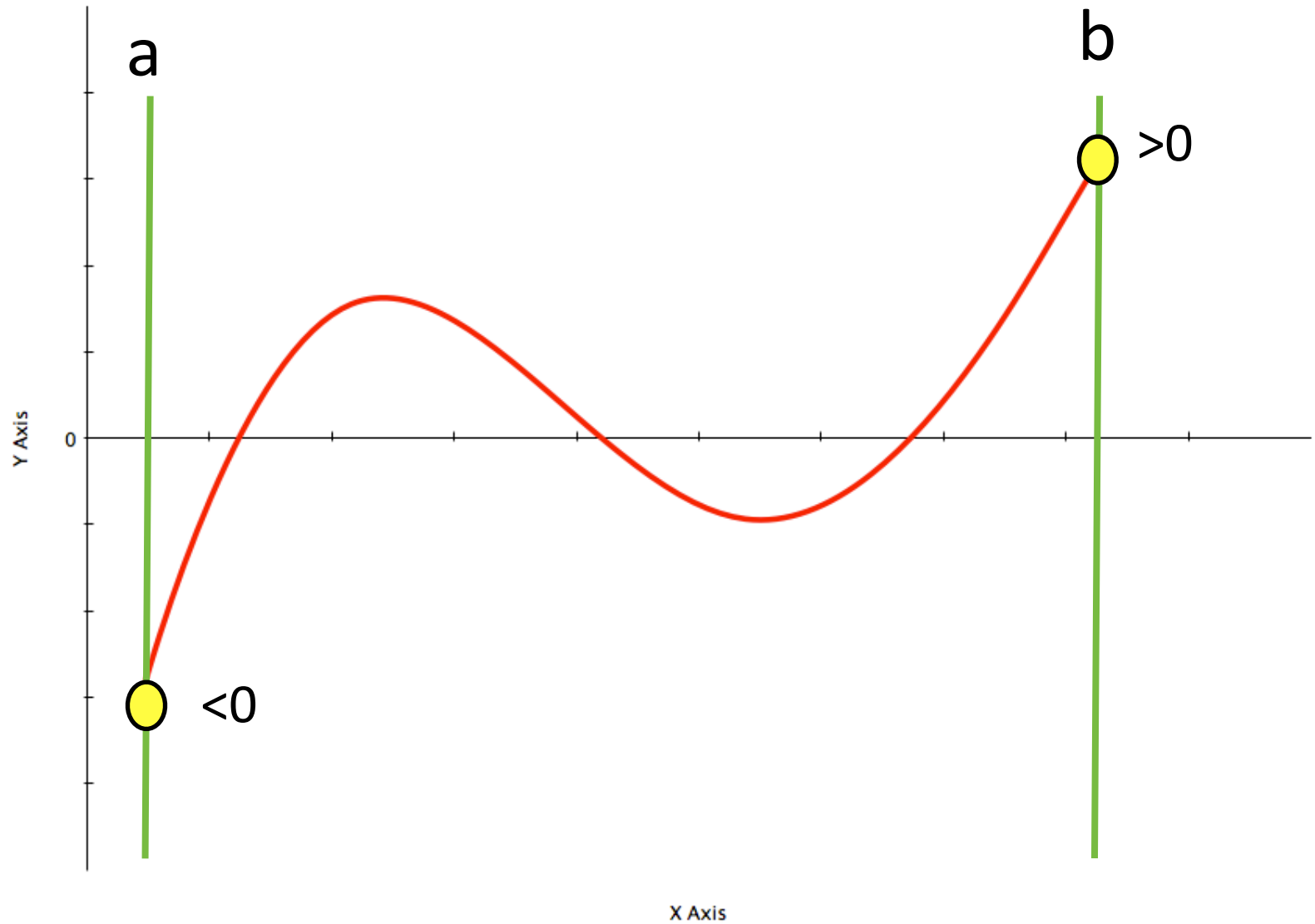
- Logarithmic!
- Binary Search = $O(\log N)$
 - Compare to Linear Search $O(N)$

Binary Search Application- Find the root (Bisection Method)

