

$$1. \quad T(n) = a T(n/b) + O(n^d)$$

$a$  = number of subproblems

$n/b$  = size of subproblem

$n^d$  = time needed to conquer

when  $d = \log_b a$  then  $T(n)$  is  $O(n^d \log_b n)$

"  $d > \log_b a$  then  $T(n) = O(n^d)$

"  $d < \log_b a$  then  $T(n) = O(n^{\log_b a})$

(a)  $a = 5$

$$n/b = n/2$$

$$n^d = n$$

$$T(n) = 5T(n/2) + O(n)$$

$$d = 1$$

$$\log_2 5 = 2.32$$

$$T(n) = O(n^{2.32} \log_2 n)$$

$$d < \log_b a$$

$$\therefore T(n) = O(n^{\log_2 5})$$

$$= O(n^{2.32})$$

(b)  $a = 2$

$$n/b = n-1$$

$$n^d = 1$$

$$T(n) = 2T(n-1) + O(1)$$

say,

$$T(1) = 2T(1-1) + O(1)$$

$$= 0$$

$$T(2) = 2T(1)$$

$$= 2 \times T(1)$$

$$T(3) = 2T(2)$$

$$= 2 \times 2T(1)$$

$$\therefore T(n) = 2^n T(1)$$

$$(c) \quad a = 9$$

$$n/b = n/3$$

$$O(n^d) = O(n^2)$$

$$T(n) = 9T(n/3) + O(n^2)$$

$$b = 3$$

$$d = 2$$

$$\log_b a = \log_3 9 = 2$$

$$b > \log_b a$$

$$\therefore T(n) = O(n^d)$$

$$= O(n^2)$$

$$d = \log_b a$$

$$\therefore T(n) = O(n^d \log_b n)$$

$$= O(n^2 \log_3 n)$$

$\therefore$  I would choose algorithm B, because its time complexity is minimum.

2. (a) we split array A into 2 subproblems subarrays.  $A_1$  and  $A_2$  of half of the size. After that we do a linear time equality operation to decide whether it is possible to find a majority element. The recurrence therefore given by  $T(n) = 2T(n/2) + O(n)$ . The complexity of algorithm comes to  $O(n \log n)$ .

GetMajorElement( $a[1 \dots n]$ )

Input = Array  $a$  of objects.

Output = Majority element of  $a$

If  $n = 1$  : return  $a[1]$

$k = n/2$

$elem_{lsub} = \text{GetMajorElement}(a[1 \dots k])$

$elem_{rsub} = \text{GetMajorElement}(a[k+1 \dots n])$

If  $elem_{lsub} = elem_{rsub}$ :

return  $elem_{lsub}$

$lcount = \text{GetFrequency}(a[1 \dots n], elem_{lsub})$

$rcount = \text{GetFrequency}(a[k+1 \dots n], elem_{rsub})$

If  $lcount > k+1$

return  $elem_{lsub}$

else if  $\text{count} > k+1$ :

return  $\text{elem}_{\text{sub}}$

else return NoMajorityElement :

(b) Recurrence relation for the algorithm is given below

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

The processing of array in the recursive function

is done in  $O(n)$  time.

GetMajorityElement ( $a[1 \dots n]$ )

Input : Array  $a$  of objects

Output : Majority element of  $a$

If  $n \geq 2$ :

If  $a[1] = a[2]$  return  $a[1]$

else

return No-Majority-Element;

For  $i = 1$  to  $n$ :

If  $(a[i] = a[i+1])$ :

Insert  $a[i]$  into temp

$i = i+1$ :

return GetMajorityElement(temp)

Checkcan( $a[1 \dots n]$ )

Input : Array  $a$  of objects

Output : majority of a element of  $a$

$m = \text{Getmajorelement}(a[1 \dots n])$

$\text{freq} = \text{Getfrequency}(a[1 \dots n], m)$

If  $\text{freq} > \lfloor \frac{n}{2} \rfloor + 1$ :

return  $m$ ;

else

return no-majority-element.

