

## Tom's Kitchen

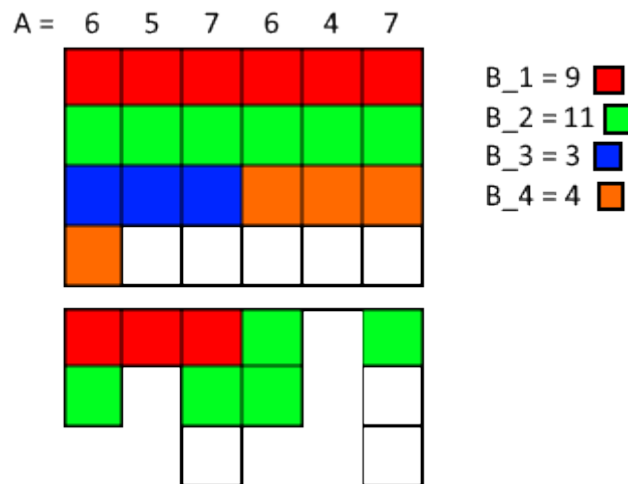
Spoiler

Subtask 1 was intended to permit simple case-analysis based solutions.

Subtask 2 was intended to permit brute force search solutions.

Subtask 3 was a reduction to a standard Dynamic Programming problem (commonly stated as: you have certain coins, find if you can pay exactly  $X$  amount of money with them).

Subtask 4 was intended to permit task-specific but suboptimal Dynamic Programming solutions.



The full solution uses Dynamic Programming. Let us mentally reorder the hours spent on each meal such that for the first  $K$  hours, all chefs are different. This way we can visualize these  $K$  hours forming an  $N \times K$  "diversity box", with all "non-diverse hours" coming afterwards (as shown in the figure above). Now let's make the following observations:

1. Each chef can add at most 1 hour to any column of the "diversity box".
2. A chef  $j$  can fill the "diversity box" by at most  $\min(B_j, N)$ .
3. In a correct solution "diversity box" must be filled by an amount at least  $N \cdot K$ .
4. Suppose you have decided to hire chefs for a total of  $H$  hours. Then it's optimal to hire such set of chefs that the "diversity box" is filled as much as possible (perhaps even overfilled).

Now let  $D[c][h]$  be the maximum amount we can fill the "diversity box" by picking a subset of chefs  $1, \dots, c$  such that they are hired for a total of  $h$  hours. Now let us notice that for the value  $D[c][h]$  there are two possibilities:

1. The maximal subset contains chef  $c$ . Thus the other chefs in this subset form a maximal solution for  $D[c-1][h - B_c]$  (otherwise we could pick a better subset). Thus  $D[c][h] = D[c-1][h - B_c] + \min(B_c, N)$ .
2. The maximal subset doesn't contain chef  $c$ . Thus this subset also forms a maximal solution for  $D[c-1][h]$  (a better solution to  $D[c-1][h]$  would contradict the maximality of this subset). Thus  $D[c][h] = D[c-1][h]$ .

Now we simply need to consider the two cases and see which gives us a better solution. Thus  $D[c][h] = \max(D[c-1][h - B_c] + \min(B_c, N), D[c-1][h])$ . For performing the dynamic programming computation, we can initialize  $D[0][0]$  to 0 and every other  $D[i][j]$  to  $\infty$ . Note that once we have computed  $D[c][*]$  we don't care about  $D[c-1][*]$  anymore, so we can optimize memory consumption. The answer will be minimum non-negative  $h - \sum_i A_i$  such that  $D[N][h] \geq N \cdot K$ .

```
import array
N = 300
n,m,k = [int(x) for x in input().split()]
a = [int(x) for x in input().split()]
b = [int(x) for x in input().split()]
supply = array.array('i', [-N*N for i in range(N*N+1)])
supply[0] = 0;
def solve():
    if(min(a) < k):
        return 'Impossible'
    bsum = 0
    for x in b:
        bsum += x
        for i in range(bsum,-1,-1):
            supply[i+x] = max(supply[i+x],supply[i]+min(x,n))
    for i in range(sum(a),N*N+1):
        if(supply[i] >= n*k):
            return i-sum(a)
    return 'Impossible'
print(solve())
```

## Credits

- Task: Bernhard Linn Hilmarsson (Iceland)
- Solutions and tests: Oliver-Matis Lill, Andres Unt (Estonia)