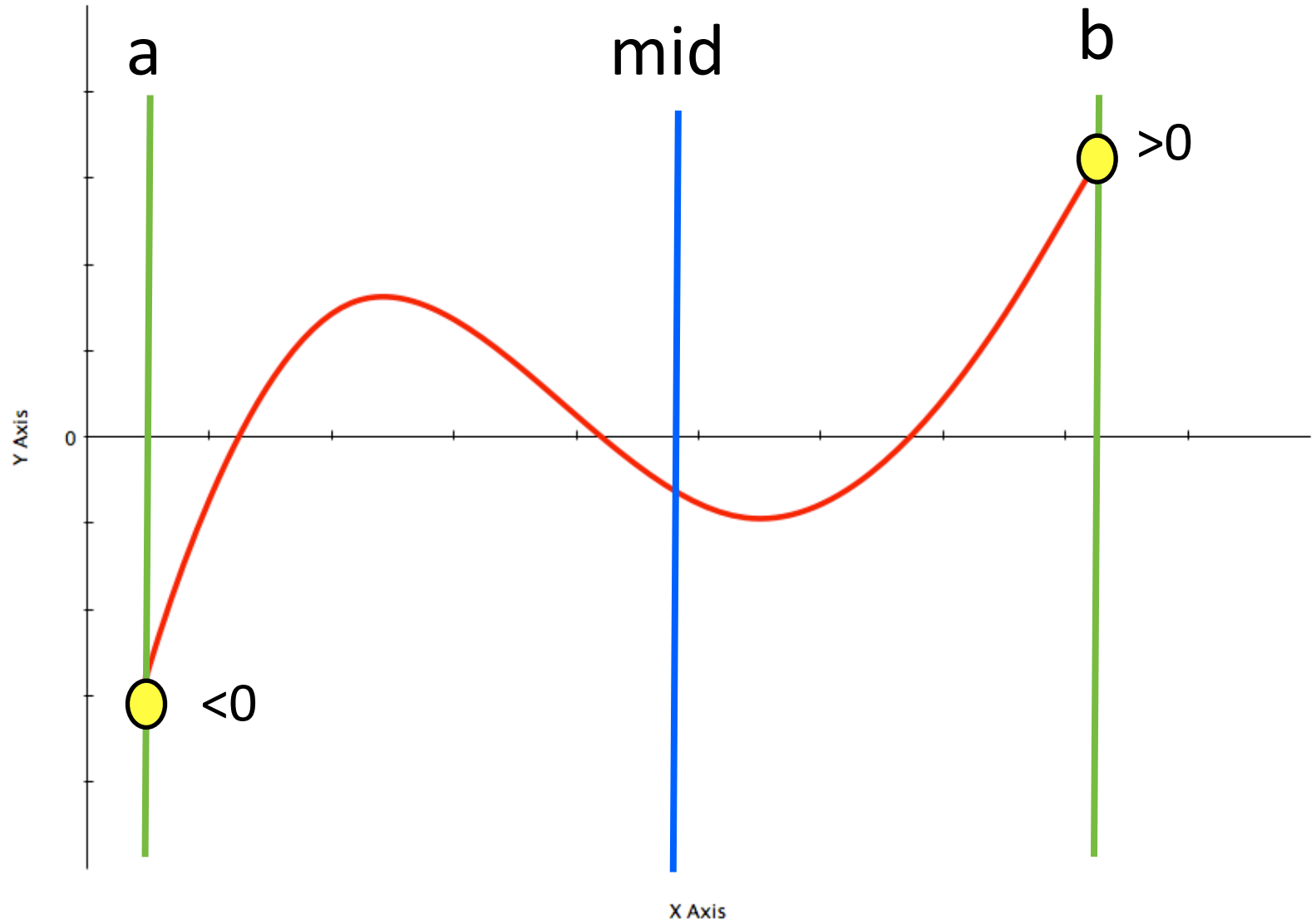
Find a root

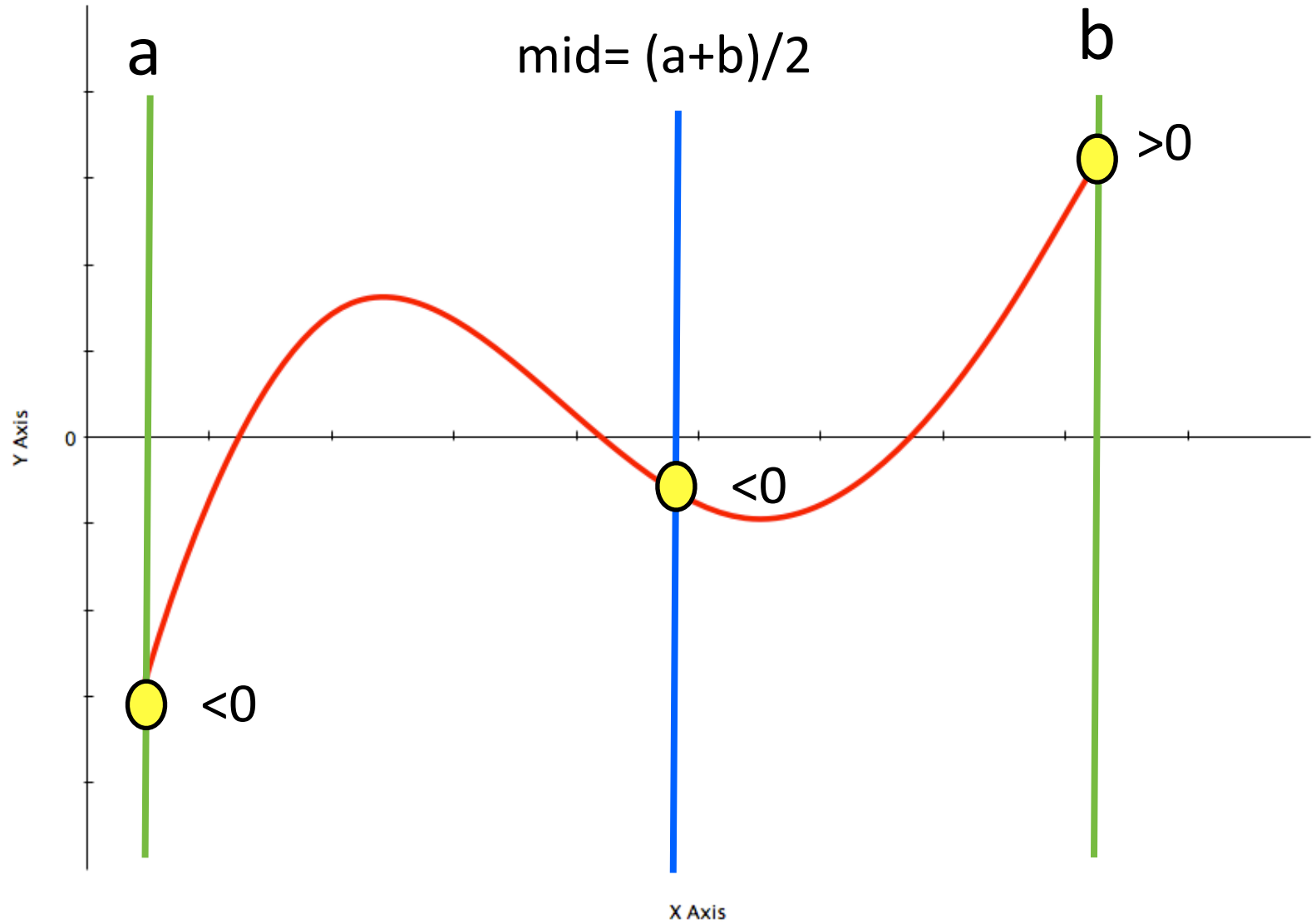


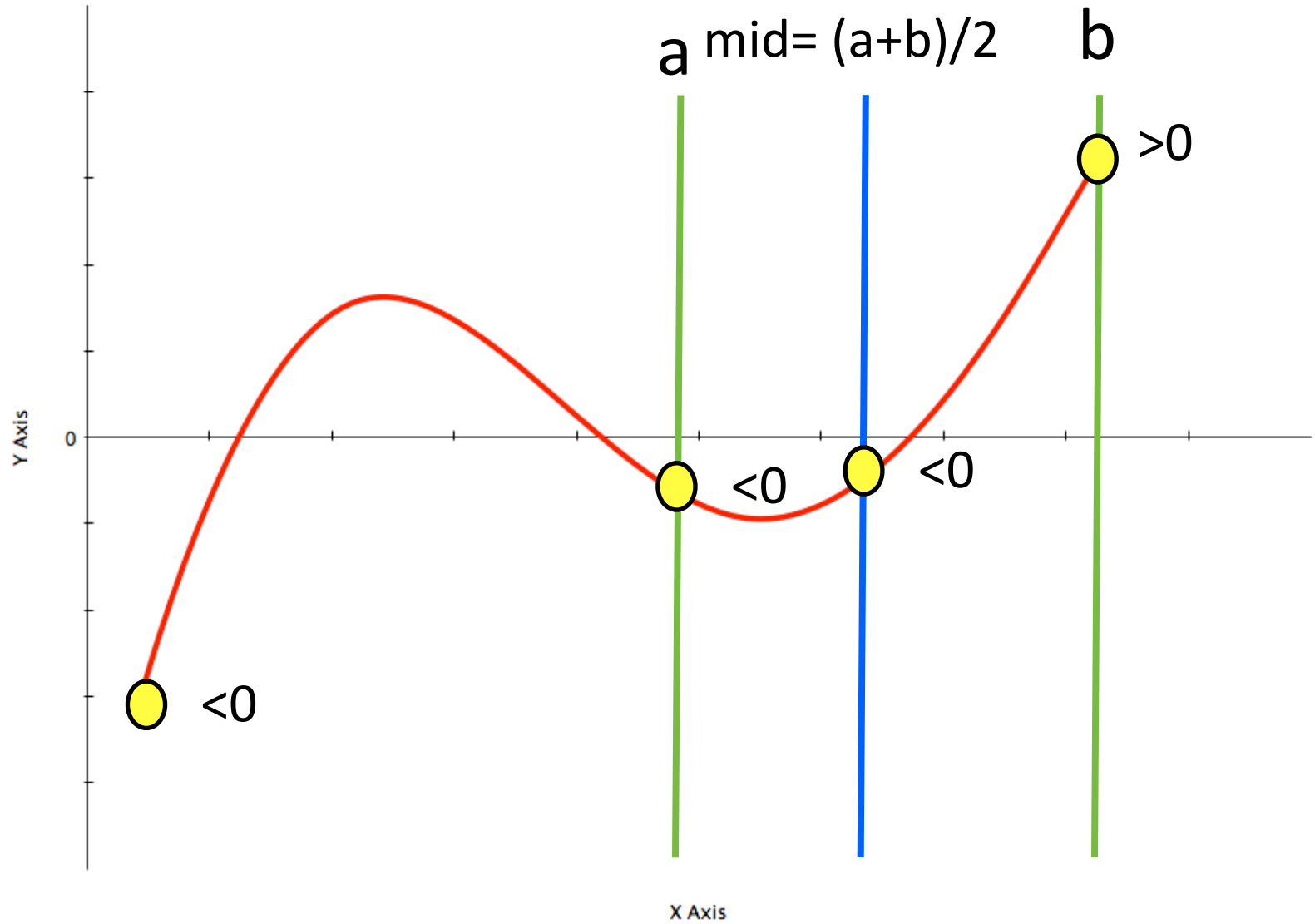


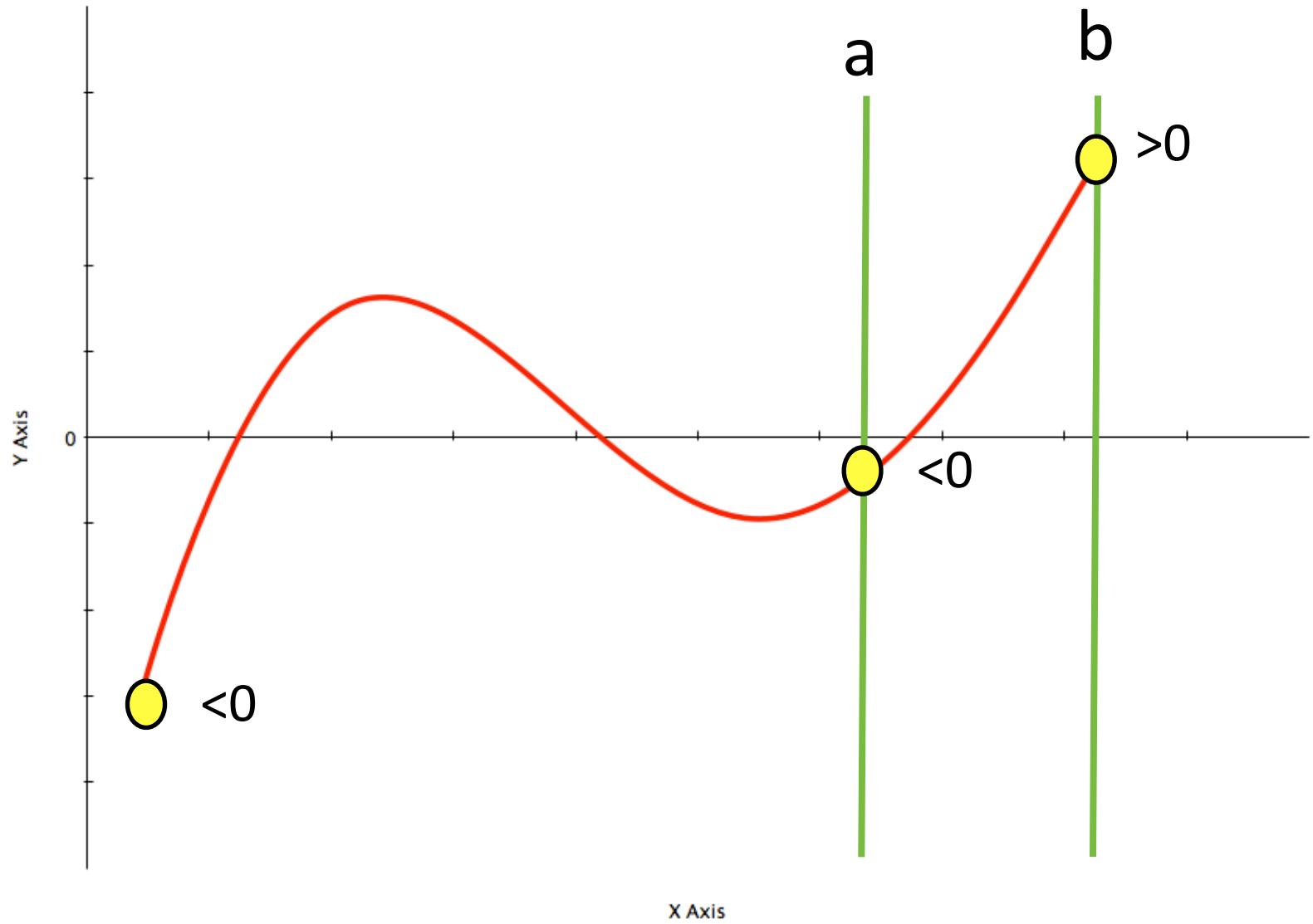


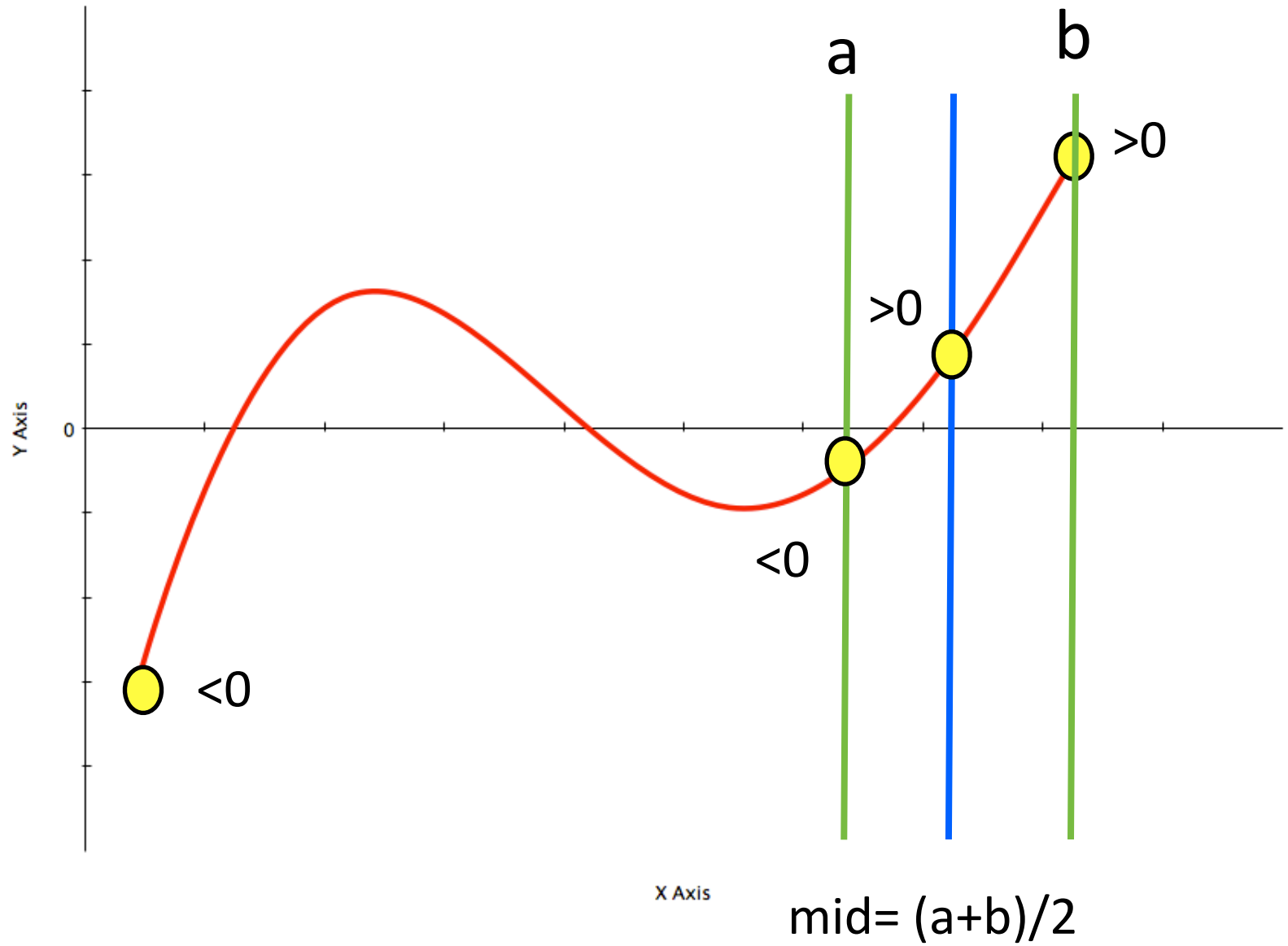
