# #1 Assignment - CMPT 405

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#1 - Let C be the array containing all the possible coins  $\{1, 5, 10, 25, 100, 200\}$ . Let V be the total change value.

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Algorithm:  \begin{split} &\textbf{Input:} \ \ C, \ V \\ & d \leftarrow \text{sort C such that } d_1 \geq d_2 \geq ... \geq d_n \\ & res \leftarrow \emptyset; \ i \leftarrow 1 \\ & \textbf{while } V > 0 \ \textbf{do} \\ & \textbf{if } V \geq d_i \ \textbf{then} \\ & n_{coins} \leftarrow \left\lfloor \frac{V}{d_i} \right\rfloor \\ & V \leftarrow V - (n_{coins} * d_i) \\ & res \leftarrow res \cup \{(d_i, n_{coins})\} \\ & \textbf{end if} \\ & i \leftarrow i + 1 \\ & \textbf{end while} \\ & \textbf{return } res \end{split}
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Intuition: For all  $d_i, d_j \in C$ ,  $1 \le i < j \le n$ ,  $d_i \ge 2 * d_j$ , meaning that if I the algorithm chooses any  $d_j$  over any  $d_i$ , it will have to pick at least 2 times more coins for some value V that satisfies both  $d_i$  and  $d_j$ .

*Proof.* The only way to give more coins than the smallest possible number of coins for any change would be in a case where the algorithm chooses  $d_j$  over a  $d_i$  (see *Intuition*). Now, imagine that the algorithm chooses  $d_j$  over  $d_i$ , then we have two cases:

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- Case 1. d_i > V:
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If so, we are done because there are no possible ways of choosing  $d_i$  for value V.

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- Case 2. d_i \leq V:
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If that was the case (as a mean of contradiction), we would have an optimal set OPT such that  $OPT_{i-1} \cup d_j \subseteq OPT$ , which is not the case, once the iteration i will happen before the iteration j, causing the algorithm to choose  $d_i$  over  $d_j$  (and never the opposite) for any value V that satisfies both  $d_i$  and  $d_j$ .

#### #2 a)

Greedy approach to the fractional knapsack:

- n objects and a knapsack
- item i weighs  $w_i > 0$  and has utility  $u_i > 0$
- fill knapsack so as to  $\mathbf{maximize}$  total utility/weight, not exceeding total capacity W

### Algorithm approach:

- sort items in decreasing order of their utility-to-weight ratio  $u_i/w_i$
- repeatedly add item with max ratio  $u_i/w_i$ . If not possible to add the whole object, add a fraction  $\alpha \in (0,1)$  of it, if possible.

*Proof.* Let  $K_{opt} \subseteq \{i_1, i_2, i_3, ..., i_n\}$  be the optimal set of items in a knapsack and let  $K_j$  be the chosen items after an iteration j,  $0 \le j \le n$ . Let  $K_j$  be considered "promising" if  $K_j \subseteq K_{opt}$ .

Base case:  $K_0$ :  $K_0$  is promising since the total number of chosen objects, in this case none, does not exceed total capacity W. Thus, there exists some optimal  $K_{opt}$  such that  $K_0 \subseteq K_{opt} \subseteq K_0 \cup \{i_1, i_2, ..., i_n\}$ .

Induction step: Assume  $K_{j-1}$  is promising for stage j-1, meaning that  $K_{j-1} \subseteq K_{opt} \subseteq K_{j-1} \cup \{i_j, i_{j+1}, ..., i_n\}$ . We want to show  $K_j$ . On a stage j we have two cases:

Case 1.  $i_j$  is rejected. Then  $K_{j-1} \cup \{i_j\}$  or  $K_{j-1} \cup \{i_j * \alpha\}$  (any fraction  $\alpha \in (0,1)$  of  $i_j$ ) exceed the capacity W; thus,  $K_{j-1} = K_j$ . Since  $K_{j-1} \subseteq K_{opt}$  and  $K_{opt}$  does not exceed the total capacity W,  $i_j \notin K_{opt}$ . So  $K_j \subseteq K_{opt} \subseteq K_j \cup \{i_{j+1}, i_{j+2}, ..., i_n\}$ .

