

Application of Divide and Conquer

Binary Search

Binary Search :

input : An array of n elements and what element "x" we want to search in an array

output : Position of an element x if it is found and if it is not present in the array then our function will return -1

Implementation of Binary Search :

0 1 2 3 4 #index

2 4 6 8 10 #values

i = 0

j = 4

x => That we want to search in an array = 8

mid = $(0 + 4)/2 = 2$

a[mid] = a[2] = 6 == 8 -> false

a[mid] = a[2] = 6 < 8 -> true

BinarySearch(a,mid+1,j,x);

i = 3

j = 4

mid = $(3+4)/2 = 7/2 = 4$

a[mid] = a[4] = 10 == 8 -> false

a[mid] = a[4] = 10 > 8

BinarySearch(a,i,mid-1,x);

i = 3

j = 3

i == j (3 == 3)

a[i] = a[3] = 8 == 8 - true

Return -> 3

BinarySearch(a,i,j,x):

 if(i == j):

 // small problem

 if(a[i] == x):

 return i # O(1)

 return -1

// Big problem -> we apply Divide and Conquer

Strategy

 while(i < j):

 mid = (i + j)/2 #O(1)

 if(a[mid] == x):

 return mid # O(1)

 if(a[mid] < x):

 BinarySearch(a,mid+1,j,x) #T(n/2)

 elif(a[mid] > x)

 BinarySearch(a,i,mid-1,x) # T(n/2)

 return -1;

Recurrence Relation :

$T(n) = T(n/2) + c \rightarrow$ Binary Search Algorithm

$$a = 1$$

$$b = 2$$

$$n^{(\log_b a)} = n^{(\log_2 1)} = n^0 = 1$$

$$f(n) = c$$

Which one is greater?

Both are equal i.e. constant

Overall time complexity is : $O(f(n) \log n) \Rightarrow O(c \log n) \Rightarrow O(\log n)$

Discussion about Best case, worst case and average case time

complexity Best case : $O(1)$

Worst case : $O(\log n)$

Average case : $O(\log n)$