

BTH004 - Laboratory assignment 1— Multiple knapsacks problem and solutions

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Part I

1 Greedy algorithm

1.1 mainly description for the algorithm

allocate knapsack for every items in the order of values per weight unit.

1.2 implementation step

step 1: Create a new list calculate all items' unit value(value per weight unit)in the items list and record its id, and then sort it in the order of unit value.

step 2: For every item in the sorted list, allocate knapsack for the item in the order. if there are enough room in this knapsack , then include it in the current knapsack. If all knapsacks don't have enough room for the item, then it will not be included in all knapsacks.

step 3: All items have been allocated knapsack once, algorithm them end.

1.3 two fit method are alternative: first fit and best fit

first fit: Once the current knapsack has enough room, then include in the knapsack immediately.

best fit: Search all knapsacks, and look for the best knapsack(which left capacity most closed to the item weight)

1.4 descriptions for some criteria

termination criteria: All items have been allocated knapsacks once, then end the algorithm.

1.5 pseudo code

```
1 def greedyForMultipleKnapsack_firstFit(knapsacks, items):
2
3     valuesPerUnit = []
4     leftCapacity = knapsacks.copy()
5
6     for i in range(len(items)):
7         valuesPerUnit.append([items[i][0]/items[i][1], i])
8
9     #sort the array by the order of value per unit
10    totalValue = 0
11    for i in range(len(items)):
12        for j in range(len(knapsacks)):
13            #if find the knapsack with enough space:
14                #include the item in this knapsack, add the value to
                the total value and reduce the knapsack space.
15    return items, totalValue
16
17 def greedyForMultipleKnapsack_bestFit(knapsacks, items):
18     valuesPerUnit = []
19     leftCapacity = knapsacks.copy()
20     for i in range(len(items)):
21         valuesPerUnit.append([items[i][0]/items[i][1], i])
22     #sort the array by the order of value per unit
23
24     totalValue = 0
25     for i in range(len(items)):
26         knapIndex = -1
27         for j in range(len(knapsacks)):
28             #find the knapsack who have closest space for current
            item,
29             if knapIndex != -1:
30                 totalValue += items[valuesPerUnit[i][1]][0]
31                 leftCapacity[knapIndex] -= items[valuesPerUnit[i]
32 ] [1] [1]
33                 items[valuesPerUnit[i][1]][2] = knapIndex
34     return items, totalValue
```

1.6 code & comment

```
1 def greedyForMultipleKnapsack_firstFit(knapsacks, items):
2 # input:
3 # @para knapsacks -> [1,2] represent 2 knapsacks with (weight)
    capacity 1,2
4 # @para items -> [[1,2,-1],[3,4,0]] represent 2 items
5 # items[0] means this item worth 1 unit, weigh 2 but not include in
    any knapsack,
6 # while item[1] means this item worth 3 unit, weigh 4 unit and is
    included in knapsacks[0]
```

```

7 print("\033[1;35mgreedy algorithm for multiple knapsacks ---
first fit start->")
8 print('input:')
9 print('knapsacks: \n{}'.format(knapsacks))
10 print('items: \n{}\n'.format(items))
11
12 valuesPerUnit = []
13 #used to record the value per weight unit for item and item
index.
14 leftCapacity = knapsacks.copy()
15 #we will modify the list so copy a new knapsacks to avoid to
modify the original
16 for i in range(len(items)):
17     #fill the lists mentioned above
18     valuesPerUnit.append([items[i][0]/items[i][1],i])
19 print('and the value per weight unit with the relevant item id
are: \n{}'.format(valuesPerUnit))
20
21 valuesPerUnit.sort(reverse=True)
22 #descending sort by the value per unit
23 print('after sort, it becomes: \n{}'.format(valuesPerUnit))
24 print("\nnow we start allocate knapsack for each items in the
order of the value per unit ->\033[0m")
25 totalValue = 0
26 # record the total values of these items have included
27 for i in range(len(items)):
28     #allocate knapsack for the items in the order of values per
unit
29     isInclude = False
30     print("\033[1;32m-----\n\033[0m")
31     print("\033[1;32mnow allocate knapsack for items[{}]\n
\033[0m".format(valuesPerUnit[i][1]))
32     print("\033[1;36mknapsacks' condition:{}\n\033[0m".format(
leftCapacity))
33     for j in range(len(knapsacks)):
34         #search rest space in knapsack for the item
35         print("for knapsacks[{}], it still has {} weight left
".format(j,leftCapacity[j]))
36         print("and items[{}] weight {} ".format(valuesPerUnit[
i][1],items[valuesPerUnit[i][1]][1]))
37         input()
38         if items[valuesPerUnit[i][1]][1] <= leftCapacity[j]:
39             # if have enough space for the current item
40             print("\033[1;34mit has enough space for the item,
so we include it in this knapsack\n\033[0m")
41             totalValue += items[valuesPerUnit[i][1]][0]
42             # add the item value to the total
43             leftCapacity[j] -= items[valuesPerUnit[i][1]][1]
44             # it take up a part of space of the current
knapsack,so we decrease the room
45             items[valuesPerUnit[i][1]][2] = j
46             # record its position
47             isInclude = True
48             break
49             # current item have found its place,next item
50         else:
51             print("\033[1;33mknapsacks[{}] not has enough space

