

P1

$f_7 < f_1 < f_4 < f_{16} < f_{13} < f_2 < f_8 < f_5$

P2

ii) Algorithm

P = ii) Algorithm

① Sort the weight from small to large, the sorted list is  $S_i, i \in n$

② Let Height = 0,  $T = \{\}$ , ~~Sum-W~~ Sum-W = 0

for  $k = 0 \rightarrow n-1$

if ~~Sum-W~~  $\leq S_k$

$T.add(S_k)$ ;  $Sum-W = Sum-W + S_k \cdot weight$ ;

Height = Height + 1;

end if

end for

return Height

Time complexity:  $O(n \log n) + O(n) = O(n \log n)$

(5) Proof,

Assume greedy is not optimal, and let's ~~G<sub>0</sub> G<sub>1</sub> ... G<sub>R</sub>~~ be the stones selected by Greedy algorithm

$OPT_0, OPT_1, \dots, OPT_m$  denotes the stones selected by optimal solution

Suppose that  $OPT_0 = G_0, \dots, OPT_j = G_j$ ,  $\xrightarrow{j}$  the largest value of  $g_j$

Then we have  $OPT_{j+1} > G_{j+1}$ , However, since <sup>both</sup>  $OPT_{j+1} \& G_{j+1}$  can hold the [on j] ~~st~~ previous stones, and  $G_{j+1}$  ~~is~~ is lighter than  $OPT_{j+1}$  which is better for latter layers, therefore, we can replace  $OPT_{j+1}$  by  $G_{j+1}$  which also maintains the optimal  $\therefore OPT_{j+1} = G_{j+1}$ , contradiction  
 $\therefore$  greedy is optimal

### P3 (ii) Algorithm

Sort the timetable from small start time to large, as set (A)

then;  $M = \infty$ ,  $L = \{\emptyset\}$ ,  $\text{check\_p} = \{\emptyset\}$

for  $i=0:n-1$

if  $A_i.\text{start} \leq M$

$L.\text{add}(A_i);$

$M = \min(A_i.\text{finish}, M);$

else

$\text{clear } L, i;$

$\text{check\_p}.\text{add}(\{M\});$

$M = \infty;$

endif

endfor

return  $\text{check\_p}$

(2) Proof. Assume Greedy is not the optimal solution, and let  $G_0 \dots G_k$  be the check point selected by our algorithm,  $\text{OPT}_0 \dots \text{OPT}_m$  be the check point selected by ~~optimal~~ optimal solution, then assume  $G_0 = \text{OPT}_0 \dots G_j = \text{OPT}_j$  for the largest possible value of  $j$ .

Then ~~Assume that~~  $r^{\text{th}} \sim n^{\text{th}}$  people need to be checked after  $G_0 - G_j$  check points

and the  $r^{\text{th}} \sim (r+1)^{\text{th}}$  people has overlaped timetable.

Then, for the Greedy algorithm, after  $G_{j+1}$  check point, we need to do Greedy  $((r+1)^{\text{th}} \sim n^{\text{th}})$

for ~~optimal~~ optimal solution, after  $\text{OPT}_{j+1}$ , it will be optimal  $((r+1-x)^{\text{th}} \sim n^{\text{th}})$

which means the the optimal solution leaves some people which can be checked

$\Leftrightarrow$  we can replace  $\text{OPT}$  by Greedy, which will not affect the latter solution

i.  $\text{OPT}_{j+1} = G_{j+1}$ , contradiction

P4. Input: ( Array, Card )

Note: Judge( ) means use the equivalence test

abandon means we don't consider those cards in the following test of this algorithm

Algorithm:

F ( Array Card[], Int n )

C<sub>1</sub> = card[ 0:n  $\frac{n}{2}$  ],

C<sub>2</sub> = card[  $\frac{n}{2}$ :n ].

If n == 2

k = Judge( C<sub>1</sub>, C<sub>2</sub> ) # judge whether these two cards are same or not. If same, return 2

If k == 2, mark C<sub>1</sub>, C<sub>2</sub> are the same, return 2

else abandon C<sub>1</sub>, C<sub>2</sub>, return 0

endif

k<sub>1</sub> = F [ C<sub>1</sub>,  $\frac{n}{2}$  ].

k<sub>2</sub> = F [ C<sub>2</sub>,  $\frac{n}{2}$  ].

If k<sub>1</sub> == k<sub>2</sub>: k = Judge( C<sub>1</sub>, C<sub>2</sub> ) # only test once, because C<sub>1</sub>, C<sub>2</sub> are the same, respectively.  
If k == 2, make C<sub>1</sub>, C<sub>2</sub> the same, return k+k<sub>1</sub>; else abandon C<sub>1</sub>, C<sub>2</sub>, return 0, endif

endif

If k<sub>1</sub> or k<sub>2</sub> == 0, abandon C<sub>1</sub>, C<sub>2</sub>, return 0

else, k = Judge( C<sub>1</sub>, C<sub>2</sub> )

If k == 2, mark C<sub>1</sub>, C<sub>2</sub> the same, return k+k<sub>1</sub>

else abandon the smaller one, return max( k<sub>1</sub>, k<sub>2</sub> )  
( C<sub>1</sub>, C<sub>2</sub> )

endif

endif

end

This algorithm above can check which card has the largest number then we can use this card to do another (m) test to check whether there exist  $\frac{n}{2}+1$  this card. (Compare with all other cards)  
The ~~Judge~~ invocations are  $O(\log n \cdot (\frac{1}{2}n + 1) \div 2) + O(m)$   
 $= O(n \log n)$  for the worst case.