Assume that the function is continuous

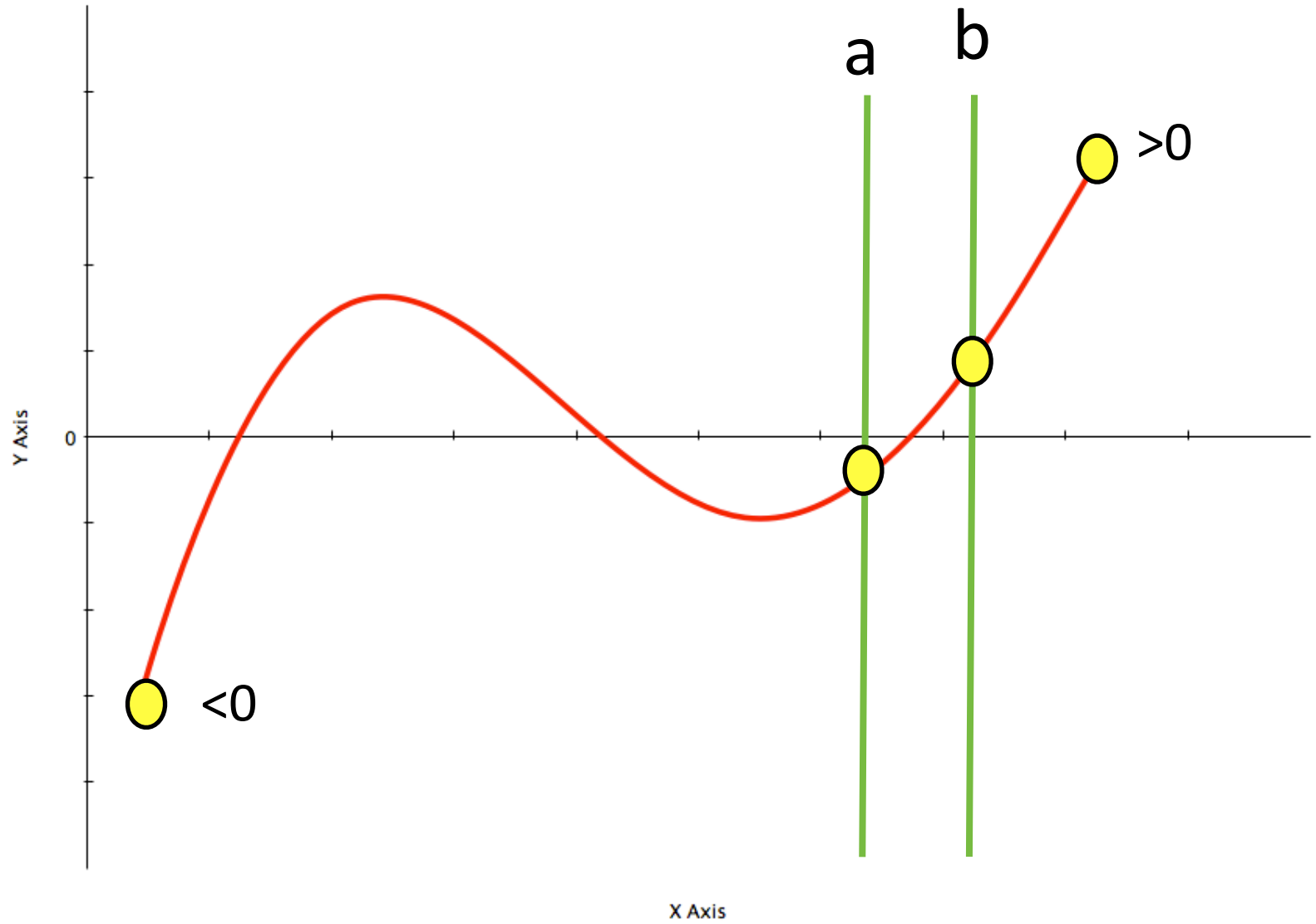


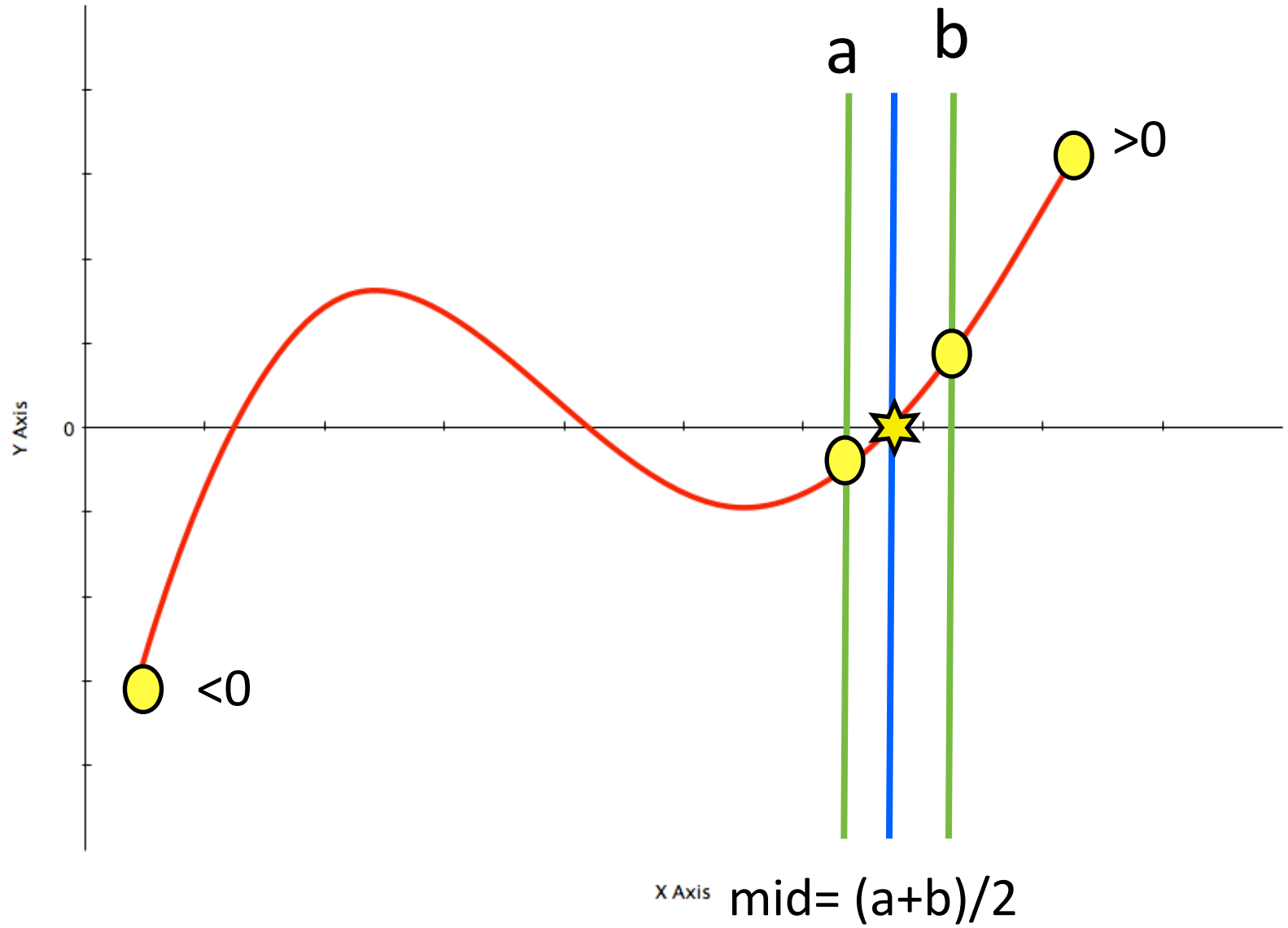










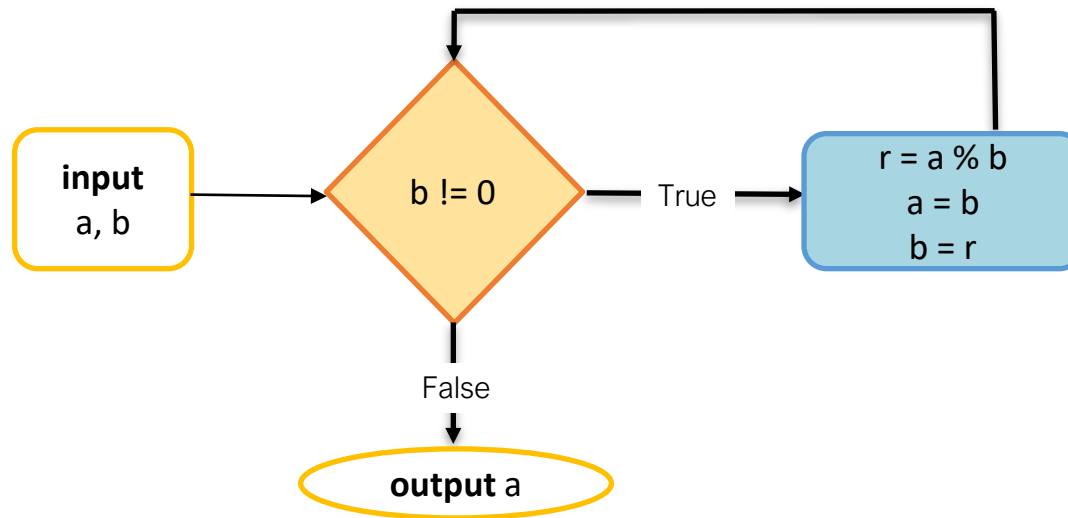


- In a computer all real values are approximations
- Two real values may never be exactly the same,