3.

$$a = 68$$

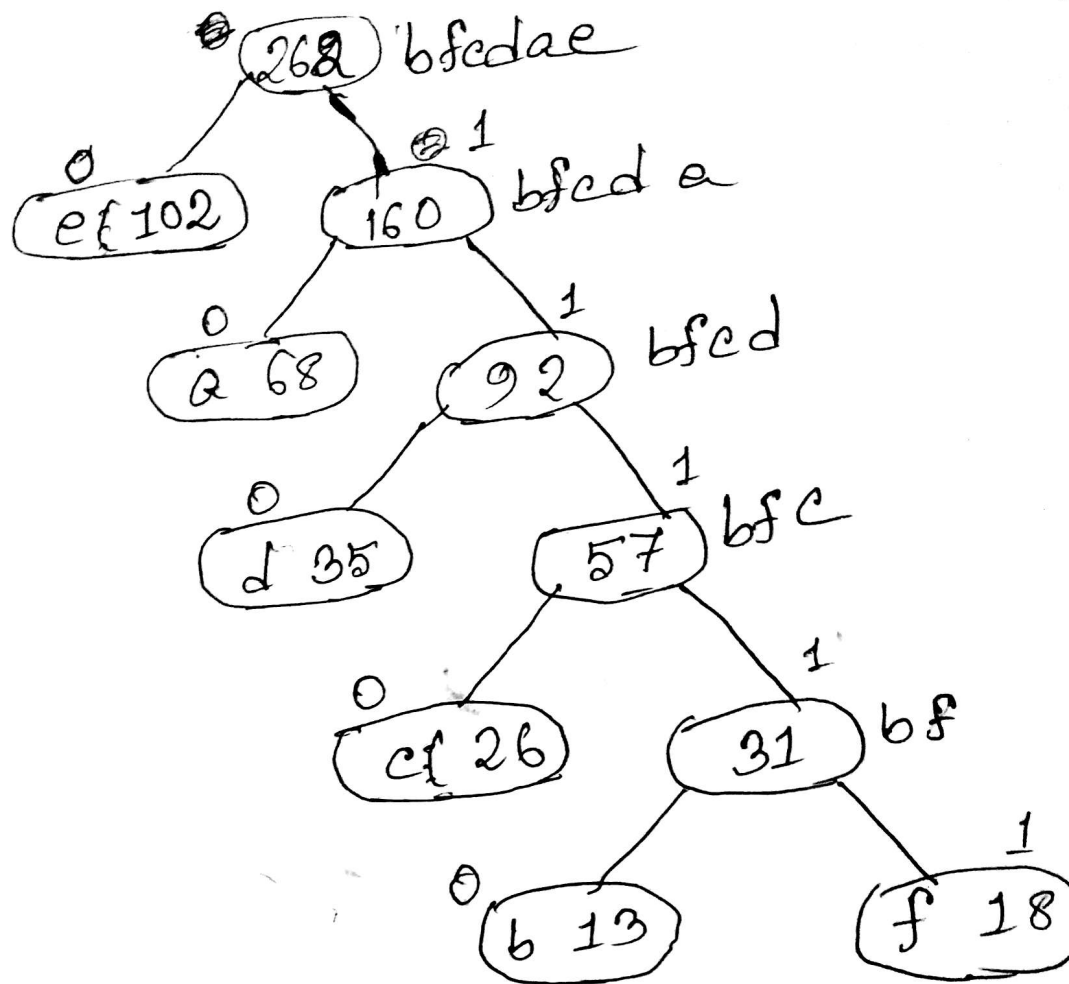
$$b = 13$$

$$c = 26$$

$$d = 35$$

$$e = 102$$

$$f = 18$$



$$a = 10$$

$$b = 11110$$

$$c = 1110$$

$$d = 110$$

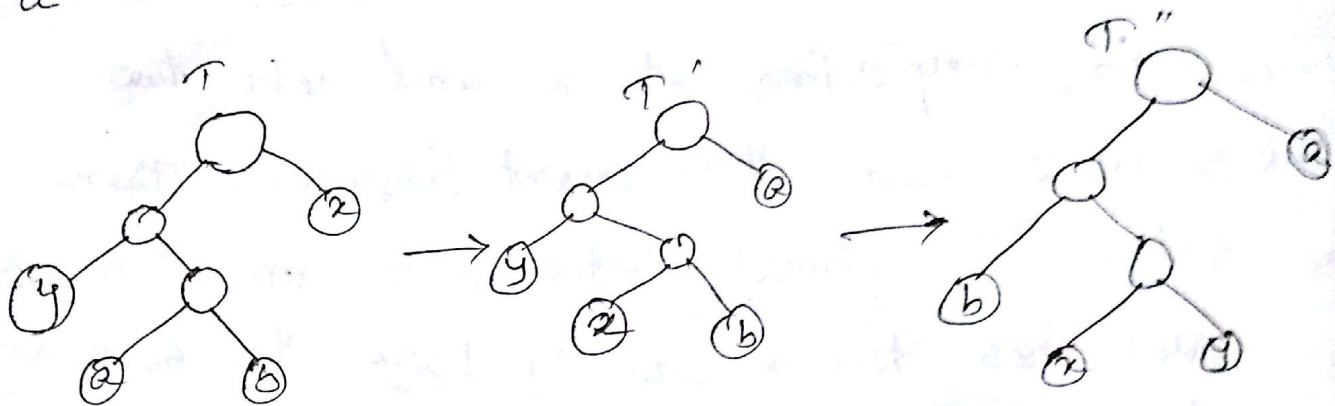
$$e = 0$$

$$f = 11111$$

Let  $C$  be an alphabet in which each character  $c \in C$  has frequency  $c.\text{freq}$ . Let  $x$  and  $y$  be two characters in  $C$  having the lowest frequencies. Then there exists an optimal prefix code for  $C$  in which the codewords for  $x$  and  $y$  have the same length and differ only in the last bit.

Proof: The idea of the proof is to take a tree  $T$  representing an arbitrary optimal prefix code such that the characters  $x$  and  $y$  appear as sibling leaves of maximum depth in the new tree. Let  $a$  and  $b$  be two characters that are sibling leaves of maximum depth in  $T$ . In the remainder of the proof, it is possible that we could have  $x.\text{freq} = a.\text{freq}$  or  $y.\text{freq} = b.\text{freq}$ . However, if we had  $x.\text{freq} = b.