```

```

    for this item\n\033[0m".format(j))
52     if not isInclude:
53         print("\033[1;31mall knapsacks have no space for this
item!\n\033[0m")
54     print('\033[1;35mall items have been allocated!')
55     print("\033[1;35m\n*****")
56     print("<-algorithm end")
57
58     print('{} {} \033[0m'.format(items,totalValue))
59     return items,totalValue
60
61 def greedyForMultipleKnapsack_bestFit(knapsacks,items):
62     print("\033[1;35mgreedy algorithm for multiple knapsacks ---
best fit start->")
63     print('input:')
64     print('knapsacks: \n{}'.format(knapsacks))
65     print('items: \n{}\n'.format(items))
66
67     valuesPerUnit = []
68     #used to record the value per weight unit for item and item
index
69     leftCapacity = knapsacks.copy()
70     #we will modify the list so copy a new knapsacks to avoid to
modify the original
71     for i in range(len(items)):
72         #fill the lists mentioned above
73         valuesPerUnit.append([items[i][0]/items[i][1],i])
74     print('and the value per weight unit with the relevant item id
are: \n{}'.format(valuesPerUnit))
75
76     valuesPerUnit.sort(reverse=True)
77     #descending sort by the value per unit
78     print('after sort, it becomes: \n{}'.format(valuesPerUnit))
79     print("\nnow we start allocate knapsack for each items in the
order of the value per unit ->\033[0m")
80
81     totalValue = 0
82     # record the total values of these items have included
83     for i in range(len(items)):
84         #allocate knapsack for the items in the order of values per
unit
85         difference = max(leftCapacity)
86         # used to record the best fit diifference, initalize with
the largest left rooms in these knapsacks
87         knapIndex = -1
88         # used to record the best fit diifference
89         print("\033[1;32m-----\n\033[0m")
90         print("\033[1;32mnow allocate knapsack for items[{}]\n
\033[0m".format(valuesPerUnit[i][1]))
91         print("\033[1;36mknapsacks' condition:{}\n\033[0m".format(
leftCapacity))
92         for j in range(len(knapsacks)):
93             #search rest space in knapsack for the item
94             print("for knapsacks[{}], it still has {} weight left
".format(j,leftCapacity[j]))
95             print("and items[{}] weight {} ".format(valuesPerUnit[
i][1],items[valuesPerUnit[i][1]][1]))

```

```

96         input()
97         newDifference = leftCapacity[j] - items[valuesPerUnit[i]
][1]][1]
98         if newDifference >= 0:
99             print("\033[1;36mknapsacks[{}] it has enough space
for this item\n\033[0m".format(j))
100             if newDifference < difference:
101                 #if found a new better fit in knapsacks, choose
it as the best fit
102                 print("\033[1;36mand it's more appropriate for
this item\n\033[0m")
103                 knapIndex = j
104                 difference = newDifference
105             else:
106                 print("\033[1;33mbut it's not better than
current one\n\033[0m")
107             else:
108                 print("\033[1;33mknapsacks[{}] not has enough space
for this item\n\033[0m".format(j))
109
110         if knapIndex != -1:
111             print("\033[1;34mso we include it in knapsacks[{}](
weigh {} capacity {})\n\033[0m".format(knapIndex, items[
valuesPerUnit[i][1]][1], knapsacks[knapIndex]))
112             totalValue += items[valuesPerUnit[i][1]][0]
113             # add the item value to the total
114             leftCapacity[knapIndex] -= items[valuesPerUnit[i]
][1]][1]
115             # it take up a part of space of the current knapsack,so
we decrease the room
116         else:
117             print("\033[1;31mall knapsacks have no space for this
item!\n\033[0m")
118             items[valuesPerUnit[i][1]][2] = knapIndex
119
120         print('\033[1;35mall items have been allocated!')
121         print("\033[1;35m\n*****")
122         print("<-algorithm end")
123
124         print('{} {} \033[0m'.format(items, totalValue))
125         return items, totalValue