- But can be close enough

Overview

1. The Ordered Search Problem
2. Binary Search
3. Revisiting Euclid's Algorithm

Recall Euclid's Algorithm

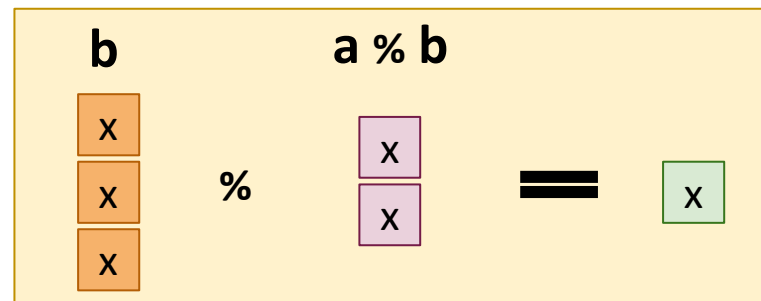
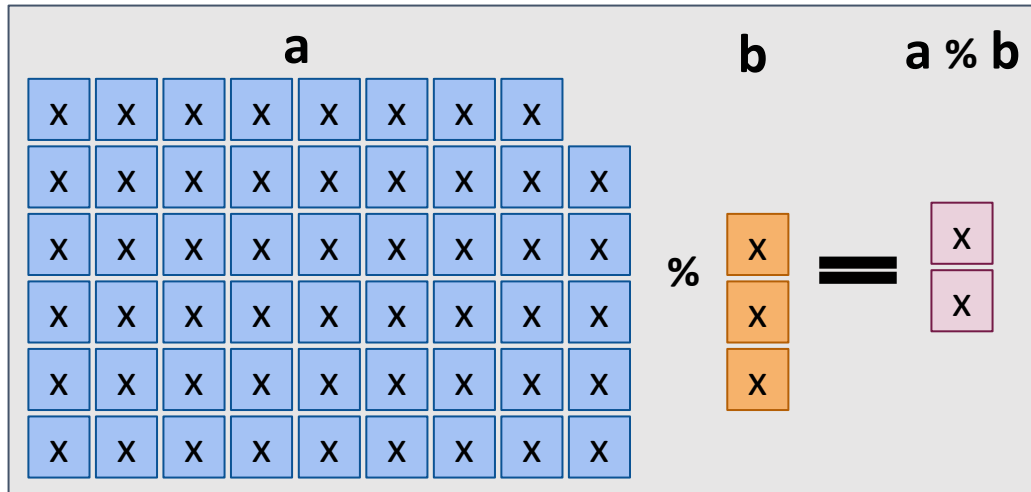


