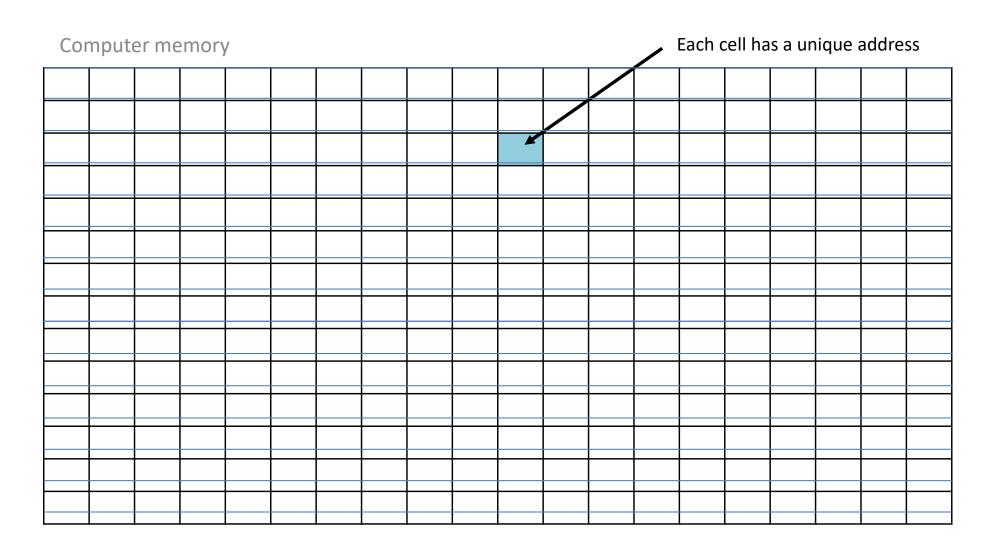
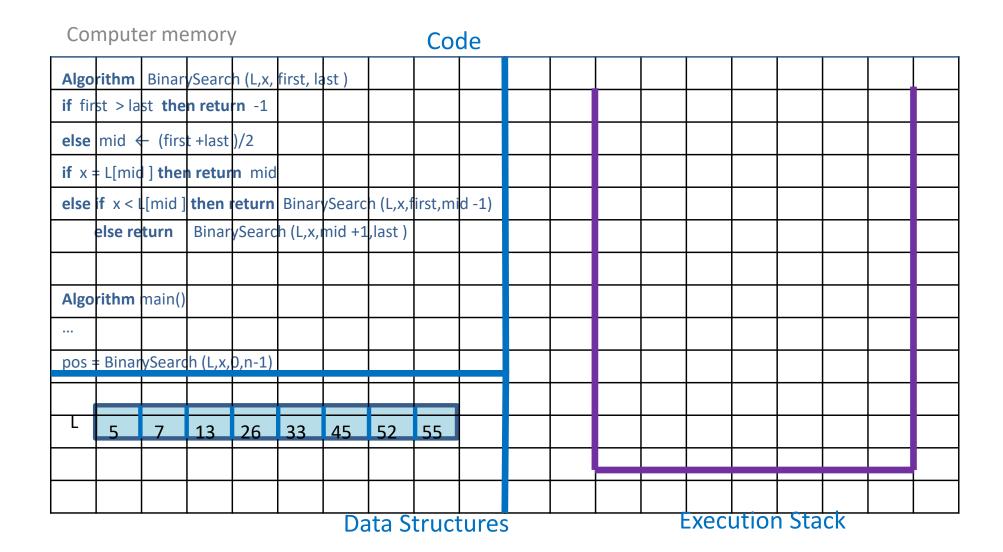
Execution of a Recursive Algorithm

We will illustrate how a computer executes a recursive algorithm using the recursive version of binary search as example.



Computer memory						Co	de							
Algo	rithm	Binar	/Searc	h (L,x,	first, la	ist)								
if fir	st > la	st the	n retu	rn -1										
else	mid 🗧	- (firs	t +last)/2										
if x	L[mic	l] ther	retur	n mid										
else	if x < l	.[mid]	then	return	Binar	ySearc	h (L,x,f	irst,mi	d -1)					
	else re	turn	Binar	ySearc	h (L,x,ı	mid +1	,last)							
Algo	rithm	main()												
•••														
pos :	Binar	ySearc	h (L,x,	0,n-1)										