P5.

- a) Line 6<sup>n|b</sup> means that it divides A into B and C, where B stores the  $a_i$  which has 1 in  $k^{\text{th}}$  bit, C stores the  $a_i$  which has 0 in  $k^{\text{th}}$  bit

Line 12<sup>n|b</sup> is the base case, means that if the  $i^{\text{th}}$  bit of A are all 0 or all 1, then we ~~can~~ always can find a number  $x_i$  makes  $x_i \oplus A_{i^{\text{th}} \text{bit}} = 0$   
 $\therefore$  If  $B \& C$  is ~~not~~ empty we return 0  
else ~~return~~ return 1, obviously

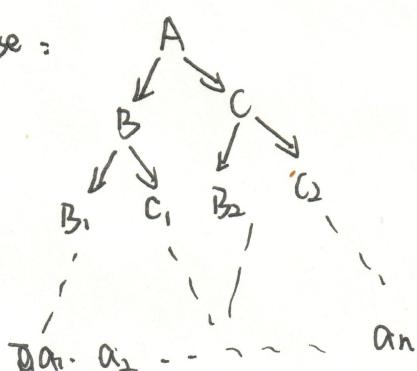
Lines 27 are the recursion step, which means that:

If B and C are not empty, then the <sup>maximum</sup> XOR result must be 1 at  $k^{\text{th}}$  bit, this is why Line 26 add +.

If B or C is empty, which means that there must exist  $x_k$  makes  $x_k \oplus A_{k^{\text{th}} \text{bit}} = 0$  therefore we only need to consider the latter bits  
this is why we do recursion in Line 18, 21, and only compute  $\min\{f(A \setminus B), f(A \setminus C)\}$  in Line 25

- b) The time complexity is ~~O(K:N)~~  $O(K:N)$

because:



In each layer, since  $A \geq B + C$   
therefore every layer has  $n$  components  
in all  $(a_1, a_2, \dots, a_n)$

And in each ~~function~~ function, we have  
only one ~~loop~~ loop, therefore in each  
layer, we will access all the  $a_i$   
 $\Leftrightarrow O(n)$

and the depth of this tree is  $K$   
~~because~~ obviously (from  $K$  to 1)

$$\therefore O(n) \cdot O(K) = O(Kn)$$

### P6. Algorithm.

$\text{OPT}(A, B, C, i, j, k)$  #  $i \geq j = k \geq 0$  at beginning

If  $A_i \neq C_k$  and  $B_j \neq C_k$

endif return  $R=0$

If  $A_i = C_k$  and  $B_j \neq C_k$ :

$R = \text{OPT}(A, B, C, i+1, j, k+1)$

endif

If  $A_i \neq C_k$  and  $B_j = C_k$ :

$R = \text{OPT}(A, B, C, i, j+1, k+1)$

endif

If  $A_i = C_k$  and  $B_j = C_k$ :

$R = \max \{ \text{OPT}(A, B, C, i+1, j, k+1), \text{OPT}(A, B, C, i, j+1, k+1) \}$

endif

return  $R$

end

Let  $K = \text{OPT}(A, B, C, 0, 0, 0)$

then if  $K = m n$

return true

else

return false.

endif

Note, since the ~~the~~ range of  $i, j, k$  should be  ~~$[0, m-1], [0, n-1]$~~   $[0, m-1] \times [0, n-1]$  and  $[0, m+n-1]$  respectively

so the algorithm ~~will~~ <sup>will</sup> error due to the range,

Since this error is not vital to the algorithm, I do not fix them

in the algorithm, but needs to add some limitation ~~when I~~ \* to realize this algorithm  
for the simplicity

P7. Algorithm. This problem is a RMQ problem, we use ST table to solve it.

① pre-processing

the range,  
limit the input  $(l, r)$  into  $[1, n]$ ; (constant time)

② build ST-table.

define:  $f[i][j]$  denotes the max value start from  $i^{\text{th}}$ , end with  $(i + 2^{j-1})^{\text{th}}$

then every  $f[i][j]$  can be compute by,

$$f[i][j] = \max(f[i][j-1], f[i+2^{j-1}][j])$$

use this function to build a table

which stores the value of  $f[0][1], f[0][2], f[0][3], \dots$

$f[0][2], f[0][3], \dots$

⋮

$f[0][\log n], f[1][\log n], \dots$

the time complexity is  $O(\log n) + O(\log(n)) = O(\log^2 n)$

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

③ search:

$L = r - l + 1$  is the length of this interval. Let  $k = \log_{\text{down rounding}} L$

divide  $L$  into two parts:

$$\text{answer} = f[r-2^k][k] = \max\{f[l][k], f[r-2^k+1][k]\}$$

is the ~~right~~ answer, time complexity is  $O(1)$  from all questions

in total complexity  $O(n \log n + k)$