
Searching



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Searching

Imagine a data structure containing 9 billion **unordered** names

# 1					Peart
# 2					Schock
# 3					Crump
.					
.					
.					
# 8,999,999,998					White
# 8,999,999,999					Purdie
# 9,000,000,000					Bruford

Searching

Imagine a data structure containing 9 billion **unordered** names and we want to locate **one** of them.

check # 1				Peart
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.				
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← Is this it? If not, $n-1$ to go.

← Is this it? If not, $n-2$ to go.

← Is this it? If not, $n-3$ to go.

← Is this it? If not, 2 to go.

← Is this it? If not, 1 to go.

← Is this it? If not, it's not here.

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← Is this it? If not, 2 to go.
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Sometimes we will find the target person early.

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Sometimes we will find the target person early.
Sometimes we will find the target person late.

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This is called
Linear Search or
Sequential Search

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Q: What's the average — or expected — case for n items?

Linear / Sequential Searching

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A: The expected case is $\frac{1}{2} n$, which requires examining 4.5B rows in this example.

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That's $O(n)$.

Best case $O(1)$
Worst case $O(n)$
Average case $O(n)$

Recurrence:
 $T(n) = T(n-1) + c$

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LIST-SEARCH(L, k)

```
1  $x = L.head$ 
2 while  $x \neq \text{NIL}$  and  $x.key \neq k$ 
3      $x = x.next$ 
4 return  $x$ 
```

To search a list of n objects, the LIST-SEARCH procedure takes $\Theta(n)$ time in the worst case, since it may have to search the entire list.

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Average case $O(n)$

Can we do better?

LIST-SEARCH(L, k)

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To search a list of n objects, the LIST-SEARCH procedure takes $\Theta(n)$ time in the worst case, since it may have to search the entire list.

Searching 9 billion people

What if we could **search** through **sorted** data?

After all, we are good at sorting...

... in $O(n^2)$ time — Selection and Insertion sort

... in $O(n \log_2 n)$ time — Merge sort and Quicksort

Searching 9 billion people

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How would you do it?
What's your strategy?

Want to play a number guessing game?

Searching 9 billion people

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We could pick from the middle. If that's not our target, then we **exclude** the *lower* or *upper half of the data*, depending on whether our target is greater or lesser than the value we picked. Then we pick the middle of the remaining half. Repeat.

Have we seen this before?

Divide and . . .

Take a big problem and divide it into two smaller problems.

Take a those problems and divide them into two smaller problems.

Take a those problems and divide them into two smaller problems.

Take a those problems and divide them into two smaller problems.

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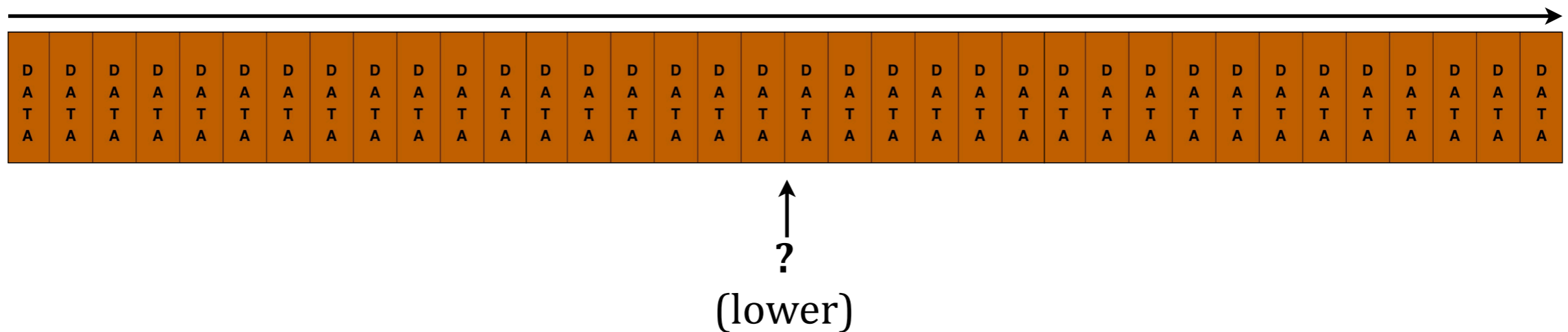
... until the problems get small enough that they are solved.

In this case, it's really just divide.

Let's consider **Binary Search**.