```

1.7 correctness check & test

As you see in the code above, a lot of print statement are in the code section. They are used to show the process of the program.

```

greedy algorithm for multiple knapsacks --- first fit start->
input:
knapsacks:
[5, 1, 2]
items:
[[2, 3, -1], [5, 3, -1], [2, 2, -1], [8, 5, -1], [1, 1, -1]]

and the value per weight unit with the relevant item id are:
[[0.6666666666666666, 0], [1.6666666666666667, 1], [1.0, 2], [1.6, 3], [1.0, 4]]
after sort, it becomes:
[[1.6666666666666667, 1], [1.6, 3], [1.0, 4], [1.0, 2], [0.6666666666666666, 0]]

now we start allocate knapsack for each items in the order of the value per unit ->
-----

now allocate knapsack for items[1]
knapsacks' condition:[5, 1, 2]

for knapsacks[0], it still has 5 weight left
and items[1] weight 3

```

That means we can observe the process of the program, that's the most direct way of correctness check.

for current example, the input of this algorithm will be:

```

knapsacks[0] not has enough space for this item
for knapsacks[1], it still has 1 weight left
and items[0] weight 3

knapsacks[1] not has enough space for this item
for knapsacks[2], it still has 0 weight left
and items[0] weight 3

knapsacks[2] not has enough space for this item
all knapsacks have no space for this item!
all items have been allocated!

*****
<-algorithm end
[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]] 8

```

2 Neighbourhood search

2.1 mainly description for the algorithm

Choose a suitable solution as the start solution, then find its best "neighbourhood" from all neighbourhood as the solution for the next iteration. When the best neighbourhood is even not better than the current one, then the algorithm end and output the current solution.

2.2 implementation step

step 1: Choose a suitable solution as the start solution.

step 2: while there are better neighbourhood existing, then choose the neighbourhood as the solution for next iteration. Redo this step continually.

step 3: If the best neighbourhood is even not better than the current solution, the algorithm end and output the current solution.

2.3 descriptions for some criteria

neighbourhood definition: On the base of the current solution, move one item to a certain knapsack. After removing a lowest value per unit item from this knapsack(if need), if there is enough space for this new item, then it is neighbourhood of the current solution.

termination criteria: If the best neighbourhood is even not better than the current solution, the algorithm end and output the current solution.

2.4 pseudo code

```
1 def neighbourhood(knapsacks, items):
2     # search neighbourhoods of current solution
3     #the definition of neighbour is: move one item's from a
4     knapsack (or not included) to another knapsack
5     Fx = []
6     for i in range(len(items)):
7         for j in range(len(knapsacks)):
8             if items[i][2] != j:
9                 # search the neighbours but not include itself
10                # move items[Xn[i]][1] to knapsack j
11                # calculate value for current solution.
12                # if the total weight out of the capacity, then
13                give up an item with the lowest value per unit.
14                #then add this solution to the Fx
15            return Fx
16
17 def bestNeighbourhood(Fx, knapsacks, items):
18     # find the best neighbour with the largest value in all
19     neighbours
20
21 def neighbourhoodSearchForMultipleKnapsack(knapsacks, items):
22     currentBestValue = 0
23     # record the total value with current solution
24     for item in items:
25         if item[2] != -1:
26             currentBestValue += item[0]
```

```

27     bestNeighbourhoodValue = currentBestValue + 1 # record the best
        values of all neighbourhoods
28     firstR = True
29     count = 0
30     bestSolution = []
31     for item in items:
32         bestSolution.append(item[:])
33
34     while currentBestValue < bestNeighbourhoodValue:
35         currentBestValue = bestNeighbourhoodValue
36         if firstR:
37             firstR = False
38             currentBestValue -= 1
39             items = []
40             for item in bestSolution:
41                 items.append(item[:])
42             nbhs = neighbourhood(knapsacks, items)
43             bestSolution, bestNeighbourhoodValue = bestNeighbourhood(
nbhs, knapsacks, items)
44             count+=1
45
46     return items, currentBestValue