\text{freq}$ , then we would also have  $a.\text{freq} = b.\text{freq} = x.\text{freq} = y.\text{freq}$ . As Fig 1 shows, we exchange the positions in  $T$  of  $a$  and  $x$  to produce a tree  $T'$ , and then we exchange

the positions in  $T'$  of  $b$  and  $y$  to produce a tree  $T''$ .



$$B(T) - B(T')$$

$$= \sum_{c \in L} c \cdot \text{freq} \cdot d_T(c) - \sum_{c \in L} c \cdot \text{freq} \cdot d_{T'}(c)$$

$$= x \cdot \text{freq} \cdot d_T(x) + a \cdot \text{freq} \cdot d_T(a) - x \cdot \text{freq} \cdot d_{T'}(x) - a \cdot \text{freq} \cdot d_{T'}(a)$$

$$= x \cdot \text{freq} \cdot d_T(x) + a \cdot \text{freq} \cdot d_T(a) - x \cdot \text{freq} \cdot d_T(a) - a \cdot \text{freq} \cdot d_T(x)$$

$$= (a \cdot \text{freq} - x \cdot \text{freq}) (d_T(a) - d_T(x))$$

$$\geq 0$$

Because both  $a \cdot \text{freq} - x \cdot \text{freq}$  and  $d_T(a) - d_T(x)$  are nonnegative. More specifically,  $a \cdot \text{freq} - x \cdot \text{freq}$  is nonnegative because  $x$  is a minimum-frequency leaf, and  $d_T(a) - d_T(x)$