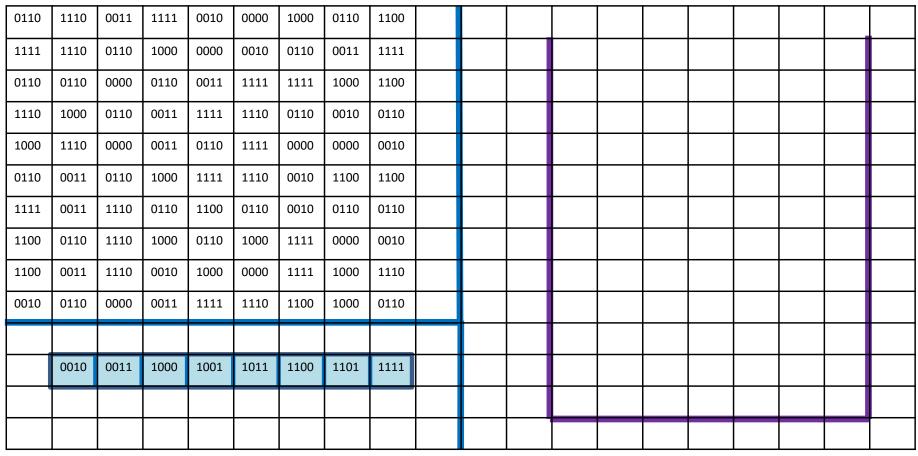
Computer memory						Co	de								
Algo	rithm	Binar	ySearc	h (L,x,	first, la	ıst)									
if fir	st > la	st the	n retu	rn -1											
else	mid 🗧	- (firs	t +last)/2											
if x =	L[mic	l] ther	ı retur	n mid											
else	if x < l	[mid]	then	eturn	Binar	/Searc	h (L,x,f	irst,mi	id -1)						
	else re	turn	Binar	ySearc	h (L,x,ı	mid +1	,last)								
Algo	rithm	main()													
pos =	Binar	ySearc	h (L,x,	0,n-1)											
L	5	7	13	26	33	45	52	55							
						Da	ta S	truct	tures	5					



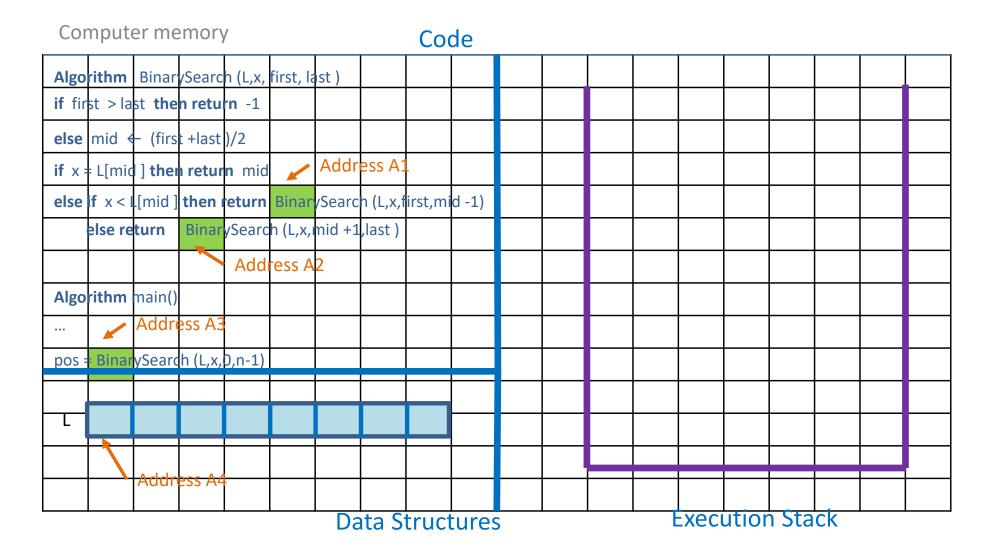
Information is Stored in Binary

Computer memory

Code



Data Structures



```
Algorithm BinarySearch (L,x, first, last)
if first > last then return -1
else mid \leftarrow (first +last )/2
if x = L[mid ] then return mid
else if x < L[mid] then return BinarySearch (L,x,first,mid-1) A1
    else return BinarySearch (L,x,mid +1,last ) A2
Algorithm main()
pos = BinarySearch (L,x,0,n-1) A0
                   23
                         26
                               35
                                      49
                                                  88
 x = 23
```

Activation Records

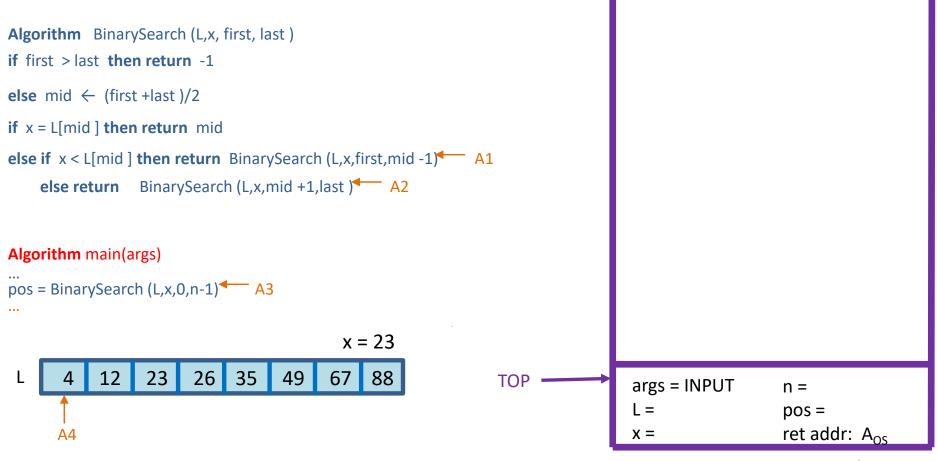
Every time that an algorithm (or method) is invoked an activation record is created at the top of the execution stack.