Case 2.  $i_j$  or  $i_j * \alpha$  is added to  $K_{j-1}$ . Let item  $i_{chosen}$  be  $i_j$  or  $i_j * \alpha$  (whichever was added to  $K_{j-1}$ ). Then  $K_{j-1} \cup \{i_{chosen}\}$  does not exceed the total capacity W and we have  $K_{j-1} \cup \{i_{chosen}\} = K_j$ .

Case 2.1.  $i_{chosen} \in K_{opt}$ . Then we have  $K_j \subseteq K_{opt} \subseteq K_j \cup \{i_{j+1}, i_{j+2}, ..., i_n\}$ .

Case 2.2.  $i_{chosen} \notin K_{opt}$ . We show that there is another maximum set of utility-to-weight items  $K'_{opt}$  that witnesses the fact that  $K_j$  is promising. For example, consider an item  $i_{chosen}$  added to  $K_{opt}$ . This will exceed the capacity W and the knapsack will contain at least one item of  $\{i_{j+1}, i_{j+2}, ..., i_n\}$ .

Proof of claim (Case 2.2):  $K_{opt}$  contains all elements of  $K_{j-1}$  and can be obtained from  $K_{j-1}$  by adding some items from the set  $\{i_j, i_{j+1}, ..., i_n\}$ . Adding  $i_j$  does not exceed capacity W, so the excess in  $K_{opt} \cup \{i_{chosen}\}$  must contain some elements other items in  $\{i_{j+1}, i_{j+2}, ..., i_n\}$ .

#2 b)

$\pi = 0$				
item	utility	weight		
1	2	1	W = 1000	
2	1000	1000		

 $K_{opt} = \{i_2\}$  (utility = 1000),  $K_{greedy} = \{i_1\}$  (utility = 2)

## #3

The idea here is to use the greedy approach to check ahead and count the number of tiles  $a_{xy}$ ,  $1 \le i < x \le k$ ,  $1 \le y \le n_x$  adjacent to a tile  $a_{ij}$ . Basically, if a tile  $a_{ij}$  touches the bounds of a tile  $a_{xy}$ , we say they are adjacent. The MAXnumber of adjacent tiles is then then minimum required number of colors for our solution. For example, if we have the tiles  $a_{11} = 0.6$ ,  $a_{12} = 0.4$  and  $a_{21} = 0.35$ ,  $a_{22} = 0.35$ ,  $a_{23} = 0.3$ , then  $a_{11}$  is adjacent to  $a_{21}$  and  $a_{22}$  and  $a_{12}$  is adjacent to  $a_{22}$  and  $a_{23}$ . The minimum number of colours required would then be 4 in this case.

### Algorithm:

```
Input: wall, k
  for i from 1 to k do
      for j from 1 to n_i do
           wall_i \leftarrow \text{sort tiles such that } a_{ij} \geq a_{ij+1} \geq ... \geq a_{in_i}
      end for
  end for
  C \leftarrow 0
  for i from 1 to k-1 do
      c \leftarrow 0
      for j from 1 to n_i do
          x \leftarrow i+1
          for y from 1 to n_x do
               if a_{ij} adjacent to a_{xy} then
                   c \leftarrow c + 1
               else
                   break //not adjacent to a_{xy+1} through a_{xn_x} as well
               end if
          end for
      end for
      C \leftarrow MAX(C,c)
  end for
  return C
Counterexample:
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Suppose:
col\ 1: a_{11} = 0.6, a_{12} = 0.4
col \ 2: a_{21} = 0.4, \ a_{22} = 0.4, \ a_{23} = 0.2
```