is nonnegative because  $a$  is a leaf of maximum depth in  $T$ . Similarly, exchanging  $y$  and  $b$  does not increase the cost, and so  $B(T) - B(T'')$  is nonnegative. Therefore,  $B(T'') \leq B(T)$ , and since  $T$  is optimal, we have  $B(T) \leq B(T'')$ , which implies  $B(T'') = B(T)$ . Thus  $T''$  is an optimal tree in which  $x$  and  $y$  appear as sibling leaves of maximum depth, from which the lemma follows.

We first show how to express the cost  $B(T)$  of tree  $T$  in terms of the cost  $B(T')$  of tree  $T'$ , by considering the component costs in equation. For each character  $c \in C - \{x, y\}$ , we have that  $d_T(c) = d_{T'}(c)$ , and hence  $c \cdot \text{freq} \cdot d_T(c) = c \cdot \text{freq} \cdot d_{T'}(c)$ . Since  $d_T(x) = d_{T'}(x) = d_{T'}(z) + 1$ , we have that

$$x \cdot \text{freq} \cdot d_T(x) + y \cdot \text{freq} \cdot d_T(y) = (x \cdot \text{freq} + y \cdot \text{freq}) (d_{T'}(z) + 1)$$

$$= z \cdot \text{freq} \cdot d_{T'}(z) + (x \cdot \text{freq} + y \cdot \text{freq})$$

from which we conclude that

$$B(T) = B(T') + x \cdot \text{freq} + y \cdot \text{freq}$$

or, equivalently

$$B(T') = B(T) - x \cdot \text{freq} - y \cdot \text{freq}$$

we now prove the lemma by contradiction.  
 suppose that  $T$  does not represent an optimal  
 prefix code for  $C$ . Without loss of the  
 generality,  $T''$  has  $x$  and  $y$  as siblings. let  
 $T'''$  be the tree  $T''$  with common parent  
 of  $x$  and  $y$  replaced by a leaf  $z$  with  
 frequency  $z \cdot \text{freq} = x \cdot \text{freq} + y \cdot \text{freq}$ . Then

$$\begin{aligned} B(T''') &= B(T'') - x \cdot \text{freq} - y \cdot \text{freq} \\ &< B(T) - x \cdot \text{freq} - y \cdot \text{freq} \\ &= B(T') \end{aligned}$$

yielding a contradiction to the assumption  
 $T'$  represents an optimal code for  $C'$ . Thus  
 $T$  must represent an optimal code for  
 alphabet  $C$ .

#### 4. 0/1 knapsack problem

Algorithm:

// Input

// values (stored in array  $v$ )

// weights (stored in array  $w$ )

// Number of distinct items ( $n$ )

// knapsack capacity ( $w$ )

for  $j$  from 0 to  $w$  do:

$m[0, j] := 0$

for  $i$  from 1 to  $n$  do:

for  $j$  from 0 to  $w$  do:

if  $w[i] > j$  then:

$m[i, j] := m[i-1, j]$

else:

$m[i, j] := \max(m[i-1, j],$

$m[i-1, j - w[i]] + v[i])$

④ The running time =  $O(1)O(nw)$   
 $= O(nw)$

6. The max reliability for constructing system with stage  $1 \dots j$  with budget  $b$ . Redundancy( $b, j$ )

for  $j=1$  to  $n$

$Pro(0, j) = 0$

for  $b=1$  to  $B$

$Pro(b, 1) = r_0 - 1$

for  $b=1$  to  $B$

for  $j=1$  to  $n$

for  $k=1$  to  $b/c - j$

$Pro(b, j) = \max \{ Pro(b-c, j-1) * (1 - (1-r)^k) \}$

$Red[B][j] = k$

print ( $Red[ ][ ]$ )