

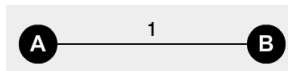
## #3 Assignment - CMPT 405

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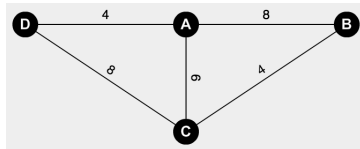
### #1a)

Let  $G_1$  be a graph with two vertices  $A$  and  $B$  and an edge  $(A, B)$  with weight 1. For every shortest path tree  $T_v$ ,  $v \in V$ ,  $T_v$  is also a MST (it is easy to see, as there is only one tree).



### #1b)

Let  $G_2$  be a graph with vertices  $A, B, C$  and  $D$  and edges  $(A, B)$ ,  $(A, D)$ ,  $(A, C)$ ,  $(B, C)$  and  $(C, D)$ , with weights 8, 4, 6, 4 and 8, respectively. Then, no shortest path tree  $T_v$  given by Dijkstra's algorithm is a MST.



MST =  $(A, C), (A, D), (B, C)$

$T_a = (A, B), (A, C), (A, D)$

$T_b = (A, B), (A, D), (B, C)$

$T_c = (A, C), (B, C), (C, D)$

$T_d = (A, B), (A, D), (C, D)$

### #2a)

Intuition: Considering an unweighted graph  $G = (V, E)$ , every edge comprises same weight (let us say weight 1) and therefore, the least number of edges used, the better. For example while the path  $i \rightarrow j$  will have total distance 1 by summing through all weights in the path, the path  $i \rightarrow \dots \rightarrow k \dots \rightarrow j$  will have at least total weight equal to 2 (in case that the path is only  $i \rightarrow k \rightarrow j$ ).

(1) Because of the intuition above, it is correct to say that, for a distance  $d_{ij}$ , every edge used in  $d_{ij}$  will corroborate with exactly 1 (their weight) in the distance of the path  $i$  through  $j$ . Let  $(i, k) \in E$  be such edge. Then,  $d_{ij} \leq d_{ik} + d_{kj}$ ,

where  $d_{ik} = 1$ . Therefore,  $d_{ij} \leq 1 + d_{kj}$  and  $d_{ij} - 1 \leq d_{kj}$ .

(2) Likewise, we do the same for  $d_{kj} \leq d_{ki} + d_{ij}$ , as  $(i, k) = (k, i) \in E$ . So,  $d_{kj} \leq 1 + d_{ij}$ .

Finally, we can join notions (1) and (2) so that we prove that  $d_{ij} - 1 \leq d_{kj} \leq d_{ij} + 1$ .

### #2b)

To show that if  $d_{ij}$  is even, then  $d'_{kj} \geq d_{ij}$ , we have the cases:

**Case 1:**  $d_{ij} = 0$ , then  $d'_{ij} = 0$  and  $d'_{kj} \in \{0, 1, 2\} \geq d'_{ij}$

**Case 2:**  $d_{ij}$  is even and  $d_{ij} \geq 2$ , then  $d'_{ij} = 0$  and  $d'_{kj} \in \{0, 1, 2\} \geq d'_{ij}$

**Case 3:**  $d_{ij} = 2$ , then  $d'_{ij} = 2$ , and  $d'_{kj} = 2$ , thus  $d'_{kj} \geq d'_{ij}$ .

To show that if  $d_{ij}$  is odd, then  $d'_{kj} \leq d'_{ij}$  and  $d'_{kj} < d'_{ij}$  for at least one  $k$ . Because of distance matrix  $D'$ , we have two cases:

**Case 1:**  $d_{ij} = 1$ , then it is proven by part a that  $d'_{kj} \leq d'_{ij}$ , because of the symmetry in the distance matrix it holds that for at least one  $k$ ,  $d'_{kj} < d'_{ij}$ .

**Case 2:**  $d_{ij}$  is odd and  $d_{ij} \geq 3$ . Then  $d'_{ij} = 0$  and therefore any  $d'_{kj}$  is also equal to zero, thus  $d'_{kj} \leq d'_{ij}$  holds.