Our Greedy algorithm selects  $a_{11}$  as being adjacent to  $a_{21}$  and  $a_{22}$ . After that, it selects  $a_{12}$  as being adjacent to  $a_{22}$  and  $a_{23}$ , returning 4 as the minimal number of colours. The algorithm could not rearrange  $a_{22}$  and  $a_{23}$  so that  $a_{11}$  would be adjacent to  $a_{21}$  and  $a_{23}$ ; and  $a_{12}$  would be adjacent to  $a_{22}$ , returning 3, the optimal number of colors for this problem. Thus, the Greedy approach given is not optimal. See below:

```
W_{greedy} = \{(a_{11}, a_{21}), (a_{11}, a_{22}), (a_{12}, a_{22}), (a_{12}, a_{23})\}. min: 4 colors W_{opt} = \{(a_{11}, a_{21}), (a_{11}, a_{23}), (a_{12}, a_{22})\}. min: 3 colors
```

#### #4

Definition: For  $0 \le i \le m$ ,  $0 \le j \le n$ , define M[i,j] as the optimal number of paths in the Cartesian plane from (0,0) to (i,j) that uses the combined number of steps of type U(up, M[i-1,j]), R(right, M[i,j-1]) and D(diagonal, M[i-1,j-1]). The optimal number of paths from (0,0) to (m,n) is then M[m,n].

Recurrence:

$$M[i,j] = \begin{cases} 1 & \text{if } i=0 \text{ or } j=0 \\ M[i-1,j]+M[i,j-1]+M[i-1,j-1] & \text{otherwise} \end{cases}$$

Algorithm:

```
Input: m, n

Make matrix M with dimensions m \times n

for i from 0 to m do

M[i,0] \leftarrow 1

end for

for j from 0 to n do

M[0,j] \leftarrow 1

end for

for i from 1 to m do

for j from 1 to m do

M[i,j] \leftarrow M[i-1,j] + M[i,j-1] + M[i-1,j-1]

end for
end for
end for
return M[m,n]
```

Running time: The running time of the loops are respectively O(m), O(n) and O(mn). All operations on M (inside the loops) are constant O(1) and therefore the total running time of the algorithm is O(mn).

### Expression:

For d steps of type D(diagonal), there must have m-d steps of type U and n-d steps of type R, in order to reach (m,n). It can then be represented by:

$$M(m,n) = \sum_{d=0}^{\min(m,n)} \binom{m+n-d}{m} \binom{m}{d}$$

#5

The idea of the problem is to use the Greedy approach on the **set cover** problem to get maximal rectangles, meaning that if I have a position  $p_{v_i} = 0$ , where v is a leaf of T and i is an index of u's binary string, and I also have  $p_{u_i} = 0$ , for a leaf u sibling of v, then I can maximize a rectangle for a w parent of v and u, such that  $p_{w_i} = 0$  (the same holds for sequences of 0's). We start the algorithm by setting up the sets for the set cover problem such that the longest number of consecutive zeroes and ones stay make a group within a string (e.g  $S_v$  for v = 1011000 would be  $S_v = \{1, 0, 11, 000\}$ . We then traverse T such that we get maximal rectangles, see the algorithm below:

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Algorithm:
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```
Input: T
  for each v \in T, children(v) = \emptyset do
       S_v \leftarrow \text{group 0's and 1's} \in v \text{ as a list, such that no 0 in position } i \text{ is followed}
  by another (group of) 0 in position i+1 and no 1 in position i is followed by
  another (group of) 1 in position i+1
  end for
  for each w \in T_{postorder}, children(w) = \{u, v\} do
       S_w \leftarrow \text{matching-indexes 0's in both } S_v \text{ and } S_u
       S_v \leftarrow S_v - S_w
       S_u \leftarrow S_u - S_w
  end for
  Res \leftarrow \emptyset
  for each v \in T do
       if S_v \neq \emptyset then
            rectangle \leftarrow \forall (v, (i, j)) \in S_v \text{ where } p_{v_i j} = \{0\}^{|v|}
            Res \leftarrow Res \cup rectangle
       end if
  end for
  return Res
```

## Running time:

Let b be the number of bits in a binary string on the leaves of T. The algorithm then takes O(nlogn) to traverse T and O(b) for each set-related operation inside the loops. Therefore, the total running time of the algorithm is O(bnlogn).