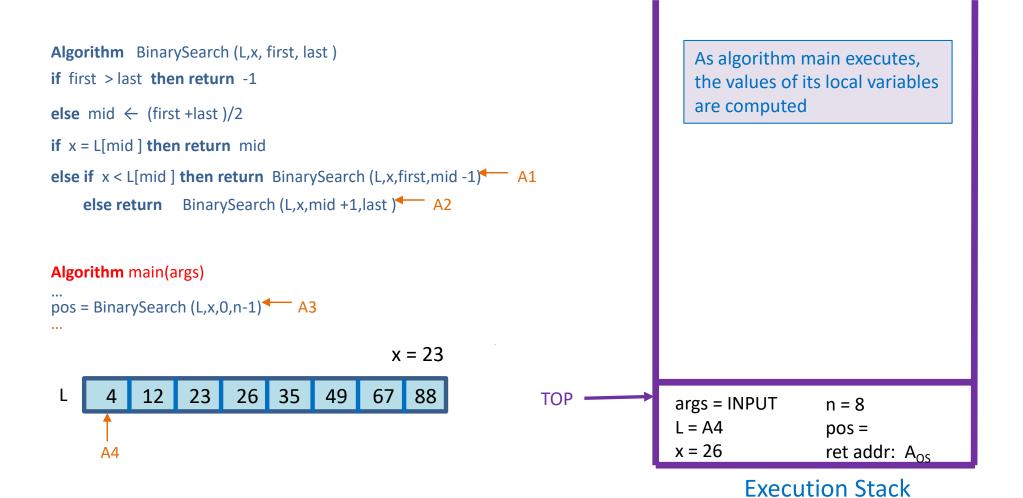
Activation Records

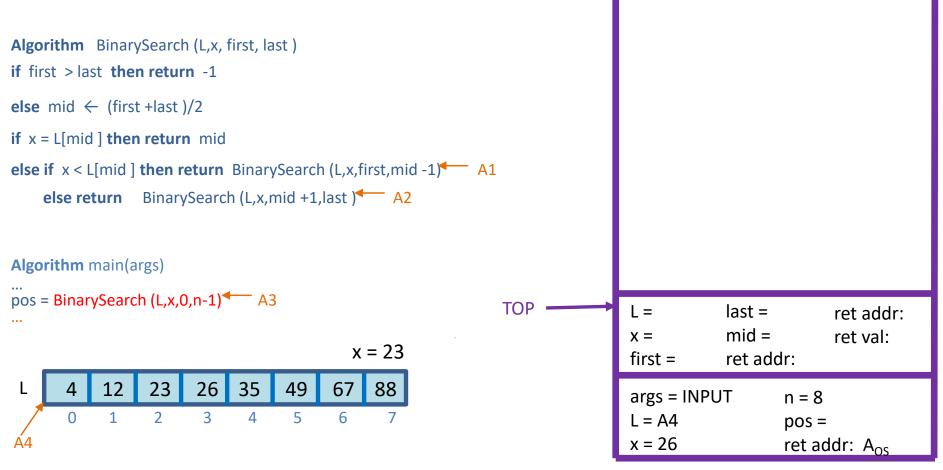
Every time that an algorithm (or method) is invoked an activation record is created at the top of the execution stack.

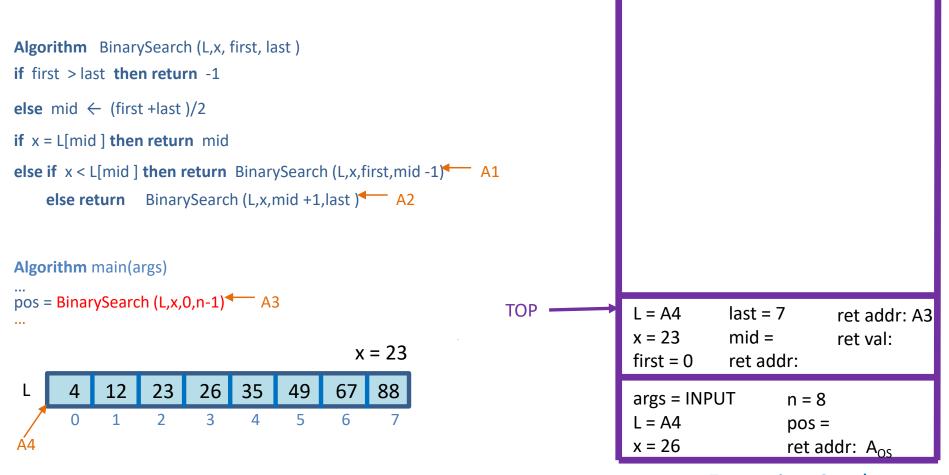
An activation record stores all the information that an algorithm needs to be executed:

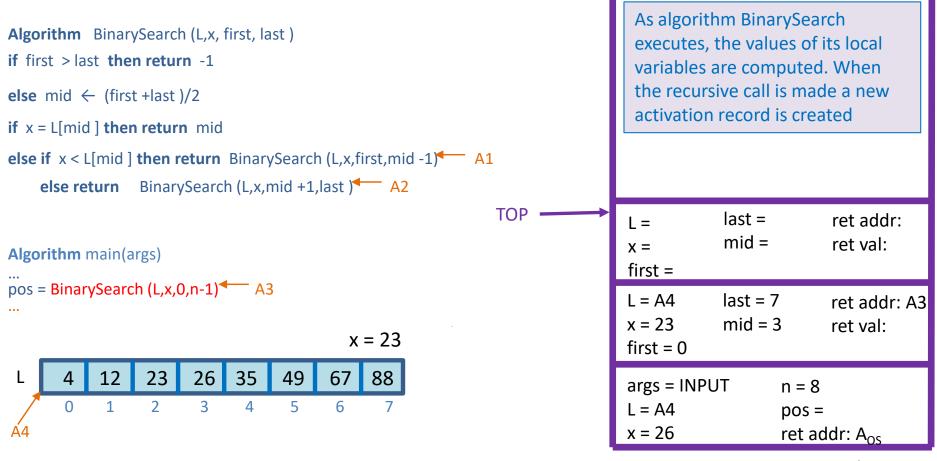
- parameters
- local variables
- return address
- return value (if any)