### #2c)

### #2d)

### #3)

The idea of the algorithm is to use an approach similar to the Floyd-Warshall algorithm for transitive closures, as transitive reductions are essentially the inverse of transitive closures. In Floyd-Warshall, we augment the number of edges whenever we find a path  $i \rightarrow k \rightarrow j$  with total weight less than the weight of the path  $i \rightarrow j$ . However, in transitive reduction, we are interested in minimizing the number of edges that are used, thus, we always prefer paths  $i \rightarrow j$  than  $i \rightarrow k \rightarrow j$ . The algorithm is very straightforward and can be thought as a modification of Floyd-Warshall and has same complexity ( $O(n^3)$ ):

Algorithm:

**Input:**  $G = (V, E)$

$E' \leftarrow \text{copy } E$

**for**  $i$  **from** 1 to  $n$  **do**

**for**  $j$  **from** 1 to  $n$  **do**

**for**  $k$  **from** 1 to  $n$  **do**

**if**  $\{(i, j), (i, k), (k, j)\} \subseteq E'$ , where  $(i, j) \neq (i, k) \neq (k, j)$  **then**

$E' \leftarrow E' - (i, k)$

**end if**

**end for**

**end for**

**end for**  
**return**  $G' = (V, E')$

**#4)**

Let  $T$  be the unrooted tree decomposition of  $G$  and  $T'$  be a nice tree decomposition of  $T$ . The idea of the algorithm is to make any node of  $T$  (preferably a internal node with many edges or minimum bag width) its root. So to ease the algorithm, we preprocess the input  $T$ : after we root  $T$ , we go through all the bag leaves  $b$  in  $T$  and create a new bag for every  $v_b \in (b - \text{parent}(b))$  and make  $b$  their parent. We then run a postorder traversal on  $T$  and apply the following rules:

- **Case 1.** If bag  $b$  is a leaf in  $T$ :

If  $b \neq \text{parent}(b)$ , it must have come from the preprocessing of the original bag  $b' - \text{parent}(b')$ , and therefore  $|b| = 1 \leq |\text{parent}(b)|$ , meaning that we need to add introduce nodes to our nice tree decomposition  $T'$  by adding some vertices  $v_{\text{parent}} \in \text{parent}(b)$  and stop when they have the same elements, so we can join the bags later; otherwise, if  $b$  and  $\text{parent}(b)$  have the same elements, we are done.

- **Case 2.** If bag  $b$  is an internal node in  $T$ :

We know that if  $b$  is an internal node, then  $b$  is a parent of at least one bag  $b'$ . Then, the first step is to add a join node in  $T'$  for  $b$  and every  $b'$ , where  $\text{parent}(b') = b$ . Also, because  $b$  is an internal node,  $b$  has a parent. We need first to get rid of all nodes in  $b$  that are not elements of  $\text{parent}(b)$  (by adding forget nodes to  $T'$ ) and finally add some vertices  $v_{\text{parent}} \in \text{parent}(b)$  and stop when  $b$  and  $\text{parent}(b)$  have the same elements, so we can join the bags later (by adding introduce nodes to  $T'$ ).

- **Case 3.** If bag  $b$  is the root in  $T$ :

Finally, if  $b$  is the root, we only need to create a join node of all bags  $\text{children}(b)$  in our final nice tree decomposition  $T'$ .

The algorithm runs in  $O(nk)$  (as per the pseudocode and demonstration below)

Algorithm:

**Input:**  $T$

Make any internal bag (or the bag with minimum width) of  $T$  its root.