Eukleides of Alexandria
3xx BC – 2xx BC

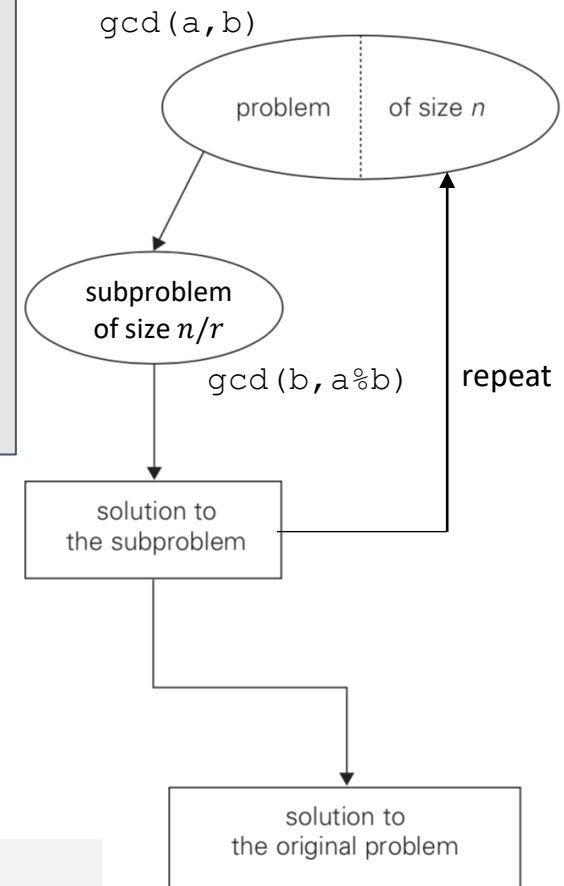
```

def gcd_euclid(a, b):
    """
    Input : integers a and b such that not a==b==0
    Output: the greatest common divisor of a and b
    """
    while b != 0:
        a, b = b, a % b
    return a
  
```


Instance of decrease and conquer

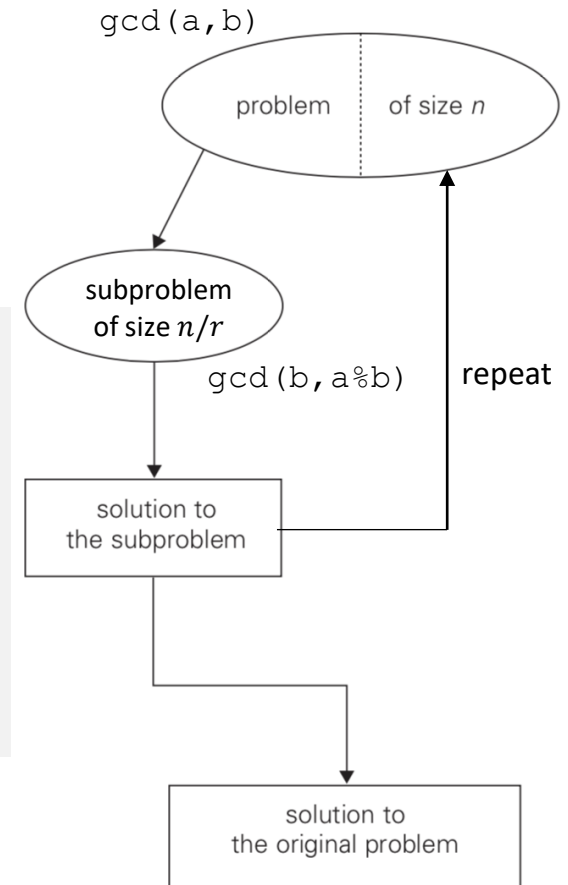


```
def gcd_euclid(a, b):
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```



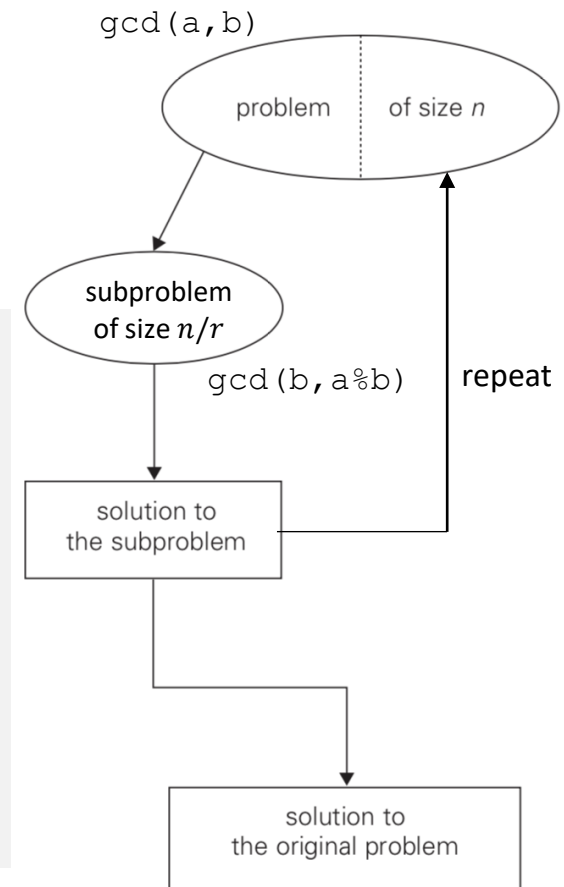
Exercise: correctness via loop invariant

```
def gcd_euclid(a, b):
    """
    I: integers a and b such
        that not a==b==0
    O: gcd(a,b)
    """
    while b != 0:
        a, b = b, a % b
    return a
```



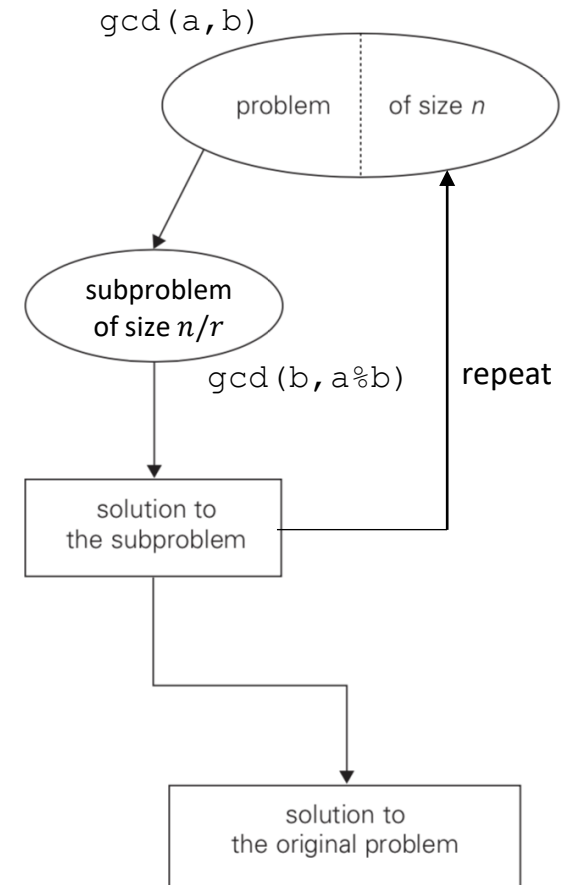
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    """
    I: integers a and b such
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    O: gcd(a,b)
    """
    #PRC: a,b==a0,b0 (original input)
    while b != 0:
        a, b = b, a % b
    return a
```



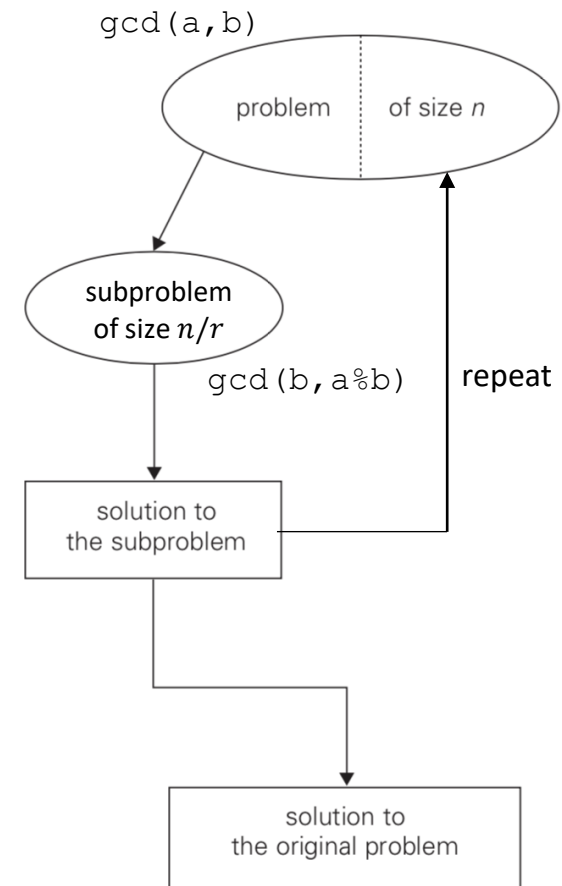
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    """
    #PRC: a,b==a0,b0 (original input)
    while b != 0:
        #I: gcd(a,b)==gcd(a0,b0)
        a, b = b, a % b
        #I: gcd(a,b)==gcd(a0,b0)
    return a
```



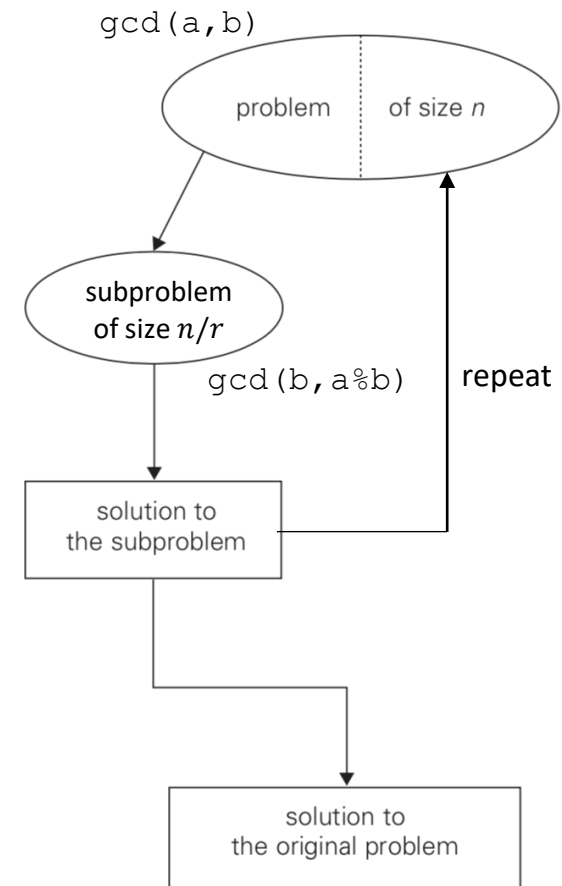
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    O: gcd(a0,b0)
    """
    #PRC: a,b==a0,b0 (original input)
    while b != 0:
        #I: gcd(a,b)==gcd(a0,b0)
        a, b = b, a % b
        #I: gcd(a,b)==gcd(a0,b0)
    #EXC: b==0
    return a
```