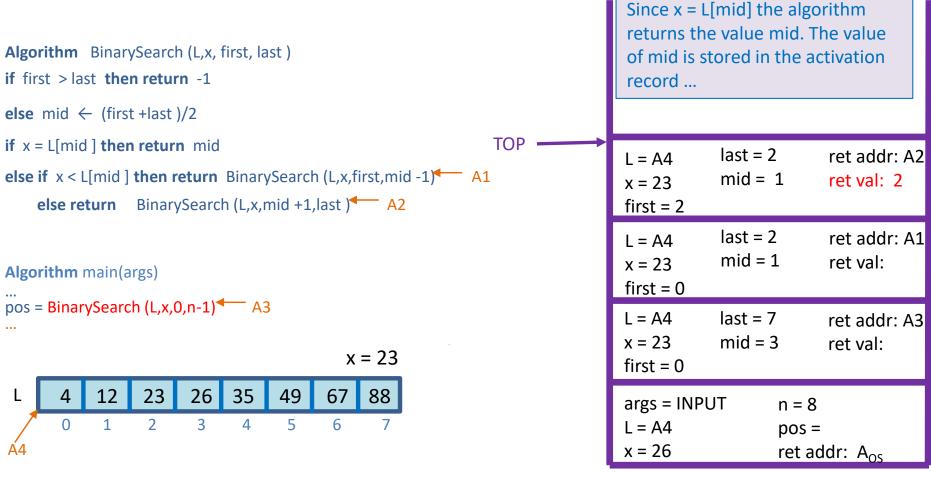


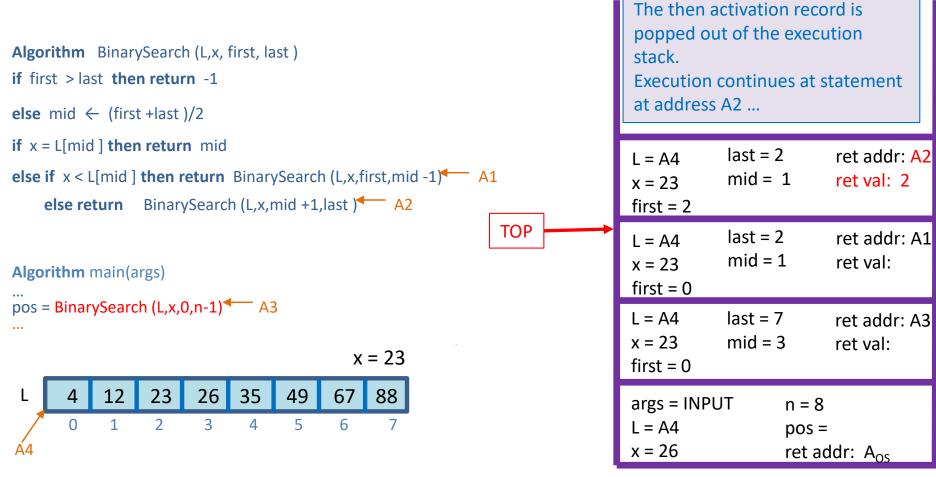


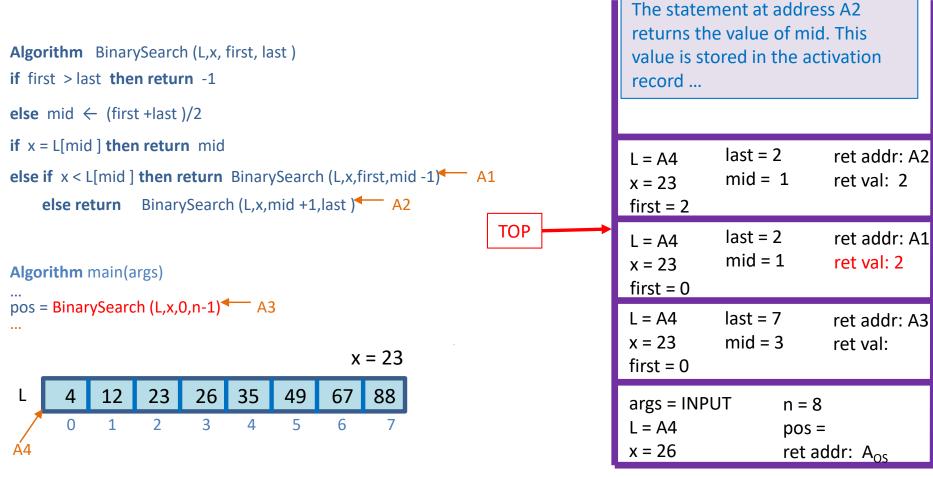


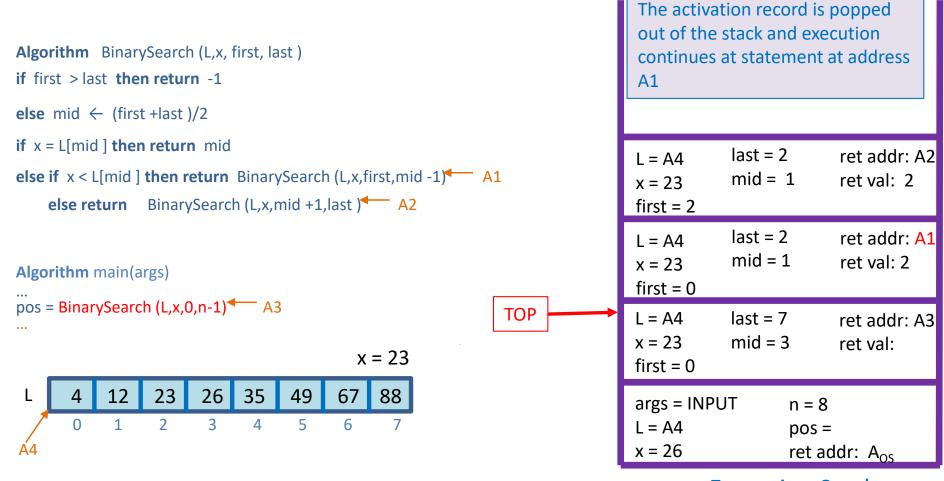


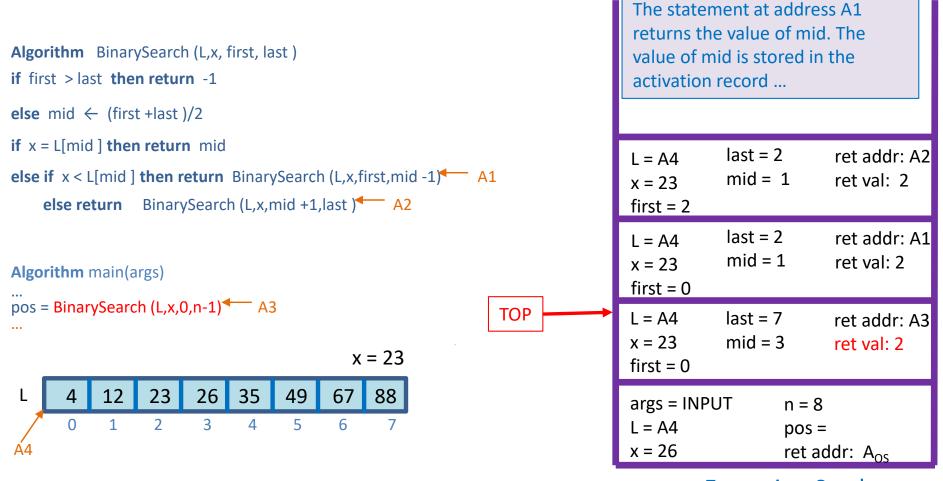


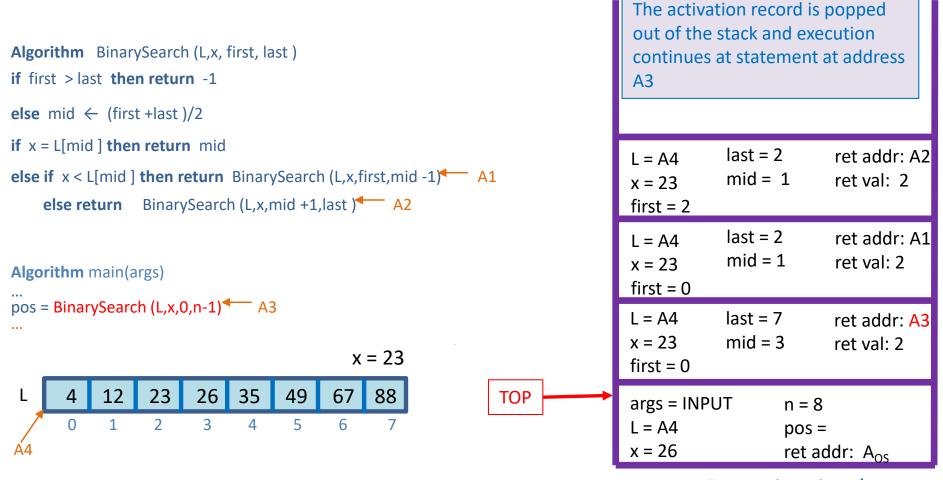


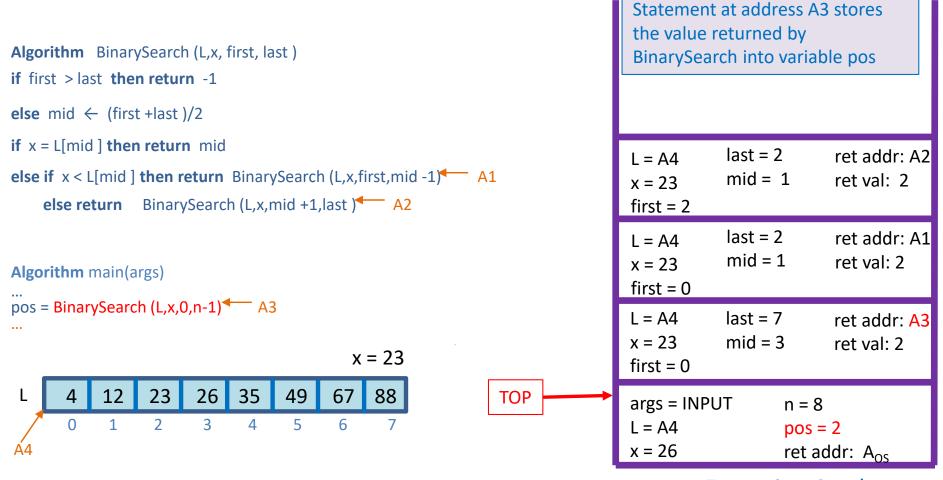












Algorithm BinarySearch (L,x, first, last) if first > last then return -1 else mid \leftarrow (first +last)/2 if x = L[mid] then return mid else if x < L[mid] then return BinarySearch (L,x,first,mid -1) \leftarrow A1 else return BinarySearch (L,x,mid +1,last) A2 **Algorithm** main(args) pos = BinarySearch (L,x,0,n-1) A3 x = 2323 26 35 49 88 12

The rest of algorithm main is executed. When the algorithm ends the activation record is popped out of the stack and control goes back to the operating system

L = A4 x = 23 first = 2	last = 2 mid = 1	ret addr: A2 ret val: 2							
L = A4 x = 23 first = 0	last = 2 mid = 1	ret addr: A1 ret val: 2							
L = A4 x = 23 first = 0	last = 7 mid = 3	ret addr: A3 ret val: 2							
arge - IND	args - INDLIT n - 9								

args = INPUT n = 8 L = A4 pos = 2x = 26 ret addr: A_{OS}

TOP

We will now compute the time complexity of binary search by first writing a recurrence equation for the time complexity function and then solving this equation using repeated substitution.

The worst case for binary search is when x is not in L. Let