**for each** bag  $b \in T, \text{children}(b) = \emptyset$  **do** // all leaves

$\text{free}_{vs} \leftarrow b - \text{parent}(b)$

**for each**  $v \in \text{free}_{vs}$  **do**

$T \leftarrow T \cup \{v\}$  such that  $\text{parent}(v) = b$

**end for**

**end for**

$T' \leftarrow \emptyset$

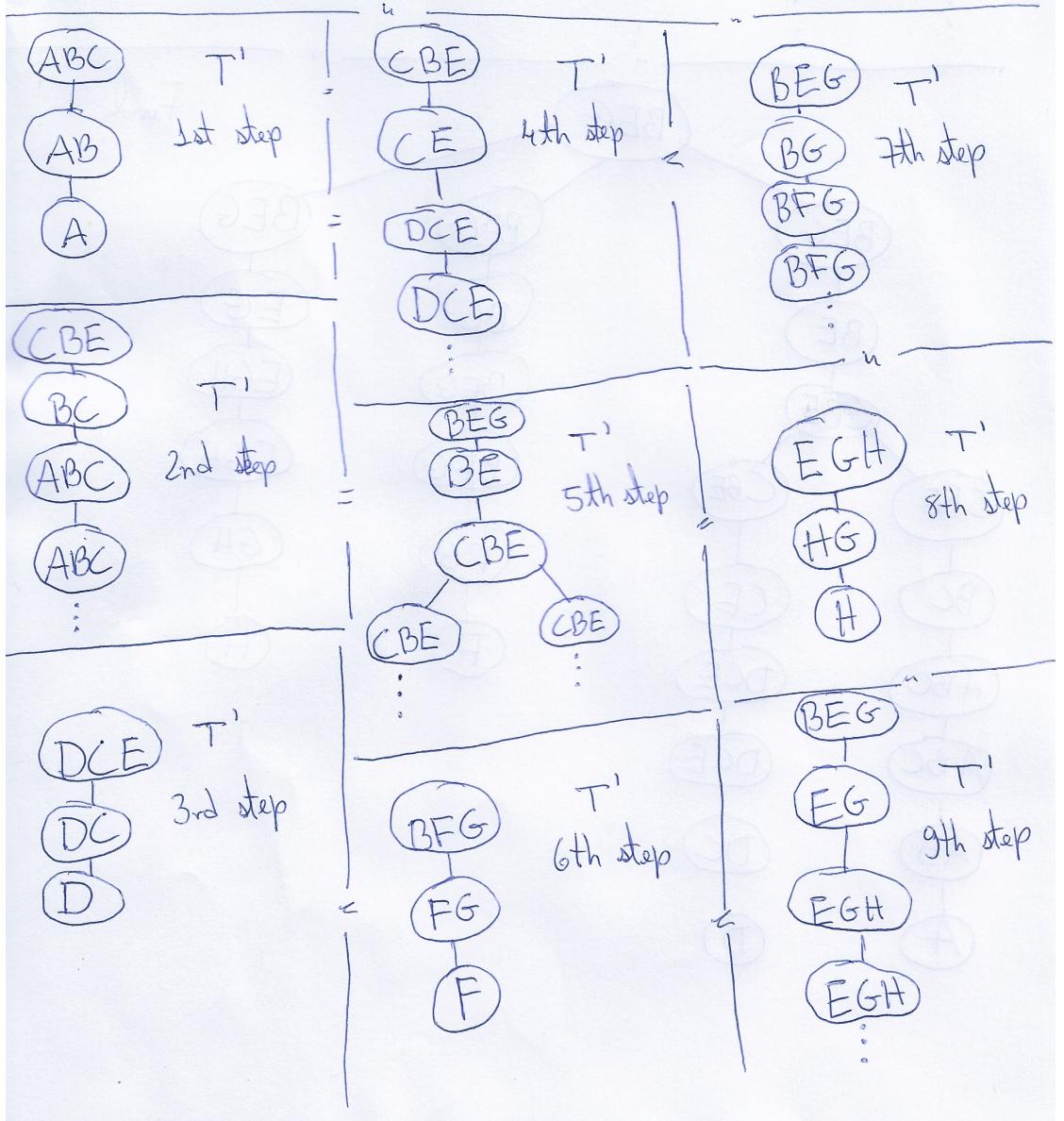
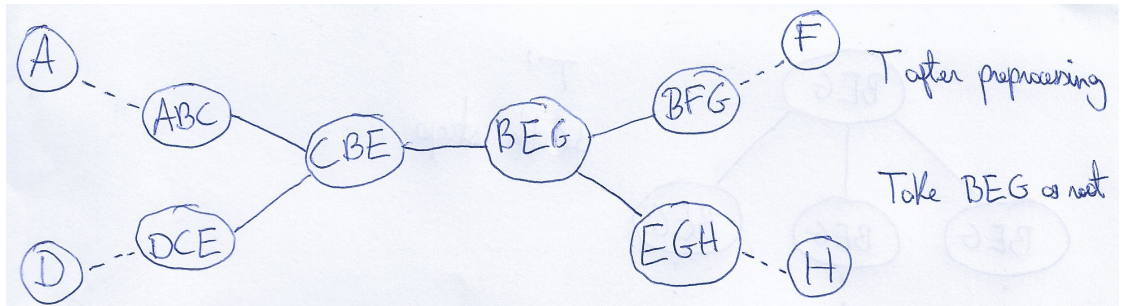
**for each** bag  $b \in T_{\text{postorder}}$  **do**