Exercise: correctness via loop invariant

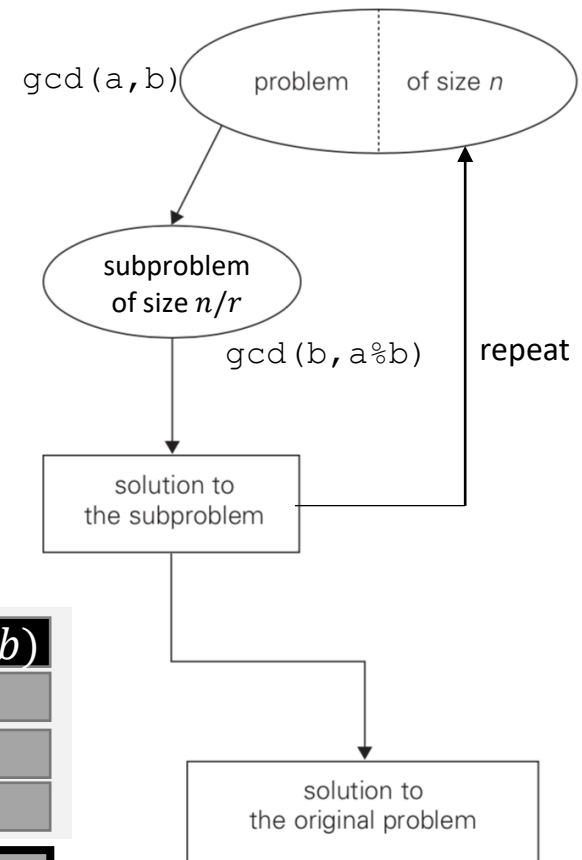
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    I: integers a and b such
        that not a==b==0
    O: gcd(a,b)
    """
    #PRC: a,b==a0,b0 (original input)
    while b != 0:
        #I: gcd(a,b)==gcd(a0,b0)
        a, b = b, a % b
        #I: gcd(a,b)==gcd(a0,b0)
    #EXC: b==0
    #POC: a==gcd(a,b)==gcd(a0,b0)
    return a
```



Can we analyse computational complexity as for binary search?

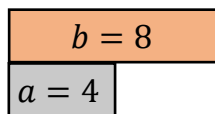
Need to determine how many iterations we can have in worst case!

<code>def gcd_euclid(a, b):</code>	<code>n = abs(a) + abs(b)</code>	<code>O(1)</code>	<code>?</code>
<code>while b != 0:</code>	<code>a, b = b, a % b</code>	<code>0</code>	<code>?</code>
<code>return a</code>		<code>0</code>	<code>0</code>
		<code>O(1)</code>	<code>?</code>

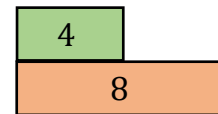


By what factor is problem decreased per iteration of Euclid's Algorithm?

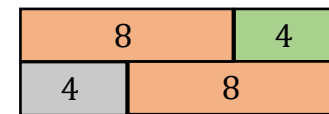
Example 1



$$\gcd(8, 4) = \gcd(4, 8)$$

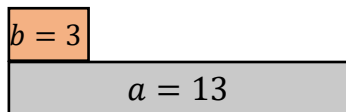


No decrease in problem size!

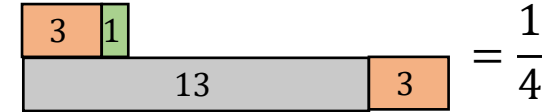
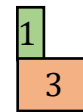


But this can only happen once in the beginning (afterwards $b > a \% b$ guarantees $a > b$)

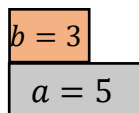
Example 2



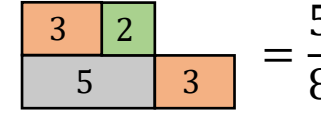
$$\gcd(13, 3) = \gcd(3, 1)$$



Example 3



$$\gcd(5, 3) = \gcd(3, 2)$$



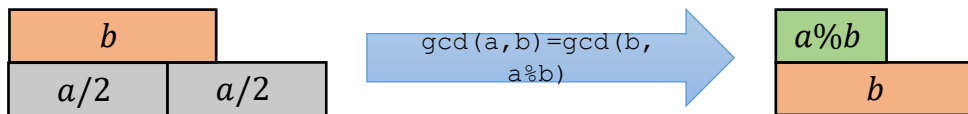
When $a > b$ there is decrease but at varying rate!

Do we need a fixed rate of decrease for logarithmic complexity?

No, just guarantee that reduction factor is always at least some $\alpha > 1$

First case: “large b”

Case $b \geq a/2$



Relative size of decreased problem with large b

Case $b \geq a/2$

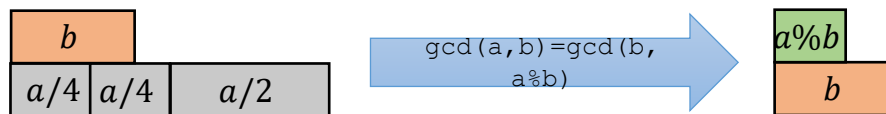
$$\begin{array}{|c|c|c|} \hline b & a \% b & \\ \hline a/2 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & a \% b & \\ \hline a/2 & a/2 & a/2 \\ \hline \end{array} = \frac{2}{3}$$

Second case: “small b”

Case $b \geq a/2$

$$\begin{array}{|c|c|} \hline b & a \% b \\ \hline a/2 & a/2 \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & a \% b & \\ \hline a/2 & a/2 & a/2 \\ \hline \end{array} = \frac{2}{3}$$