f(n) = number of primitive operations performed by binary search in the worst case when the size of the input is n

We will compute f(n) for the base case and the recursive case.

Algorithm BinarySearch (L,x, first, last)
Input: Array L of size n and value x
Output: Index i, $0 \le i < n$, such that L[i] = x if x in L, or -1 if x not in L
if first > last then return -1
else mid \leftarrow L(first +last)/2 \rfloor if x = L[mid] then return mid
else if x < L[mid] then return BinarySearch (L,x,first,mid -1)
else return BinarySearch (L,x,mid +1,last)

In the base case the algorithm performs a constant number c_1 of primitive operations. Note that in the base case fist > last, so the number of elements n is 0:

$$f(0) = c_1$$

```
Algorithm BinarySearch (L,x, first, last )
Input: Array L of size n and value x
Output: Index i, 0 \le i < n, such that L[i] = x if x in L, or -1 if x not in L
if first > last then return -1
else mid \leftarrow L(first + last)/2J
if x = L[mid] then return mid
else if x < L[mid] then return BinarySearch (L,x,first,mid -1)
else return BinarySearch (L,x,mid +1,last)
```

Ignoring the recursive calls, when n > 0 the algorithm performs a constant number c_2 of primitive operations ...

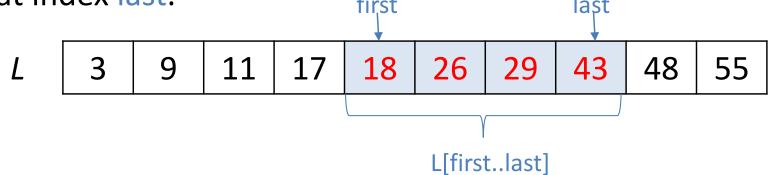
$$f(n) = c2 + ...$$
 for $n > 0$

We need to add to this the number of operations performed by the recursive calls.

```
Algorithm BinarySearch (L,x, first, last )
Input: Array L of size n and value x
Output: Index i, 0 ≤ i < n, such that L[i] = x if x in L, or -1 if x not in L
if first > last then return -1
else mid ←L(first +last )/2」
if x = L[mid] then return mid
else if x < L[mid] then return BinarySearch (L,x,first,mid -1)
else return BinarySearch (L,x,mid +1,last )</pre>
```

Let L[first..last] denote the part of array L that starts at index first and ends at index last.

| First | Let L[first..last] | Let L[



If the number of elements in L is n and the first recursive call is made, the number of elements in the first half of the array is (n-1)/2, so the number of primitive operations performed by the first recursive call is

f((n-1)/2)

Similarly, if the second recursive call is made, the number of elements in the second half of the array is (n-1)/2, so the number of primitive operations performed by the second recursive call is also

f((n-1)/2)

```
Algorithm BinarySearch (L,x, first, last)
Input: Array L of size n and value x
Output: Index i, 0 ≤ i < n, such that L[i] = x if x in L, or -1 if x not in L

if first > last then return -1
else mid ← L(first + last)/2 J

if x = L[mid] then return mid
else if x < L[mid] then return BinarySearch (L,x,first,mid-1)

else return BinarySearch (L,x,mid +1,last)

f((n-1)/2)
```

So, the number of primitive operations performed by the algorithm is

$$f(0) = c_1$$

 $f(n) = c_2 + f((n-1)/2)$ for $n > 0$

This equation is called a recurrence equation.