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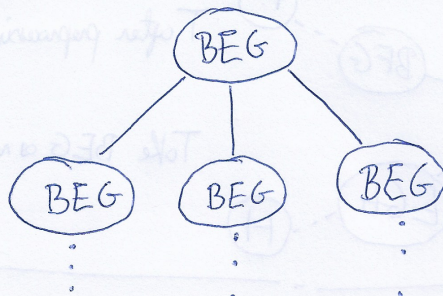
 $b_{aux} \leftarrow b$ 
if  $children(b) = \emptyset$  then // if leaf
     $T' \leftarrow T' \cup b_{aux}$  // add leaf node
    while  $b_{aux} \neq parent(b)$  do
        add introduce node  $b_{aux} \cup \{v_{parent}\}$  in  $T'$ , for any  $v_{parent} \in parent(b)$ ,
        such that  $b_{aux} \cup \{v_{parent}\} = parent(b_{aux})$ 
    end while
else
    Create join node in  $T'$ 
    if  $parent(b) \neq \emptyset$  then // if not root
        while  $\exists v \in b$ , such that  $v \notin parent(b)$  do
            add forget node  $b_{aux} - \{v\}$  in  $T'$ , for any  $v \notin parent(b)$ ,
            such that  $b_{aux} - \{v\} = parent(b_{aux})$ 
        end while
        while  $b_{aux} \neq parent(b)$  do
            add introduce node  $b_{aux} \cup \{v_{parent}\}$  in  $T'$ ,
            for any  $v_{parent} \in parent(b)$ , such that  $b_{aux} \cup \{v_{parent}\} = parent(b_{aux})$ 
        end while
    end if
end if