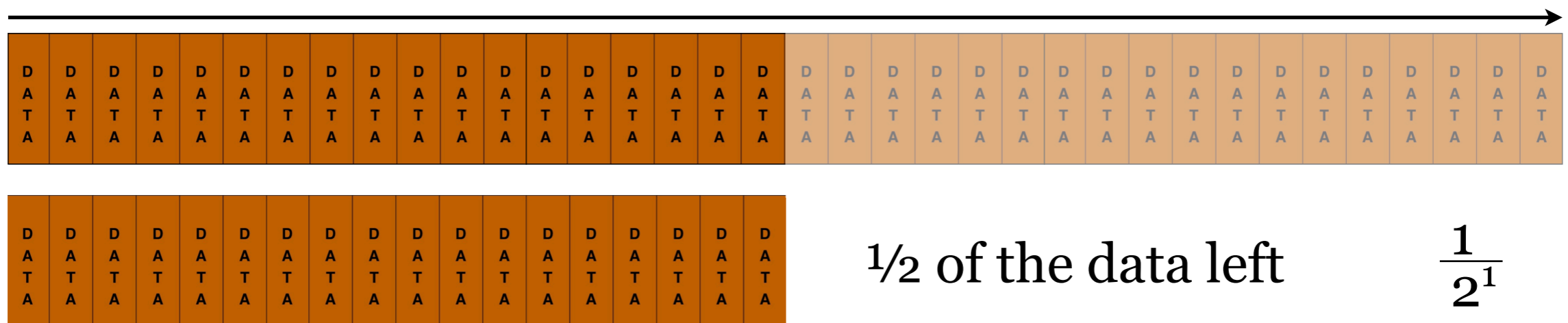
Binary Searching 9 billion people

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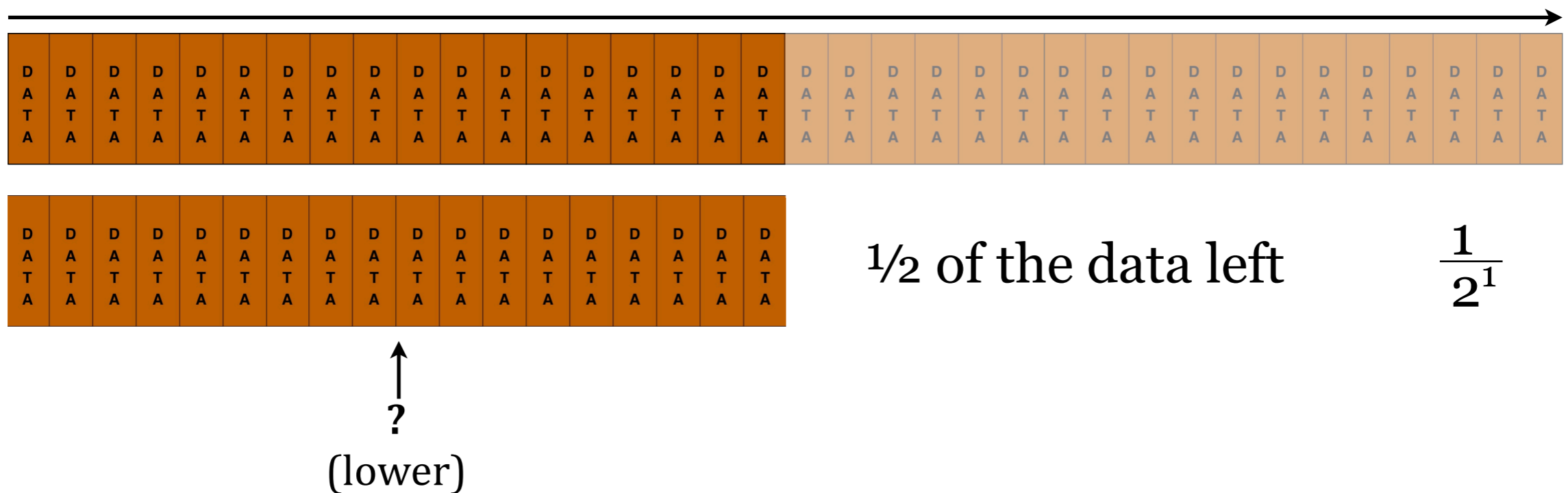
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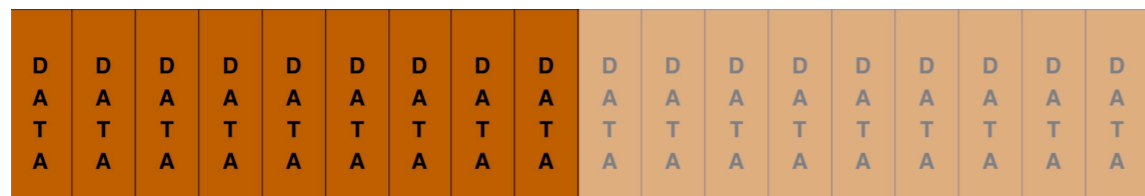
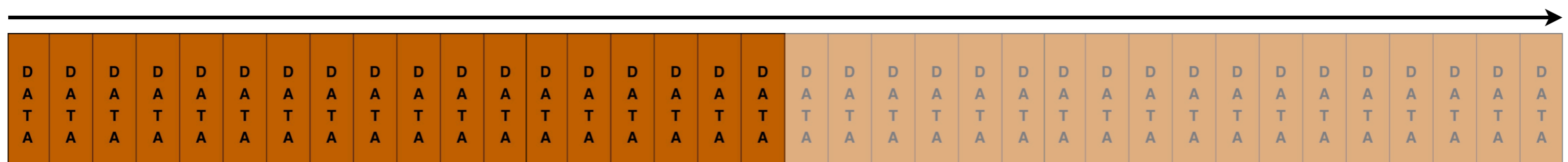
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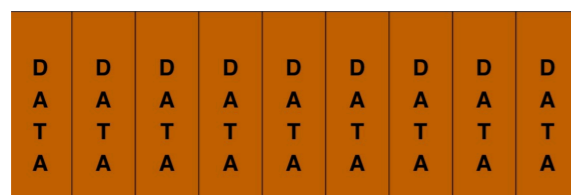
Binary Searching 9 billion people

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1/2 of the data left

$$\frac{1}{2^1}$$

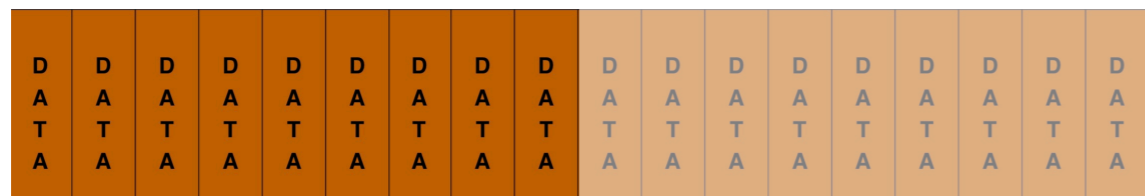
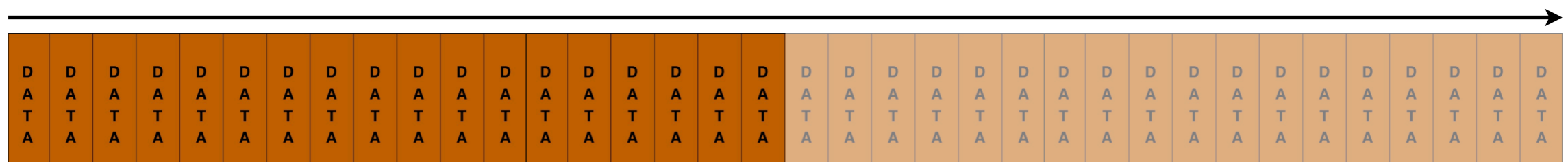


1/4 of the data left

$$\frac{1}{2^2}$$

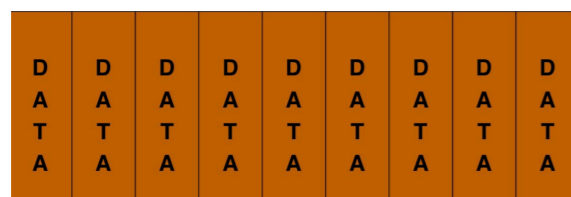
Binary Searching 9 billion people

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$$\frac{1}{2^1}$$



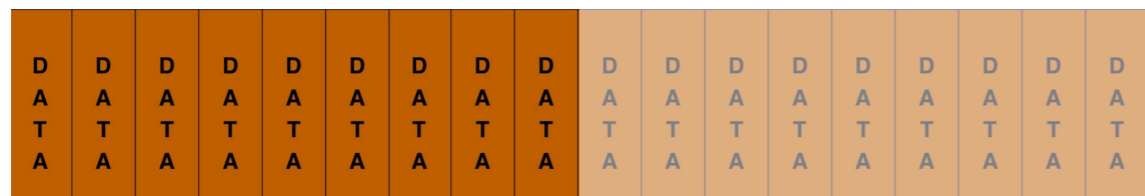
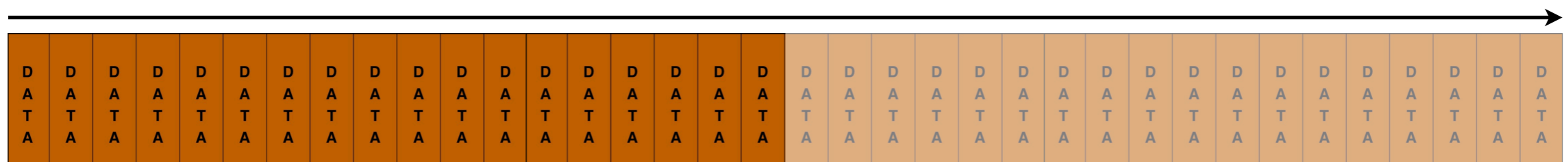
1/4 of the data left

$$\frac{1}{2^2}$$

↑
?
(higher)

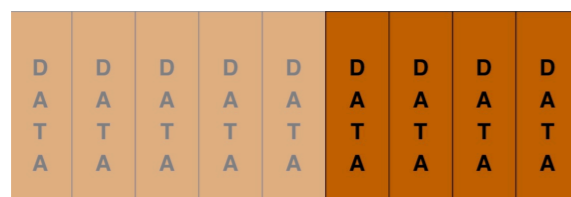
Binary Searching 9 billion people

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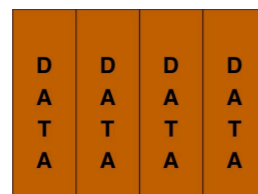
1/2 of the data left

$$\frac{1}{2^1}$$



1/4 of the data left

$$\frac{1}{2^2}$$



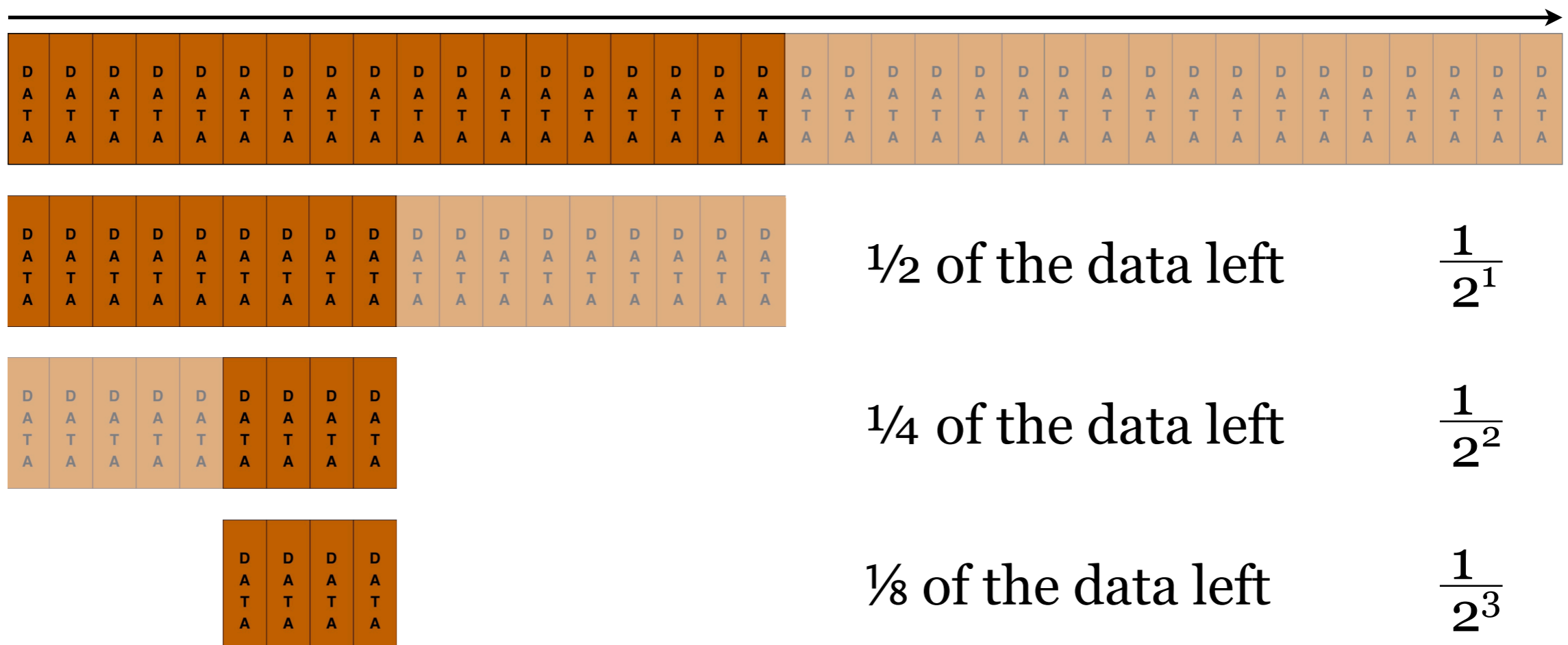
1/8 of the data left

$$\frac{1}{2^3}$$

Q: What's the average or — expected — case for n rows?

Binary Searching 9 billion people

We could pick from the middle. If that's not our target, then we exclude the *lower* or *upper* half of the data, depending on whether our target is greater or lesser than the value we picked. Then we pick the middle of the remaining half. Repeat.



Q: What's the average or — expected — case for n rows?

A: The expected case is $\log_2 n$, because we cut it in half each time.

Binary Searching 9 billion people

What if we could search through **sorted** data?

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.				
.				
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Binary Searching 9 billion people

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Now **that** is a better way!
 $33 < 4.5B$

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Binary Search is $O(\log_2 n)$.

Best case $O(1)$

Worst case $O(\log_2 n)$

Average case $O(\log_2 n)$

Recurrence:

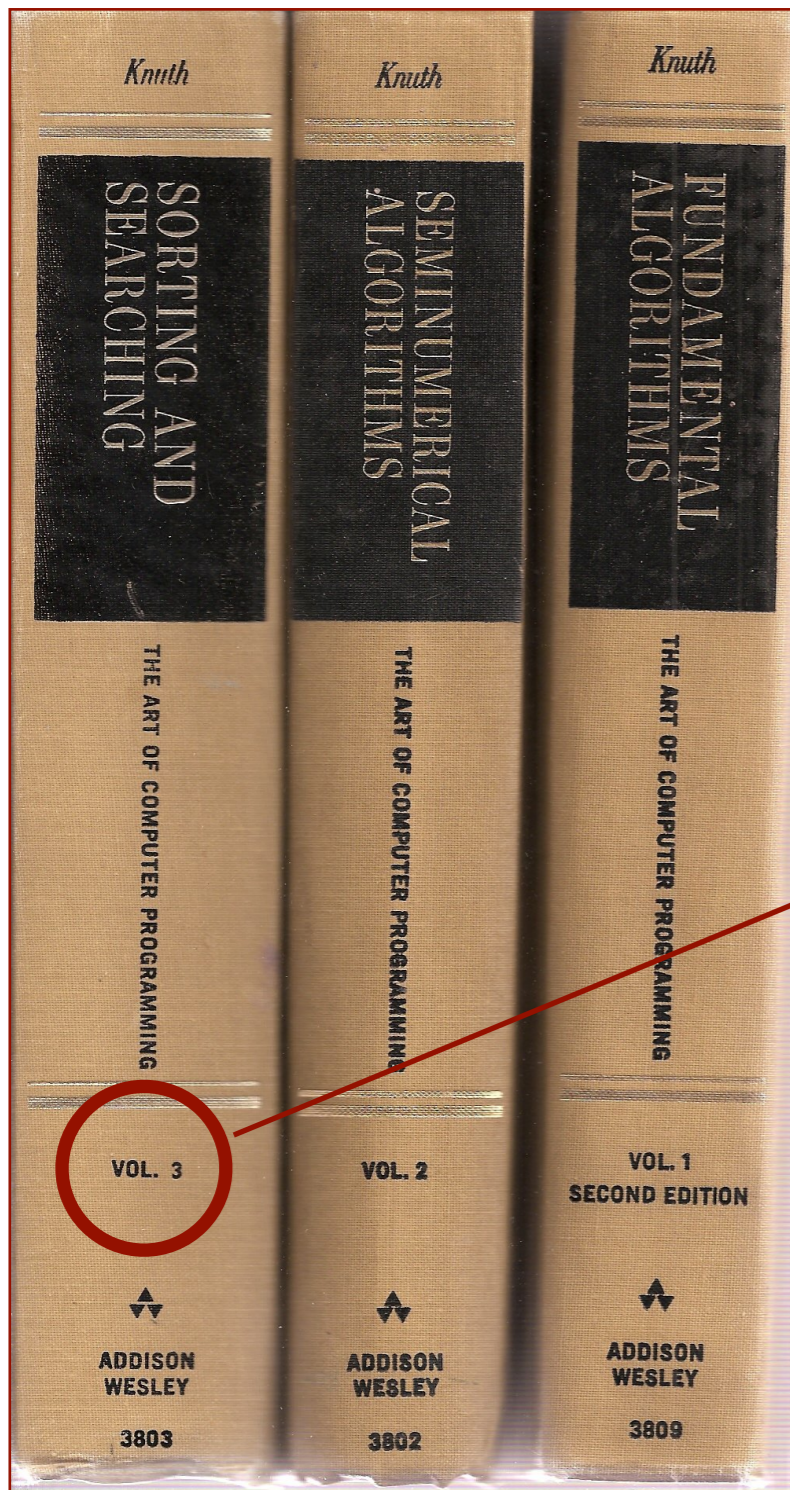
$$T(n) = T\left(\frac{n}{2}\right) + c$$

We could pick from the middle. If that's not our target, then we exclude the *lower* or *upper* half of the data, depending on whether our target is greater or lesser than the value we picked. Then we pick the middle of the remaining half. Repeat.

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A: The expected case is $\log_2 n$.

Binary Search Algorithm



6.2.1

SEARCHING AN ORDERED TABLE 407

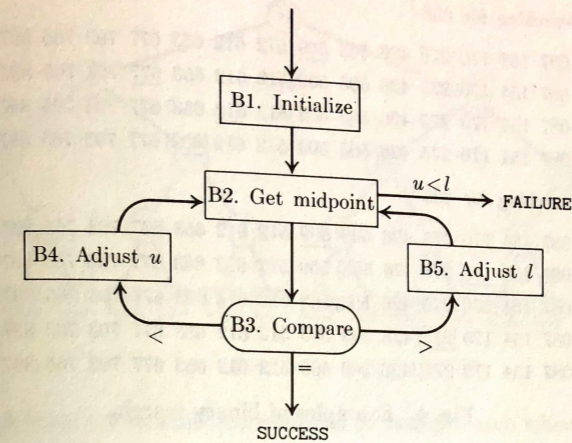


Fig. 3. Binary search.

which half of the table should be searched next, and the same procedure can be used again, comparing K to the middle key of the selected half, etc. After at most about $\log_2 N$ comparisons, we will have found the key or we will have established that it is not present. This procedure is sometimes known as “logarithmic search” or “bisection,” but it is most commonly called *binary search*.

Although the basic idea of binary search is comparatively straightforward, the details can be somewhat tricky, and many good programmers have done it wrong the first few times they tried. One of the most popular correct forms of the algorithm makes use of two pointers, l and u , which indicate the current lower and upper limits for the search, as follows:

Algorithm B (Binary search). Given a table of records R_1, R_2, \dots, R_N whose keys are in increasing order $K_1 < K_2 < \dots < K_N$, this algorithm searches for a given argument K .

B1. [Initialize.] Set $l \leftarrow 1, u \leftarrow N$.

B2. [Get midpoint.] (At this point we know that if K is in the table, it satisfies $K_l \leq K \leq K_u$. A more precise statement of the situation appears in exercise 1 below.) If $u < l$, the algorithm terminates unsuccessfully. Otherwise, set $i \leftarrow \lfloor (l + u)/2 \rfloor$, the approximate midpoint of the relevant table area.

B3. [Compare.] If $K < K_i$, go to B4; if $K > K_i$, go to B5; and if $K = K_i$, the algorithm terminates successfully.

B4. [Adjust u .] Set $u \leftarrow i - 1$ and return to B2.

B5. [Adjust l .] Set $l \leftarrow i + 1$ and return to B2. ■

Figure 4 illustrates two cases of this binary search algorithm: first to search for the argument 653, which is present in the table, and then to search for 400,

Binary Search Algorithm

Here is an iterative version of Binary Search from the CLRS text.

```
BINARY-SEARCH( $x, T, p, r$ )
1   $low = p$ 
2   $high = \max(p, r + 1)$ 
3  while  $low < high$ 
4       $mid = \lfloor (low + high) / 2 \rfloor$ 
5      if  $x \leq T[mid]$ 
6           $high = mid$ 
7      else  $low = mid + 1$ 
8  return  $high$ 
```

x - target
 T - collection of data
 p - start index
 r - stop index

CLRS 3e p.799

Binary Search Algorithm

```
proc BinarySearch(A, start, stop, target)
  midPoint = int((start+stop)/2) // round, ceil, floor?
  if (start > stop)
    return false
  else if (target == A[midPoint])
    return true
  else if (target < A[midPoint])
    BinarySearch(A, start, midPoint-1, target)
  else // target > A[midPoint] or not there at all
    BinarySearch(A, midPoint+1, stop, target)
  end if
end proc
```

Here is a recursive version of Binary Search, with some issues:

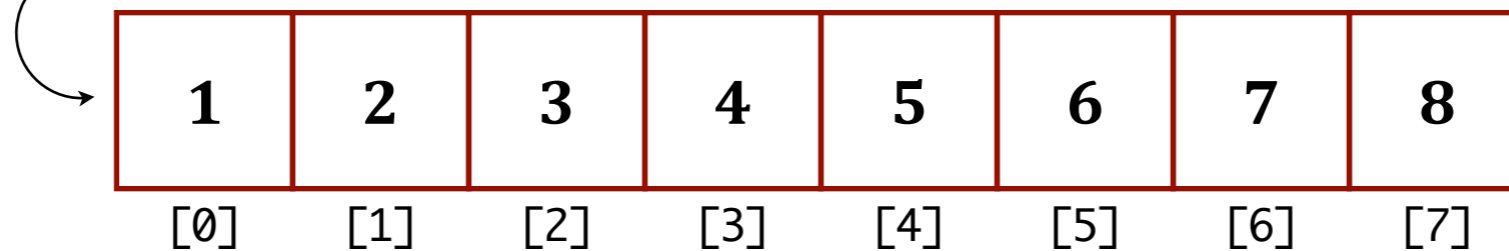
- (1) There are two exits. That's bad software craftsmanship.
- (2) It does not return **where** the target is found, just that it is.

Fix these issues when you program your own version.

Binary Search Example

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proc BinarySearch(A, start, stop, target)
  midPoint = int((start+stop)/2) // round, ceil, floor?
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end proc
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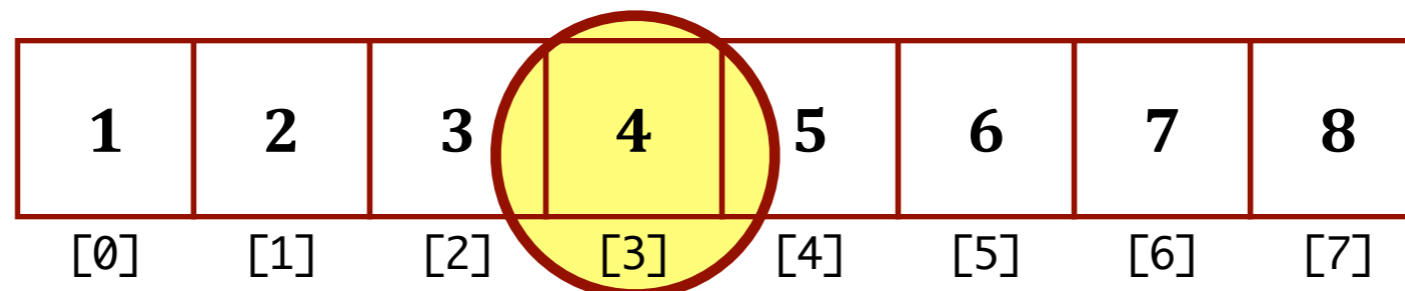
BinarySearch(A, 0, 7, 2)



Binary Search Example

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end proc
```

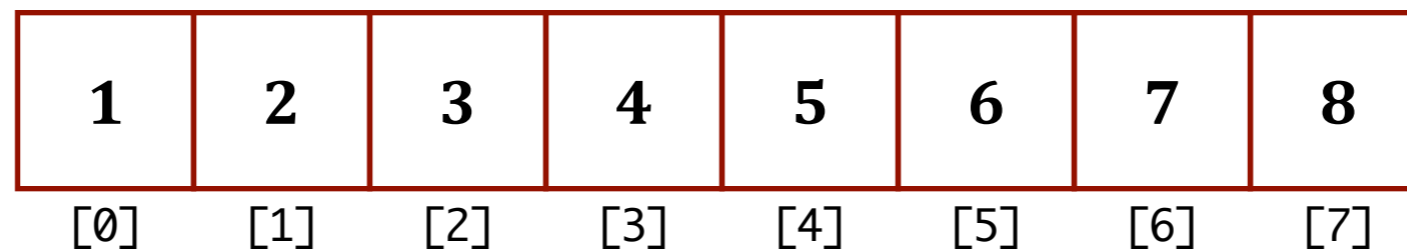
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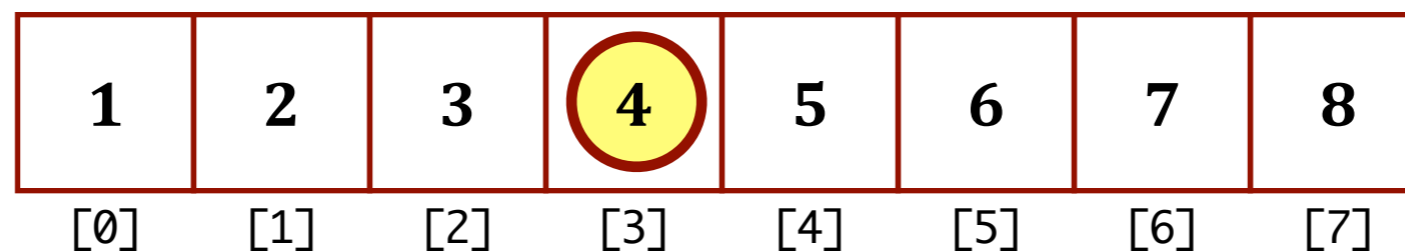
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BinarySearch(A, 0, 7, **2**) ≠



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BinarySearch(A, 0, 7, **2**) <



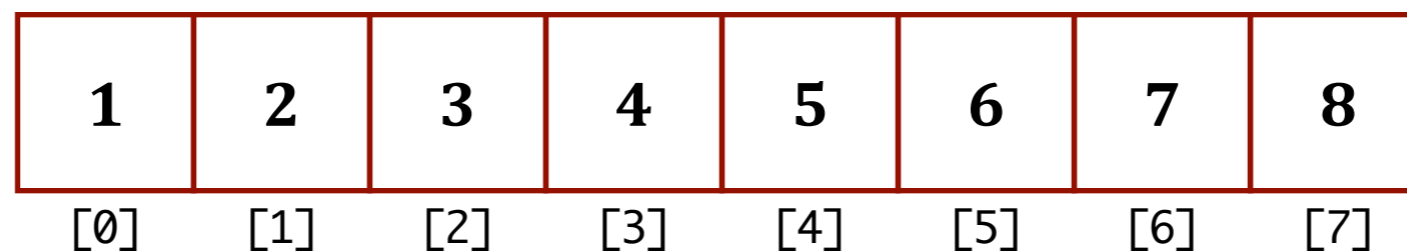
1	2	3	4	5	6	7	8
[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]

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  else if (target == A[midPoint])
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  else if (target < A[midPoint])
    BinarySearch(A, start, midPoint-1, target)
  else // target > A[midPoint] or not there at all
    BinarySearch(A, midPoint+1, stop, target)
  end if
end proc
```

BinarySearch(A, 0, 7, 2)
BinarySearch(A, 0, 2, 2)



Binary Search Example

```
proc BinarySearch(A, start, stop, target)
  midPoint = int((start+stop)/2) // round, ceil, floor?
  if (start > stop)
    return false
  else if (target == A[midPoint])
    return true
  else if (target < A[midPoint])
    BinarySearch(A, start, midPoint-1, target)
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end proc
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BinarySearch(A, 0, 7, 2)

BinarySearch(A, 0, 2, 2) → true

1	2	3	4	5	6	7	8
[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]

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```

Remember the issues:

- (1) There are two exits. That's bad software craftsmanship.
- (2) It does not return **where** the target is found, just that it is.

Fix these issues when you program your own version.