Solving the Recurrence Equation using Repeated Substitution

$$f(0) = c_1 \tag{1}$$

$$f(n) = f\left(\frac{n-1}{2}\right) + c_2 \tag{2}$$

Start with equation (2):

$$f(n) = f\left(\frac{n-1}{2}\right) + c_2$$
 Use (2) to compute $f\left(\frac{n-1}{2}\right)$

$$f\left(\frac{n-1}{2}\right) = f\left(\frac{n-1}{2}-1\right) + c_2 = f\left(\frac{n-1-2}{2}\right) + c_2 = f\left(\frac{n-2^0-2^1}{2^2}\right) + c_2 \quad \text{Use (2) to compute } f\left(\frac{n-2^0-2^1}{2^2}\right)$$

$$f\left(\frac{n-2^{0}-2^{1}}{2^{2}}\right) = f\left(\frac{\frac{n-2^{0}-2^{1}}{2^{2}}-1}{2}\right) + c_{2} = f\left(\frac{n-2^{0}-2^{1}-2^{2}}{2^{3}}\right) + c_{2}$$
 And so on ...

$$f\left(\frac{n-2^{0}-2^{1}-2^{2}}{2^{3}}\right) = f\left(\frac{n-2^{0}-2^{1}-2^{2}-2^{3}}{2^{4}}\right) + c_{2}$$

•

$$f\left(\frac{n-2^{0}-2^{1}-2^{2}-...-2^{j}}{2^{j+1}}\right) = f\left(\frac{n-2^{0}-2^{1}-2^{2}-...-2^{j}}{2^{j+2}}\right) + c_{2} = c_{1} + c_{2}$$

$$= 0 \qquad \qquad f(0) = c_{1}$$

Now we substitute each equation into the equation above it

Solving the Recurrence Equation using Repeated Substitution

$$f(0) = c_1 \tag{1}$$

$$f(n) = f\left(\frac{n-1}{2}\right) + c_2 \tag{2}$$

Start with equation (2):

$$f(n) = f\left(\frac{n-1}{2}\right) + c_2$$

$$f\left(\frac{n-1}{2}\right) = f\left(\frac{n-1}{2}-1\right) + c_2 = f\left(\frac{n-1-2}{2^2}\right) + c_2 = f\left(\frac{n-2^0-2^1}{2^2}\right) + c_2 \quad \text{Use (2) to compute } f\left(\frac{n-2^0-2^1}{2^2}\right)$$

$$f\left(\frac{n-2^{0}-2^{1}}{2^{2}}\right) = f\left(\frac{n-2^{0}-2^{1}}{2^{2}}-1\right) + c_{2} = f\left(\frac{n-2^{0}-2^{1}-2^{2}}{2^{3}}\right) + c_{2}$$
 And so on ...

$$f\left(\frac{n-2^{0}-2^{1}-2^{2}}{2^{3}}\right) = f\left(\frac{n-2^{0}-2^{1}-2^{2}-2^{3}}{2^{4}}\right) + c_{2}$$

$$f\left(\frac{n-2^{0}-2^{1}-2^{2}-...-2^{j}}{2^{j+1}}\right) = f\left(\frac{n-2^{0}-2^{1}-2^{2}-...-2^{j+1}}{2^{j+2}}\right) + c_{2} = c_{1} + c_{2}$$

We get

$$f(n) = c_2 + c_2 + c_2 + ... + c_2 + c_1 = (j+2) c_1$$

Since $n-2^0-2^1-2^2-...-2^{j+1}=0$ then $n=2^0+2^1+2^2+...2^{j+1}=2^{j+2}-1$. Taking logarithms on both sides we get $\log_2(n+1)=j+2$, therefore $f(n)=c_1\log_2(n+1)$

Using the rules we learned for computing the order of functions we finally get that f(n) is $O(\log n)$

Comparing Time Complexities

Linear search

$$f(n)$$
 is $O(n) = \{t(n) | t(n) \le c \text{ n for all } n \ge n_0, n_0, c \text{ constants}\}$

Binary search

$$f(n)$$
 is $O(\log n) = \{t(n) | t(n) \le c \log n \text{ for all } n \ge n_0, n_0, c \text{ const}\}$

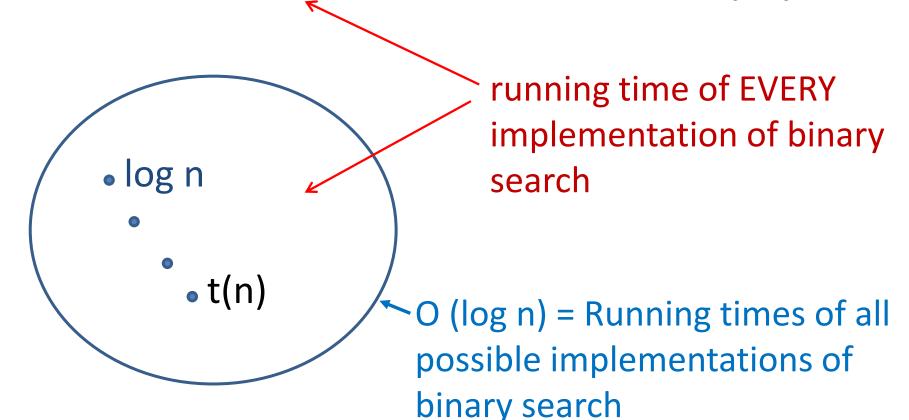
running time of EVERY implementation of binary search