end for
return  $T'$ 

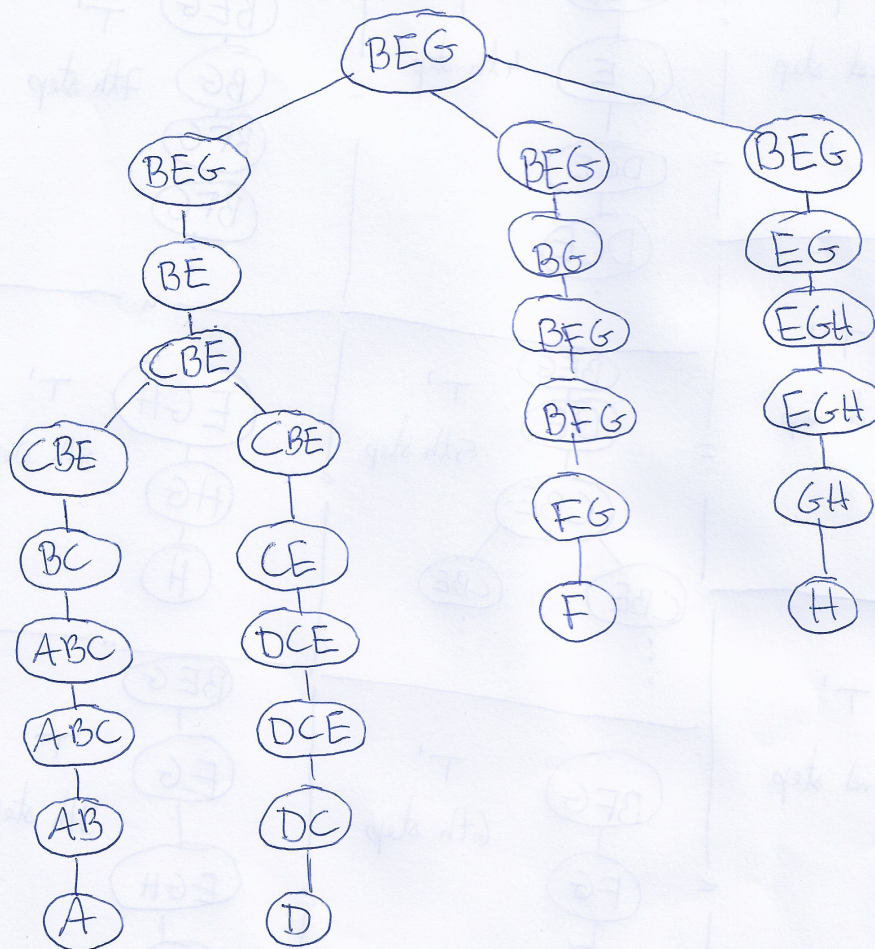
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$T'$   
10th step



Find  $T'$

**Note.:** I know joins are only possible with two bags. It was too late to fix that when I realized I did this mistake in my examples and algorithm on #4, please ignore it. Imagine we made  $T$  a binary tree and worked from there (check image on question #5).

#5)

*Definition:* (As per lecture 17). Let  $B_x$  be the vertices appearing in node  $x$  and let  $V_x$  be the vertices in the subtree rooted at  $x$ . Let  $M[x, S]$  be a matrix that keeps the size a maximum independent set  $I \subseteq V_x$  with  $I \cap B_x = S$ . At the end of the algorithm, the maximum independent set size will then be on  $M[root, S'_{root}]$ , where  $S'_{root}$  is the best  $S$  for the root.

*Recurrence:*

- **Leaf:**  $B_x$  has no children

$$M[x, S] = 1$$

- **Introduce:** 1 child  $y$ ,  $B_x = B_y \cup \{v\}$ , for some vertex  $v$

$$M[x, S] = \begin{cases} M[y, S] & \text{if } v \notin S \\ M[y, S - \{v\}] + |v| & \text{if } v \in S \text{ but } v \text{ has no neighbor in } S \\ -\infty & \text{if } S \text{ contains } v \text{ and its neighbors} \end{cases}$$

- **Forget:** 1 child  $y$ ,  $B_x = B_y - \{v\}$ , for some vertex  $v$

$$M[x, S] = \max(M[y, S], M[y, S \cup \{v\}])$$

- **Join:** 2 children  $y_1, y_2$ ,  $B_x = B_{y_1} = B_{y_2}$

$$M[x, S] = M[y_1, S] + M[y_2, S] - |S|$$

Algorithm:

**Input:**  $T', root$

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for each  $x \in T'_{postorder}$  do
  for each  $I \subseteq V_x$  with  $S = I \cap B_x$  do
    compute recurrence  $M[x, S]$  as described above
    keep best  $S$  in  $S'$  for every  $x$ 
  end for
end for
return  $M[root, S'_{root}]$ 

```

*Demonstration:* There are at most  $2^{k+1} \times n$  subproblems (entries) in  $M$  and therefore its running time is bounded on  $M$ . Demonstrate that would take a matrix of size  $2^4 \times 25$ , which is very large. Thus, the image below summarizes it:

