

Problem 1

Pseudocode:

KindOfIncreasingSubseq(input[]):

- a. Initialize array OPT[length of input] [length of input] to 0
- b. maxlen=0
- c. for i to length of input:
 - i. for j=i+1 to length of input:
 1. OPT[i] [j] = 2
 2. maxInside=-1
 3. for k=0 to i:
 - a. if (input[k]+input[i]/2 < input[j]):
 - i. maxInside= Max(OPT[k][i] and maxInside)
 - ii. OPT[i][j]=1+maxInside
 - iii. maxlen= Max(maxLen, 1+maxInside)
- d. return maxlen;

What OPT[i][j]: length of the largest subsequence that can be formed using the first j elements, including j such that the average of each pair of elements before that is less than element at j.

Relation to smaller problems: $\text{OPT}[i][j] = \max(\{1 + \max(\text{all } \text{OPT}[k][i])\})$

Here, $j > i$ and $i > k$, and the condition is that the average of kth element and the element at i is less than the value of element at j, and if there is no such element the the OPT[i][j] is simply equal to 2.

Location of the solution to the original problem: Maximum value of all the OPT 2d array is the maximum length of the “kind-of-increasing” subsequence.

Running Time estimate:

- Line **a** is initializing the 2d array of size of the length of the input which is $O(n^2)$
- Line **b** is constant time, initializing a variable
- Line **c** contains a for loop from 0 to length, which will be $O(n)$.
 - Inside this we have another for loop at line **(i)** which goes from $i+1$ to length, which is again $O(n)$
 - Inside this we have another for loop which goes from 0 to i , which is again $O(n)$
 - Inside this we have constant time tasks and assigning values to variables.

The complexity for line **c** after including 3 nested loops will be $O(n^3)$

So, the total running time for the problem will be given by:

$O(n^2) + O(1) + O(n^3)$ which is just $O(n^3)$

Problem 2

Pseudocode:

1. **getMaxCost(w1,w2,weights[], costs[], size):**
 - a. $OPT[size][w1][w2]$ initialize it to 0
 - b. for s1 from 0 to w1:
 - i. for s2 from 0 to w2:
 1. for i from 1 to size:
 - a. if $s1 \geq weights[i]$ and $s2 \geq weights[i]$:
 - i. $OPT[i][s2][s2] = \text{Max}(OPT[i-1][s1][s2],$ - ii. $OPT[i-1][s1-weights[i]][s2] + costs[i]$,
 - iii. $OPT[i-1][s1][s2-weights[i]] + costs[i]$)
 - b. else if $s1 \geq weights[i]$:
 - i. $OPT[i][s2][s2] = \text{Max}(OPT[i-1][s1][s2],$ - ii. $OPT[i-1][s1-weights[i]][s2] + costs[i]$,
 - c. else if $s2 \geq weights[i]$:
 - i. $OPT[i][s2][s2] = \text{Max}(OPT[i-1][s1][s2],$ - ii. $OPT[i-1][s1][s2-weights[i]] + costs[i]$)
 - d. else
 - i. $OPT[i][s2][s2] = OPT[i-1][s1][s2]$
 - c. $result = OPT[size][w1][w2]$
 - d. $i = size, j = w1, k = w2$
 - e. $bag1[size]$
 - f. $bag2[size]$
 - g. $m = 0, q = 0$
 - h. for i from size to 0, conditions: $result > 0$:
 - i. if $result == OPT[i-1][j][k]$:
 1. continue
 - ii. else
 1. $result = result - costs[i]$
 2. if $j - weights[i] \geq 0$:
 - a. $bag1[q] = i$
 - b. $q++$ // index for keeping track of items in bag1
 - c. $j = j - weights[i]$
 3. else if $k - weights[i] \geq 0$:
 - a. $bag2[m] = i$
 - b. $m++$ // index for keeping track of items in bag1
 - c. $k = k - weights[i]$
 - i. print bag1
 - j. print bag2
 - k. print $OPT[size][w1][w2]$

What $OPT[i][w1][w2]$: The maximum cost of a subset of first i items where the weight is w1 and w2 and items inserted in either w1 or w2 are indivisible.

