

## Home Work 10

The problems in this problem set cover lectures C11 and C12

1.
  - a. Define a recursive binary search algorithm.

```
If lb > ub
    Return -1
else
    Mid := (lb+ub)/2
    If Array(Mid) = element
        Return Mid
    Elsif Array(Mid) < Element
        Return Binary_Search(Array, mid+1, ub, Element)
    Else
        Return Binary_Search(Array, lb, mid-1, Element)
    End if
End if
```

- b. Implement your algorithm as an Ada95 program.

```
46. function Binary_Search (My_Search_Array : My_Array; Lb : Integer; Ub: Integer; Element : Integer)
return Integer is
47.   mid : integer;
48. begin
49.   if (Lb> Ub) then
50.     return -1;
51.   else
52.     Mid := (Ub+Lb)/2;
53.     if My_Search_Array(Mid) = Element then
54.       return(Mid);
55.     elsif My_Search_Array(Mid) < Element then
56.       return (Binary_Search(My_Search_Array, Mid+1, Ub, Element));
57.     else
58.       return (Binary_Search(My_Search_Array, Lb, Mid-1, Element));
59.     end if;
60.   end if;
61.
62. end Binary_Search;
63. end Recursive_Binary_Search;
```

- c. What is the recurrence equation that represents the computation time of your algorithm?

**Recursive Binary Search**

<code>if (Lb &gt; Ub) then</code>	<b>Cost</b>
<code>return -1;</code>	c1
<code>else</code>	c2
<code>Mid := (Ub+Lb)/2;</code>	c3
<code>if My_Search_Array(Mid) = Element then</code>	c4
<code>return(Mid);</code>	c5
<code>elseif My_Search_Array(Mid) &lt; Element then</code>	c6
<code>return (Binary_Search(My_Search_Array, Mid+1, Ub, Element));</code>	c7
<code>else</code>	T(n/2)
<code>return (Binary_Search(My_Search_Array, Lb, Mid-1, Element));</code>	c8
<code>end if;</code>	T(n/2)
<code>end if;</code>	c9
	c10

In this case, only one of the recursive calls is made, hence only one of the T(n/2) terms is included in the final cost computation.

Therefore  $T(n) = (c1+c2+c3+c4+c5+c6+c7+c8+c9+c10) + T(n/2)$   
 $= T(n/2) + C$

- d. What is the Big-O complexity of your algorithm? Show all the steps in the computation based on your algorithm.

$T(n) = T(n/2) + C$

$\forall T(n) = aT(n/b) + cn^k,$   
 where  $a, c > 0$  and  $b > 1$

$T(n) = \left\{ \begin{array}{l} O(n^{\log_b a}) \quad a > b^k \\ O(n^k \log_b n) \quad a < b^k \\ O(n^k) \quad a = b^k \end{array} \right.$   $1 = 2^0$ , hence the second term is used,

$T(n) =$

2. What is the Big-O complexity of :  
 a. Heapify function

A heap is an array that satisfies the heap properties i.e.,  $A(i) \leq A(2i)$  and  $A(i) \leq A(2i+1)$ .

The heapify function at 'i' makes A(i .. n) satisfy the heap property, under the assumption that the subtrees at A(2i) and A(2i+1) already satisfy the heap property.

<b>Heapify function</b>	<b>Cost</b>
Lchild := Left(I);	c1
Rchild := Right(I);	c2
if (Lchild <= Heap_Size and Heap_Array(Lchild) > Heap_Array(I))	c3
Largest:= Lchild;	c4
else	c5
Largest := I;	c6
if (Rchild <= Heap_Size)	c7
if Heap_Array(Rchild) > Heap_Array(Largest)	c8
Largest := Rchild;	c9
if (Largest /= I) then	c10
Swap(Heap_Array, I, Largest);	c11
Heapify(Heap_Array, Largest);	T(2n/3)

$$T(n) = T(2n/3) + C'$$

$$= T(2n/3) + O(1)$$

a = 1, b = 3/2, f(n) = 1, therefore by master theorem,

$$T(n) = O(n^{\log_b a} \log n)$$

$$= O(n^{\log_{3/2} 1} \log n)$$

$$= O(1 * \log n)$$

$$= O(\log n)$$

The important point to note here is the T(2n/3) term, which arises in the worst case, when the heap is asymmetric, i.e., the right subtree has one level less than the left subtree (or vice-versa).

b. Build\_Heap function

<b>Code</b>	<b>Cost t(n)</b>
Heap_Size := Size;	c1
for I in reverse 1 .. (Size/2) loop	n/2+1
Heapify(Heap_Array, I);	(n/2) log n
end loop;	n/2

Therefore

$$T(n) = c1 + n/2 + 1 + (n/2) \log n + n/2$$

$$= (n \log(n))/2 + n + (c1 + 1)$$

Simplifying

$$\Rightarrow T(n) = O(n \log(n))$$

c. Heap\_Sort

### Heap Sort

### Cost t(n)

```
Build_Heap(Heap_Array, Size);
```

$O(n \log n)$

```
for I in reverse 2.. size loop
```

n

```
  Swap(Heap_Array, I, 1);
```

$c_1(n-1)$

```
  Heap_Size := Heap_Size - 1;
```

$c_2(n-1)$

```
  Heapify(Heap_Array, 1);
```

$O(\log n)(n-1)$

$$\begin{aligned} T(n) &= 2 O(n \log n) + (c_1 + c_2 + 1)n - O(\log n) + \\ &= 2 O(n \log n) - O(\log n) + c'n \end{aligned}$$

Simplifying,

$$\Rightarrow T(n) = O(n \log n)$$