```

2.5 code & comment

```

1 def neighbourhood(knapsacks, items):
2     # search neighbourhoods of current solution
3     #the definition of neighbour is: move one item's from a
    knapsack (or not included) to another knapsack
4     Fx = []
5     for i in range(len(items)):
6         for j in range(len(knapsacks)):
7             if items[i][2] != j:
8                 # search the neighbours but not include itself
9                 temp = items[i][2]
10                items[i][2] = j
11                # move items[[Xn[i]][1]] to knapsack j
12
13                totalValue = 0
14                # calculate value for current solution.
15                for item in items:
16                    if item[2] != -1:
17                        totalValue += item[0]
18                totalWeight = 0
19                for item in items:
20                    if item[2] == j :
21                        totalWeight += item[1]
22                if knapsacks[j] >= totalWeight:
23                    Xtemp = []
24                    for item in items:
25                        Xtemp.append([item[0], item[1], item[2]])
26                    Fx.append([Xtemp, totalValue])
27                else:
28                    index = i
29                    minVal = items[i][0]
30                    for k in range(len(items)):
31                        if items[k][2] == j and items[k][0] <

```



```

minVal:
32         index = k
33         minVal = items[k][0]
34         totalValue -= items[index][0]
35         if index != i and totalWeight - items[index][1]
<= knapsacks[j]:
36             items[index][2] = -1
37             Xtemp = []
38             for item in items:
39                 Xtemp.append([item[0], item[1], item[2]])
40             Fx.append([Xtemp, totalValue])
41
42         items[i][2] = temp
43     return Fx
44
45 def bestNeighbourhood(Fx, knapsacks, items):
46     # find the best neighbour with the largest value in all
    neighbours
47     if len(Fx) == 0:
48         return [], 0
49
50     bestIndex = -1
51     maxTotalValue = 0
52     for i in range(len(Fx)):
53         if Fx[i][1] > maxTotalValue:
54             maxTotalValue = Fx[i][1]
55             bestIndex = i
56     return Fx[bestIndex][0], maxTotalValue
57
58 def neighbourhoodSearchForMultipleKnapsack(knapsacks, items):
59     print("\033[1;35mneighbourhood search algorithm for multiple
    knapsacks --- start->")
60     print('input:')
61     print('knapsacks: \n{}'.format(knapsacks))
62     print('items: \n{}\n\033[0m'.format(items))
63
64     currentBestValue = 0
65     # record the total value with current solution
66     for item in items:
67         if item[2] != -1:
68             currentBestValue += item[0]
69
70     bestNeighbourhoodValue = currentBestValue + 1 # record the best
    values of all neighbourhoods
71     firstR = True
72     print("\033[1;36mwe choose the input items as our start
    solution. it's: \n{}\033[0m".format(items))
73     input()
74     count = 0
75     bestSolution = []
76     for item in items:
77         bestSolution.append(item[:])
78
79     while currentBestValue < bestNeighbourhoodValue:
80         #termination condition: current value equals the best value
        of all of its neighbours
81         currentBestValue = bestNeighbourhoodValue

```

```

82         if firstR:
83             firstR = False
84             currentBestValue -= 1
85             items = []
86             for item in bestSolution:
87                 items.append(item[:])
88             print("\033[1;32miteration {}: ".format(count))
89             print("-----\n\033[0m")
90             print("\033[1;36mthe current solution is: \n{}".format(
bestSolution))
91             print("for this solution, the total value is: {}\n\033[0m".
format(currentBestValue))
92             nbhs = neighbourhood(knapsacks, items)
93             print("and all it's feasible {} neighbours are:".format(len
(nbhs)))
94             input()
95             for nbh in nbhs:
96                 print("solution is: \n{}".format(nbh[0]))
97                 print("value is {}\n".format(nbh[1]))
98                 bestSolution, bestNeighbourhoodValue = bestNeighbourhood(
nbhs, knapsacks, items)
99             print("\033[1;34mso we choose the neighbour \n{}\nas our
next solution, value is {}\033[0m".format(bestSolution,
bestNeighbourhoodValue))
100             count+=1
101             input()
102         else:
103             print("\033[1;33mbut we have found that even the best
neighbour is not better than the current solution, so we end
the loop\n\033[0m")
104             print("\033[1;35m\n*****\033[0m")
105             print("\033[1;35m<-algorithm end\n\033[0m")
106
107         return items, currentBestValue

```

2.6 correctness check & test

Like above saying that we can observe the process of the program. we choose the output of the greedy algorithm as the start solution.

```

neighbourhood search algorithm for multiple knapsacks --- start->
input:
knapsacks:
[5, 1, 2]
items:
[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]]

we choose the input items as our start solution. it's:
[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]]

```

and for each iteration, it will find neighbourhoods of current solution

```

iteration 0:
-----

the current solution is:
[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]]
for this solution, the total value is: 8

and all it's feasible 4 neighbours are:

solution is:
[[2, 3, -1], [5, 3, 0], [2, 2, 0], [8, 5, -1], [1, 1, -1]]
value is 7

solution is:
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, -1]]
value is 10

solution is:

```

and its output is

```

-----

the current solution is:
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, 1]]
for this solution, the total value is: 11

and all it's feasible 0 neighbours are:

so we choose the neighbour
[]
as our next solution, value is 0

but we have found that even the best neighbour is not better than the current solution, so
we end the loop

*****
<-algorithm end

 ([[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, 1]], 11)

```

Part II

3 Tabu-search

3.1 mainly description for the algorithm

On the base of the neighbourhood search, we record the best solution when we search the neighbours. The difference is when we found that best neighbour is not better than the current one, we continue the algorithm until it reached the iteration times we provided. Also a tabu list will be used, we add the current solution into the list, and when the list is full, we replace the oldest solution in the list with the current one. all solutions in tabu list will be removed from the neighbours list.

3.2 implementation step

step 1: Choose a suitable solution as the start solution.

step 2: for each iteration, choose the best from those neighbourhoods are not included in the tabu list as the solution for next iteration. Redo this step continually.

step 3: If the iteration times equals the given times, or there are no feasible alternative neighbours, end the algorithm.

3.3 descriptions for some criteria

neighbourhood definition: On the base of the current solution, move one item to a certain knapsack. After removing a lowest value per unit item from this knapsack(if need), if there is enough space for this new item, then it is neighbourhood of the current solution.

tabu list length: It was designed as a parameter that passed by user, that means we can choose to change the tabu list length to get different results.

termination criteria: If the iteration times equals the given times, or there are no feasible alternative neighbours, end the algorithm.

3.4 pseudo code

```
1 def tabuSearch(knapsacks, items, iterTimes, tabuLength):
2
3     currentBestValue = 0
4     # record the total value with current solution
5     bestNeighbourhoodValue = 0
6     # record the best values of all neighbourhoods
7     for item in items:
8         if item[2] != -1:
9             currentBestValue += item[0]
10    bestNeighbourhoodValue = currentBestValue
11    bestSolution = []
12    for item in items:
13        bestSolution.append(item[:])
14    tabuList = []
15    for i in range(iterTimes):
16        solution = []
17        for item in bestSolution:
18            #deep copy
19            solution.append(item[:])
20            if i < tabuLength:
21                tabuList.append(solution)
22        else:
23            tabuList[i % tabuLength] = solution
24        nbhs = neighbourhood(knapsacks, bestSolution)
```

```

25     for tabuSolution in tabuList:
26         # remove solutions in tabu list
27         for nbh in nbhs:
28             if tabuSolution in nbh:
29                 nbhs.remove(nbh)
30     bestSolution, bestNeighbourhoodValue = bestNeighbourhood(
nbhs, knapsacks, items)
31     if currentBestValue < bestNeighbourhoodValue:
32         items = []
33         for item in bestSolution:
34             items.append(item[:])
35         currentBestValue = bestNeighbourhoodValue
36
37     return items, currentBestValue