Comparing Time Complexities

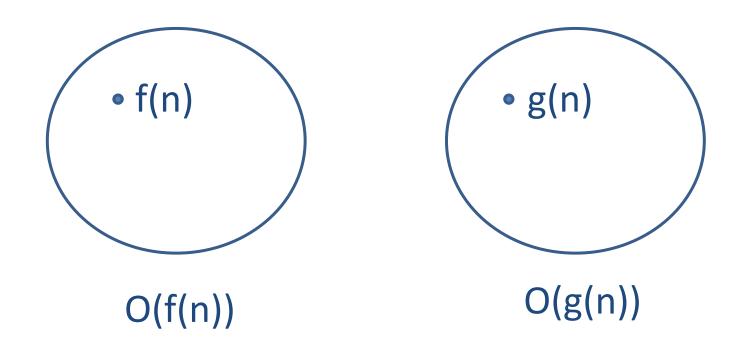
Linear search

f(n) is $O(n) = \{t(n) | t(n) \le c \text{ n for all } n \ge n_0, n_0, c \text{ constants} \}$ Binary search

f(n) is $O(\log n) = \{t(n) \mid t(n) \le c \log n \text{ for all } n \ge n_0, n_0, c \text{ const}\}$



Algorithm A has complexity O(f(n))
Algorithm B has complexity O(g(n))
Which algorithm is faster?

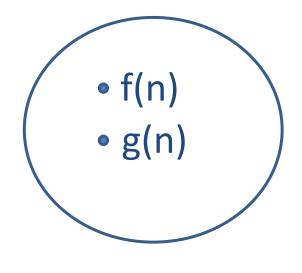


Algorithm A has complexity O(f(n))

Algorithm B has complexity O(g(n))

Two cases:

• f(n) is O(g(n)) and g(n) is O(f(n))

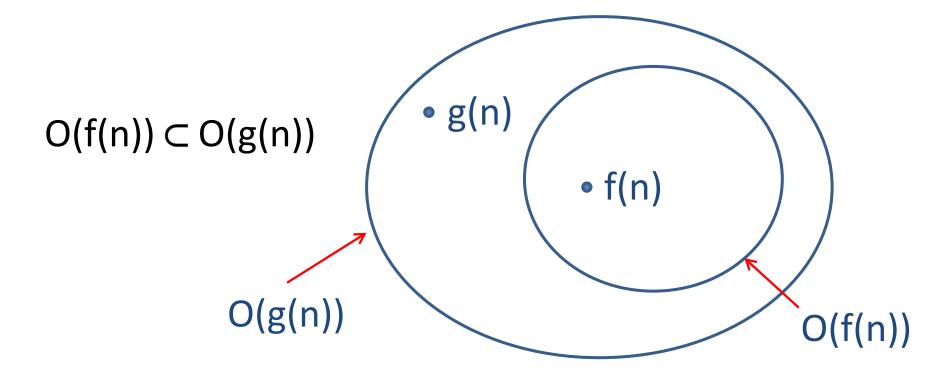


O(f(n)) = O(g(n))

Both algorithms have the same set of possible running times

Algorithm A has complexity O(f(n))
Algorithm B has complexity O(g(n))
Two cases:

• f(n) is O(g(n)) and g(n) is **not** O(f(n))

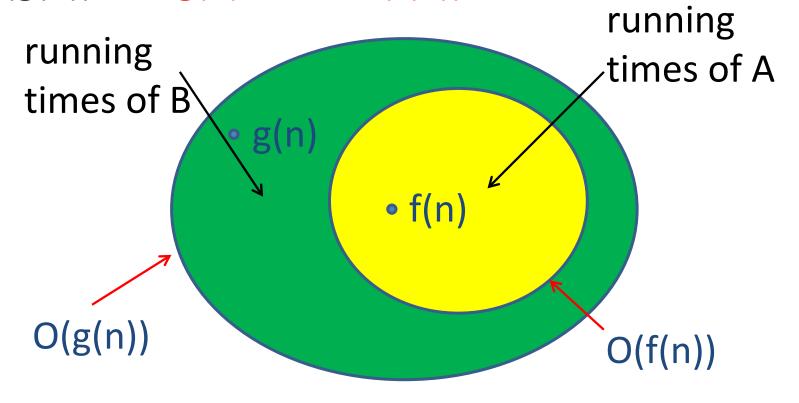


Algorithm A has complexity O(f(n))

Algorithm B has complexity O(g(n))

Two cases:

• f(n) is O(g(n)) and g(n) is **not** O(f(n))



Algorithm A has complexity O(f(n))

Algorithm B has complexity O(g(n))

Two cases:

 f(n) is O(g(n)) and g(n) is not O(f(n)): B is slower than A in ALL running **implementations** times of B g(n)g(n) > c f(n) for $n \ge n_0$ for all c, n_0 , i.e. all • f(n) implementations O(g(n))

O(f(n))

Complexity Classes

$$O(1)$$
 \subset $O(\log n)$ \subset $O(n)$ \subset $O(n \log n)$ constant logarithmic linear

$$\subset$$
 $O(n^2)$ \subset $O(n^a)$ \subset quadratic polynomial (constant $a > 2$)

○(bⁿ)
 exponential
 (b constant)

$$\subseteq$$
 O(n!) \subseteq O(nⁿ) ...

Efficient algorithms