Relation to smaller problems: There will be 4 conditions for the given problem:

- The item can be included in w_1
 - Given that the item can fit in w_1 , then :
 - $OPT[i][w_1][w_2] = \max(OPT[i-1][w_1][w_2], OPT[i-1][w_1-w_i][w_2] + \text{costs at } i)$
- The item can be included in w_2
 - Given that the item can fit in w_2 , then:
 - $OPT[i][w_1][w_2] = \max(OPT[i-1][w_1][w_2], OPT[i-1][w_1][w_2-w_i] + \text{costs at } i)$
- The item can be included in either of the 2 bags, i.e. w_1 and w_2
 - Given that the item can go in either bag, we need to check for the maximum profit/cost:
 - $OPT[i][w_1][w_2] = \max(OPT[i-1][w_1][w_2], OPT[i-1][w_1-w_i][w_2] + \text{costs at } i, OPT[i-1][w_1][w_2-w_i] + \text{costs at } i)$
- The item is not included in either of them
 - Given that item does not fit any of the bags:
 - $OPT[i][w_1][w_2] = OPT[i-1][w_1][w_2]$

Location of the solution to the original problem: $OPT[\text{size}][\text{weight } 1][\text{weight } 2]$

The items in the bag can be traced back using:

$OPT[i-1][W_1-w_i][W_2] + C_i \geq OPT[i-1][W_1][W_2]$ (if in bag 1) or

$OPT[i-1][W_1][W_2-w_i] + C_i \geq OPT[i-1][W_1][W_2]$ (if in bag 2)

Running Time estimate:

- Initializing the 3d $OPT[\text{size}][w_1][w_2]$ which will be $O(\text{size} * w_1 * w_2)$
- For loop 0 to w_1 which will be $O(w_1)$
 - For loop from 0 to w_2 which will be $O(w_2)$
 - For loop from 0 to size which will be $O(\text{size})$
 - Comparison operations, updating values which is constant time of $O(1)$
- Constant time operations from line c to h
- For loop from size to 0 which is $O(\text{size})$
 - Constant time operations inside that

Total runtime for the program will be: $O(\text{size} * w_1 * w_2) + O(\text{size} * w_1 * w_2) + O(\text{size})$

which is $O(\text{size} * w_1 * w_2)$.

Problem 3

Pseudocode:

1. canBeBuilt(c1, c2, c3, n, input):

- a. totalLengthToBeBuilt= sum of all the integers in input
- b. cracks[n-1] list of integer, that will store the crack points
- c. q=0
- d. prev=0
- e. for i from n-1 to 0:
 - i. cracks[q]+=input[i]+prev+1
 - ii. prev+=input[i]
 - iii. q++
- f. q=0
- g. OPT[totalLengthToBeBuilt + 1][c1+1][c2+1][c3+1] of type Boolean
- h. for l from 0 to totalLengthToBeBuilt:
 - i. for i from c1 to 0:
 1. for j from c2 to 0:
 - a. for k from c3 to 0:
 - i. if l==0
 1. OPT[l][i][j][k]= true
 2. Continue
 - ii. if (l>=1 and i>=1 and OPT[l-1][i-1][j][k]) or l>=2 and j>=1 and OPT[l-2][i][j-1][k]) or l>=3 and k>=1 and OPT[l-3][i][j][k-1]):
 1. OPT[l][i][j][k]=true
 - iii. If q< cracks.length and cracks[q]=l:
 1. OPT[l][i][j][k]=false
 - ii. If q<cracks.length and l>cracks[q]:
 1. q++
- i. return OPT[totalLengthToBeBuilt][c1][c2][c3]

What $\text{OPT}[l][i][j][k]$: Whether the bricks of length 1,2,3 of quantity can i,j,k respectively can be placed on top of the given bricks of total length l input of one row.