Case $a/2 > b \geq a/4$



Relative size of decreased problem in second case

Case $b \geq a/2$

$$\begin{array}{|c|c|} \hline b & a \% b \\ \hline a/2 & a/2 \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & a \% b & \\ \hline a/2 & a/2 & a/2 \\ \hline \end{array} = \frac{2}{3}$$

Case $a/2 > b \geq a/4$

$$\begin{array}{|c|c|c|c|} \hline b & a \% b & & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|c|} \hline b & b & & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|c|} \hline a/4 & a/4 & a/2 & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \frac{4}{5}$$

Final case: “tiny b”

Case $b \geq a/2$

$$\begin{array}{|c|c|} \hline b & a \% b \\ \hline a/2 & a/2 \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & a \% b & \\ \hline a/2 & a/2 & a/2 \\ \hline \end{array} = \frac{2}{3}$$

Case $a/2 > b \geq a/4$

$$\begin{array}{|c|c|c|c|} \hline b & a \% b & & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|c|} \hline b & b & & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|c|} \hline a/4 & a/4 & a/2 & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \frac{4}{5}$$

Case $a/4 > b$

$$\begin{array}{|c|c|} \hline b & \\ \hline a/4 & 3a/4 \\ \hline \end{array} \xrightarrow{\gcd(a,b) = \gcd(b, a \% b)} \begin{array}{|c|} \hline \\ \hline b \\ \hline \end{array}$$

Relative size of decreased problem in final case

Case $b \geq a/2$

$$\begin{array}{|c|c|} \hline b & a \% b \\ \hline a/2 & a/2 \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & a \% b & \\ \hline a/2 & a/2 & a/2 \\ \hline \end{array} = \frac{2}{3}$$

Case $a/2 > b \geq a/4$

$$\begin{array}{|c|c|c|c|} \hline b & a \% b & & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|c|} \hline b & b & & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|c|} \hline a/4 & a/4 & a/2 & \\ \hline a/4 & a/4 & a/2 & b \\ \hline \end{array} \leq \frac{4}{5}$$

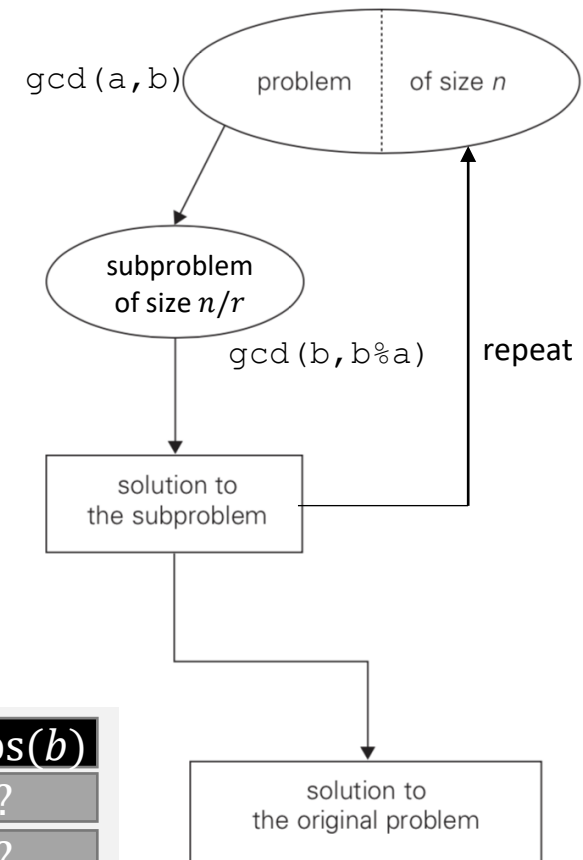
Case $a/4 > b$

$$\begin{array}{|c|c|c|} \hline b & & \\ \hline a/4 & 3a/4 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & b & \\ \hline a/4 & 3a/4 & b \\ \hline \end{array} \leq \begin{array}{|c|c|c|} \hline b & b & \\ \hline a/4 & 3a/4 & \\ \hline \end{array} \leq \frac{1}{2}$$