```

3.5 code & comment

```

1 def tabuSearch(knapsacks, items, iterTimes, tabuLength):
2     print("\033[1;35mneighbourhood search algorithm for multiple
knapsacks ---tabu search start->")
3     print('input:')
4     print('knapsacks: \n{}'.format(knapsacks))
5     print('items: \n{}'.format(items))
6     print('iteration times: \n{}'.format(iterTimes))
7     print('length of tabu list: \n{}\n\033[0m'.format(tabuLength))
8
9     currentBestValue = 0
10    # record the total value with current solution
11    bestNeighbourhoodValue = 0
12    # record the best values of all neighbourhoods
13    for item in items:
14        if item[2] != -1:
15            currentBestValue += item[0]
16    bestNeighbourhoodValue = currentBestValue
17    bestSolution = []
18    for item in items:
19        bestSolution.append(item[:])
20
21    print("\033[1;36mwe choose the input items as our start
solution. it's: \n{}\033[0m".format(items))
22    input()
23
24    tabuList = []
25    for i in range(iterTimes):
26        print("\033[1;32miteration {}: ".format(i+1))
27        print("-----\n\033[0m")
28        print("and the current solution is: \n{}".format(
bestSolution))
29        print("for this solution, the total value is: {}\n\033[0m".
format(bestNeighbourhoodValue))
30
31        solution = []
32        for item in bestSolution:
33            #deep copy
34            solution.append(item[:])
35        if i < tabuLength:
36            tabuList.append(solution)

```

```

37     else:
38         tabuList[i % tabuLength] = solution
39         print("update the tabu list")
40         print("and now the list is :\n")
41         for tabu in tabuList:
42             print("\033[1;31m{}\033[0m".format(tabu))
43         nbhs = neighbourhood(knapsacks, bestSolution)
44         for tabuSolution in tabuList:
45             # remove solutions in tabu list
46             for nbh in nbhs:
47                 if tabuSolution in nbh:
48                     nbhs.remove(nbh)
49         print("\nand all it's feasible(enough capacity and also not
in tabulist) {} neighbours are:".format(len(nbhs)))
50         input()
51         if len(nbhs) == 0:
52             print("\033[1;31mthere are no solution for next
iteration. so we end the loop\033[0m")
53             break;
54         for nbh in nbhs:
55             print("solution is: \n{}".format(nbh[0]))
56             print("value is {}\n".format(nbh[1]))
57             input()
58             bestSolution,bestNeighbourhoodValue = bestNeighbourhood(
nbhs, knapsacks, items)
59             print("\033[1;34mso we choose the neighbour \n{}\nas our
next solution, value is {}\033[0m".format(bestSolution,
bestNeighbourhoodValue))
60
61             print("\033[1;36mthe best solution now is: \n{}\nthe total
value is: {}\n".format(items,currentBestValue))
62
63             if currentBestValue < bestNeighbourhoodValue:
64                 items = []
65                 for item in bestSolution:
66                     items.append(item[:])
67                 currentBestValue = bestNeighbourhoodValue
68                 print("\033[1;34mso we update the best solution, now
the best value is {}\n\033[0m".format(currentBestValue))
69             else:
70                 print("\033[1;33mso we do not update the best solution
\033[0m")
71             print("\033[1;35m\n*****\033[0m")
72             print("\033[1;35m<-algorithm end\n\033[0m")
73
74         return items,currentBestValue

```

3.6 correctness check & test

Like above saying that we can observe the process of the program. we choose the output of the greedy algorithm as the start solution.

```

neighbourhood search algorithm for multiple knapsacks ---tabu search start->
input:
knapsacks:
[5, 1, 2]
items:
[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]]
iteration times:
15
length of tabu list:
2

we choose the input items as our start solution. it's:
[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]]

```

in one iteration, the tabu list is

```

iteration 2:
-----

and the current solution is:
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, -1]]
for this solution, the total value is: 10

update the tabu list
and now the list is :

[[2, 3, -1], [5, 3, 0], [2, 2, 2], [8, 5, -1], [1, 1, 0]]
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, -1]]

and all it's feasible(enough capacity and also not in tabulist) 1 neighbours are:

solution is:
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, 1]]
value is 11

|

```

and finally its output is

```

iteration 3:
-----

and the current solution is:
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, 1]]
for this solution, the total value is: 11

update the tabu list
and now the list is :

[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, 1]]
[[2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, -1]]

and all it's feasible(enough capacity and also not in tabulist) 0 neighbours are:

there are no solution for next iteration. so we end the loop

*****
<-algorithm end

([2, 3, -1], [5, 3, -1], [2, 2, 2], [8, 5, 0], [1, 1, 1], 11)

```