Relation to smaller problems: Following is the relation:

- if $l=0$, the given $\text{OPT}[l][i][j][k]=\text{true}$ for all the values
- if any of the following condition are true, the $\text{OPT}[l][i][j][k]$ should be:
 - $\text{OPT}[l-1][i-1][j][k]$ // when $l \geq 1$ and $i \geq 1$
 - $\text{OPT}[l-2][i][j-1][k]$ // when $l \geq 2$ and $j \geq 1$
 - $\text{OPT}[l-3][i][j][k-1]$ // when $l \geq 3$ and $k \geq 1$
- If l is any of the cracks in the input array, then the $\text{OPT}[l][i][j][k]=\text{false}$ for all the cases.

Location of the solution to the original problem: The value of the $\text{OPT}[\text{totalLengthOfAllBricksInRow}][c1][c2][c3]$ will decide whether or not the wall can be built on top a row of bricks.

Running Time estimate:

- Getting the total length of the row, is $O(n)$
- Filling in the cracks array is $O(n)$
- Initializing the $\text{OPT}[\text{totalLength}+1][c1+1][c2+1][c3+1]$ will be $O(\text{totalLength} * c1 * c2 * c3)$
- For loop from 0 to totalLength which is $O(\text{totalLength})$
 - For loop from $c1$ to 0 which is $O(c1)$
 - For loop from $c2$ to 0 which is $O(c2)$
 - For loop from $c3$ to 0 which is $O(c3)$
 - Constant time operations which is $O(1)$
 - Updating the value of q which is again $O(1)$

Total runtime of the program will be given by $O(n) + O(\text{totalLength} * c1 * c2 * c3) + O(\text{totalLength} * c1 * c2 * c3)$ which is same as :

$O(\text{totalLength} * c1 * c2 * c3)$, According to the given problem statement, the maximum length of 1 number will be 100, so the worst case, total length would be $100 * n$, n is the number of bricks

So the total runtime estimate will be: $O(n * c1 * c2 * c3)$

Problem 4

Pseudocode:

1. getMinimumCost(input[]):

- a. $n = \text{length of input}$
- b. initialize a arrays $m[][]$ and $s[][]$ of size n
- c. $j=0, q=0$
- d. for $d=1$ to $n-1$:
 - i. for $i=1$ to $n-d$
 1. $j=i+d$
 2. $\text{min} = -\text{infinity}$
 3. for $k=i$ to $j-1$:
 - a. $q = m[i][k] + m[k+1][j] + \text{input}[i-1] * \text{input}[k] * \text{input}[j]$
 - b. if $q < \text{min}$:
 - c. $\text{min} = q$
 - d. $s[i][j] = k$
 4. $m[i][j] = \text{min}$
- e. $\text{cost} = m[1][n-1]$
- f. $\text{printOptimal}(s, 1, \text{len of } s - 1)$

2. printOptimal(s[][], i, j):

- a. if $i=j$:
 - i. $\text{print}("A" + i)$
- b. else:
 - i. $\text{print}("(" + "$
 - ii. $\text{printOptimal}(s, i, s[i][j])$
 - iii. $\text{print}("x" + "$
 - iv. $\text{printOptimal}(s, s[i][j]+1, j)$
 - v. $\text{print}(")")$

What $\text{OPT}[i][j]$: minimum numbers of multiplications needed to multiply matrix A_i through A_j , where A is the input array of the size of the matrices.

Relation to smaller problems:

$\text{OPT}[i][j] = \{0 \text{ when } i=j,$

Minimum of $(\text{OPT}[i][k] + \text{OPT}[k+1][j] + A_i * A_k * A_j)$ when $i < j \}$

Location of the solution to the original problem: $\text{OPT}[1, n]$ n represents the length of the input.

Running Time estimate:

- For getMinimumCost()
 - Initializing the 2d array $m[][]$ and $s[][]$ is $2*O(n^2)$
 - For loop from 1 to length which is $O(n)$
 - Another for loop from 1 to $len - d$ which is again $O(n)$
 - Another loop from i to j which is again $O(n)$
 - Constant time operations for updating the variables
- For print optimal()
 - We have recursive calls to the function to print the paranthesis and format the solution, and depth of the tree is same as n , and the complexity should be $O(n)$.

Total running time will be given by: $2*O(n^2) + O(n^3) + O(n)$ which is : $O(n^3)$