In all cases: problem is at least decreased by a rate r of $5/4$

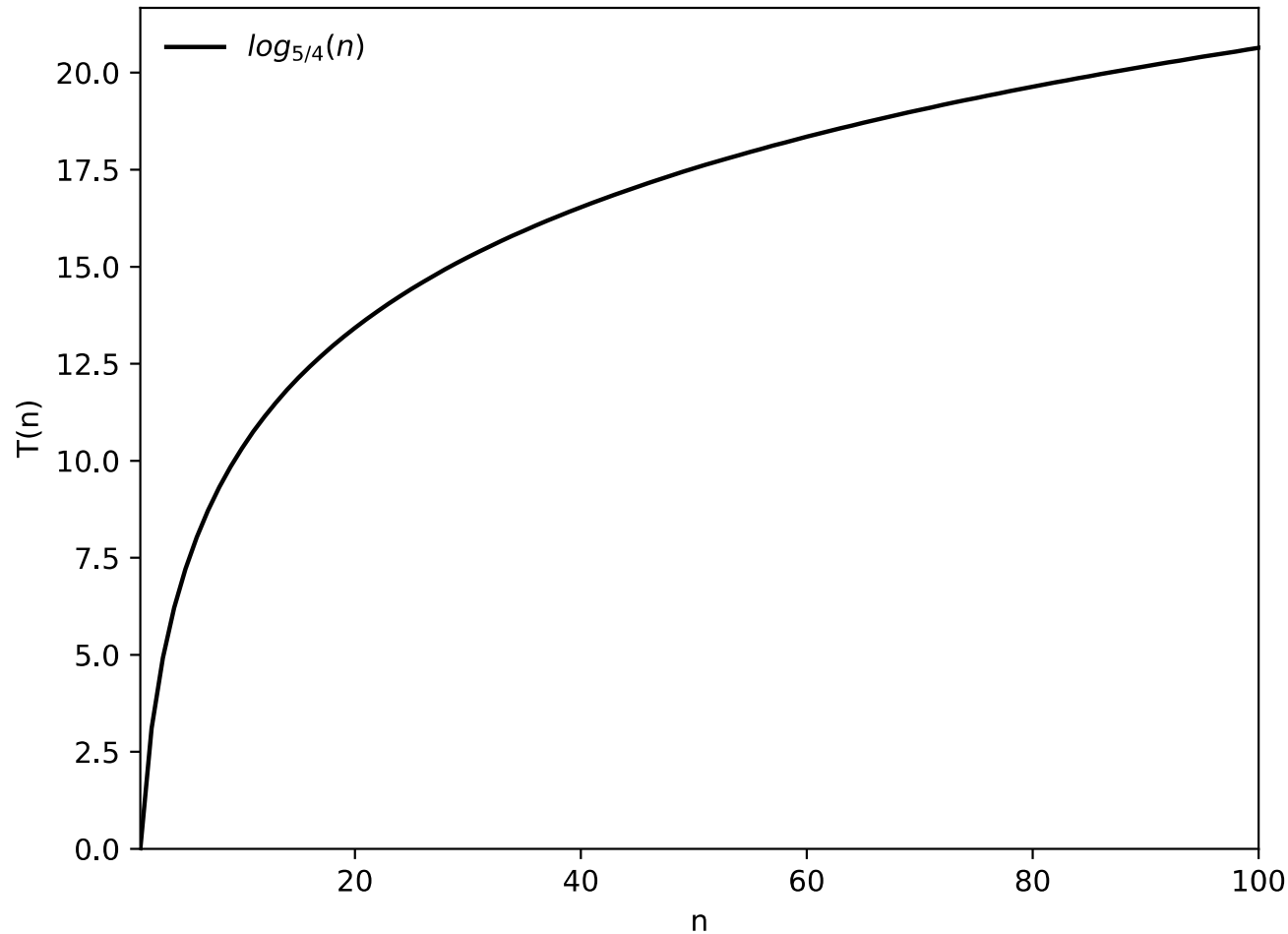
Almost identical analysis as for binary search

- Let $n_i = a_i + b_i$ be problem size after i iterations of loop
- In the beginning: $n_0 = n$
- Per iteration size is reduced by **at least**:
 $n_i = \lceil n_{i-1}/r \rceil$, i.e., $n_i = \lceil n/r^i \rceil$
- After at most $k = \lceil \log_r n \rceil$ iterations:
 $n_k = \lceil n/r^{\log_r n} \rceil = 1$, i.e., $b_k = 0$ and $a_k = 1$
- So at most $O(\log_r n)$ loop iterations

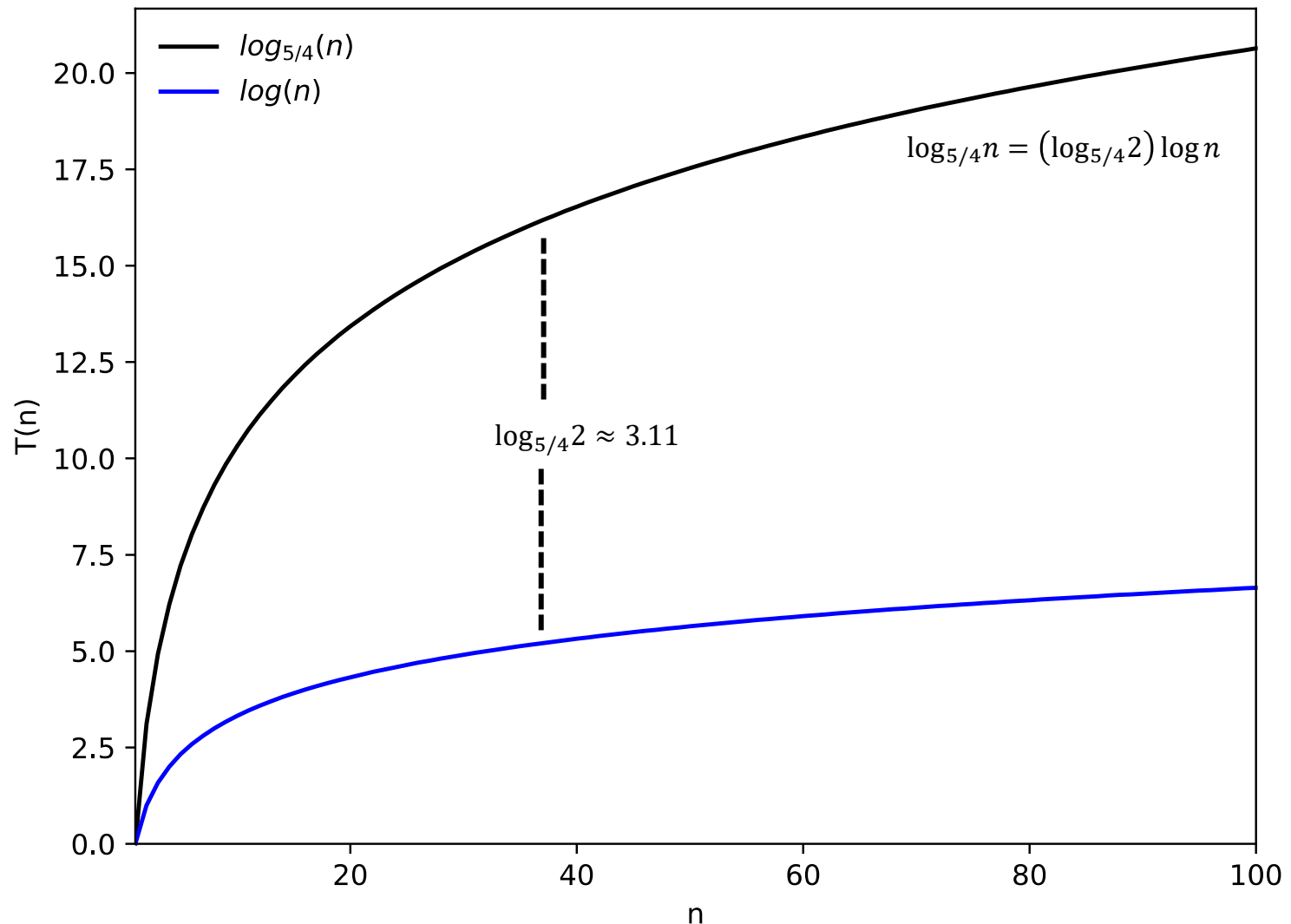


<code>def gcd_euclid(a, b):</code>	<code>n = abs(a) + abs(b)</code>	<code>O(1)</code>	<code>?</code>
<code>while b != 0:</code>	<code>a, b = b, a % b</code>	<code>0</code>	<code>?</code>
<code>return a</code>		<code>0</code>	<code>0</code>
		<code>O(1)</code>	<code>?</code>

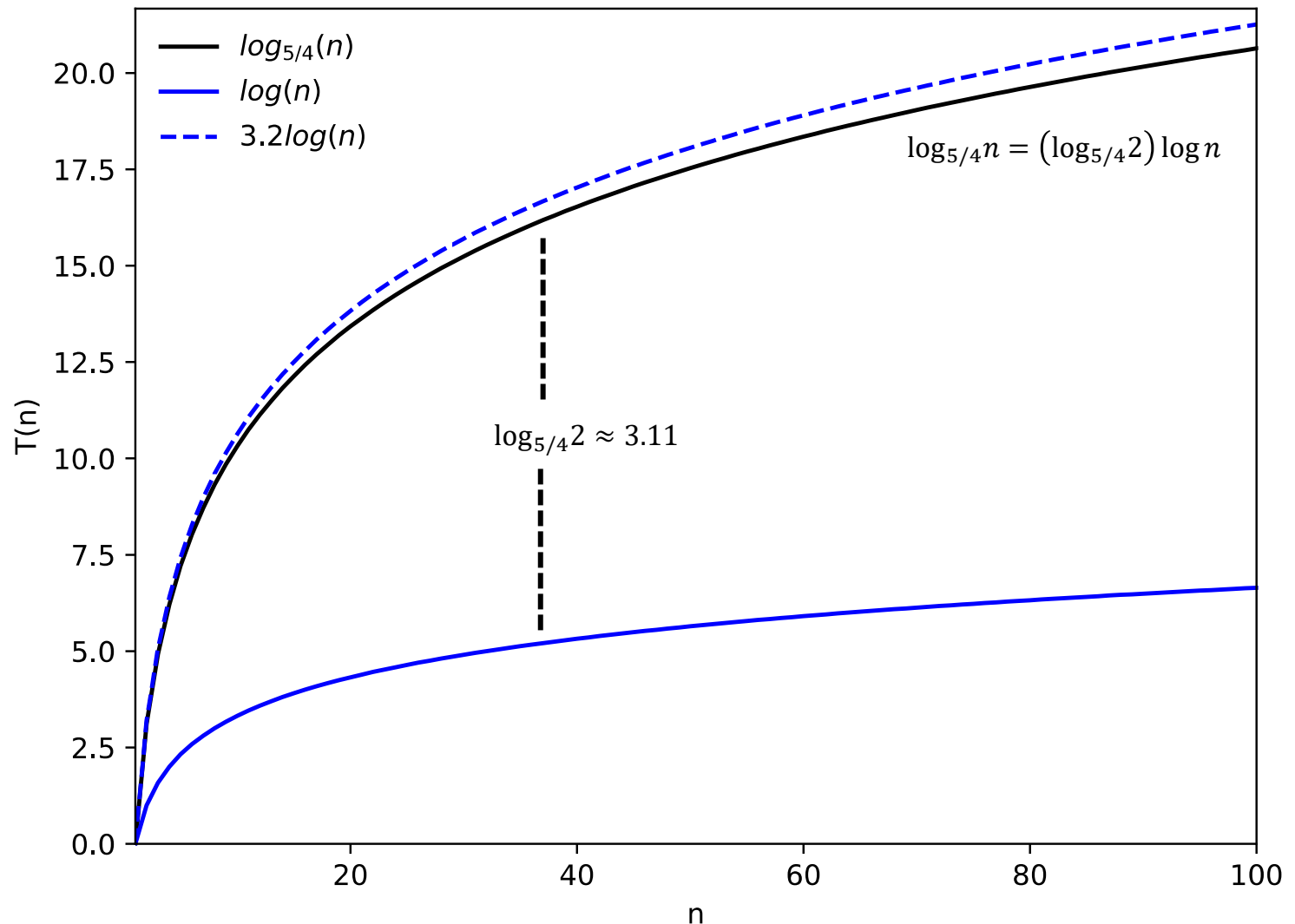
What does this mean in terms of order of growth?



Is order of growth \log base $5/4$ higher than \log base 2?

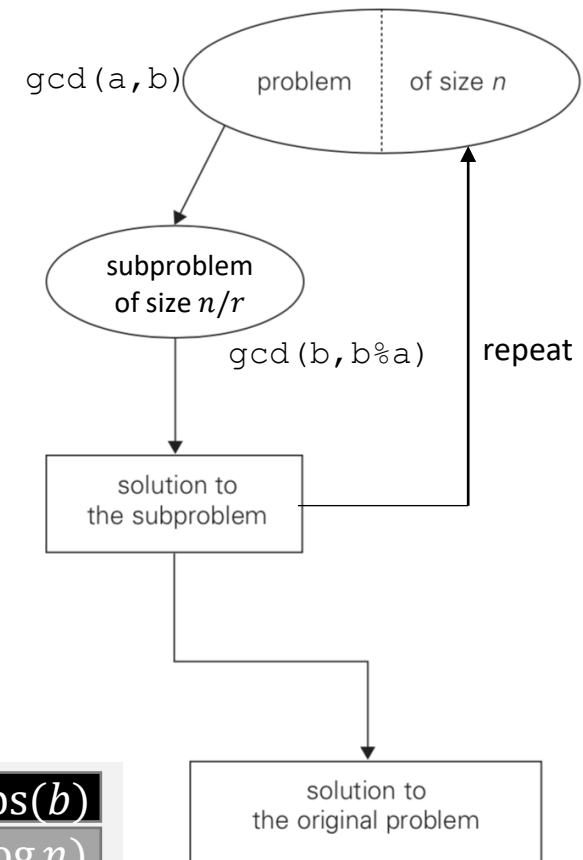


No: $O(\log n) = O(\log_r n)$



Almost identical analysis as for binary search

- Let $n_i = a_i + b_i$ be problem size after i iterations of loop
- In the beginning: $n_0 = n$
- Per iteration size is reduced by **at least**:
 $n_i = \lceil n_{i-1}/r \rceil$, i.e., $n_i = \lceil n/r^i \rceil$
- After at most $k = \lceil \log_r n \rceil$ iterations:
 $n_k = \lceil n/r^{\log_r n} \rceil = 1$, i.e., $b_k = 0$ and $a_k = 1$
- So at most $O(\log_r n)$ loop iterations



<code>def gcd_euclid(a, b) :</code>	<code>n = abs(a) + abs(b)</code>	$O(1)$	$O(\log n)$
<code>while b != 0 :</code>	<code>a, b = b, a % b</code>	$O(1)$	$O(\log n)$
<code>return a</code>		$O(1)$	$O(\log n)$

Summary

Algorithmic paradigm: **decrease-and-conquer**

- **decreasing** problem size by at least some rate $r > 1$ leads to trivial subproblems after logarithmically many reductions
- if not too much overhead: allows to replace linear complexity term by **logarithmic term**

Binary Search allows logarithmic time look-up of value in sorted sequence

Euclid's Algorithm finds gcd in time logarithmically in sum of input $\text{abs}(a) + \text{abs}(b)$

Coming Up

- More examples for algorithm analysis
- Divide and conquer