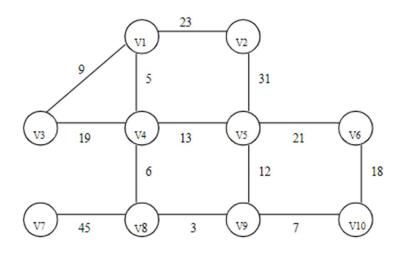
## CS.3310 Homework 3

1. Construct Minimum Spanning Tree (MST) using Prim's algorithm and find the cost of that MST



\*\*\*Notation – v12 = 23 means edge weight from v1 to v2 is 23

Step 1 – Start from v1, add it to the SetMST and compute adjacent edges weight.

V12 = 23; v13 = 9; v14 = 5. Connect v1 and v4

Step 2 – Travel to v4, add it to SetMST, and compute adjacent edges weight.

V12 = 23; v13 = 9; v45 = 13; v48 = 6. Connect v4 and v8

Step 3 – Travel to v8, add it to SetMST and compute adjacent edges weight.

V12 = 23; v13 = 9; v45 = 13; v87 = 45; v89 = 3. Connect v8 and v9

Step 4 – Travel to v9, add it to SetMST and compute adjacent edges weight.

V12 = 23; v13 = 9; v45 = 13; v87 = 45; v95 = 12, v910 = 7. Connect v9 and v10

Step 5 – Travel to v10, add it to SetMST and compute adjacent edges weight.

V12 = 23; v13 = 9; v45 = 13; v87 = 45; v95 = 12, v106 = 18. Connect v1 and v3

Step 6 – Travel to v3, add it to SetMST and compute relevant adjacent edges weight.

V12 = 23; v45 = 13; v87 = 45; v95 = 12, v106 = 18 Connect v9 and v5

Step 7 – Travel to v5, add it to SetMST and compute relevant adjacent edges weight.

V12 = 23; v87 = 45; v106 = 18; v56 = 21. Connect v10 to v6

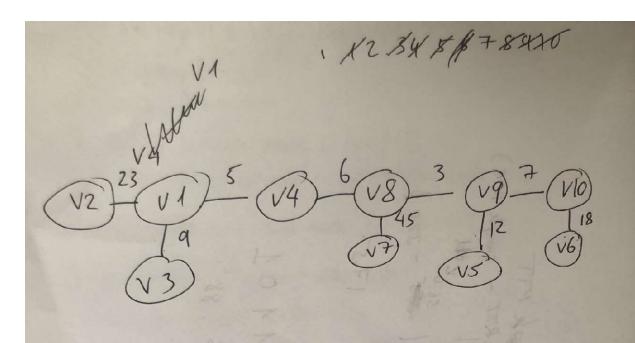
Step 8 – Add v6 to SetMST and check remaining edges weight.

V12 = 23; v87 = 45. Connect v1 to v2

Step 9 – Add v2 to SetMST and check remaining edges weight.

V87 = 45. Connect v8 to v7. Completed MST as shown below with 9 edges

Total cost/weight = 23+5+9+6+45+3+12+7+18 = 128



## 2. Construct MST using Kruskal's algorithm and cost of that MST

Sort the edges

$$(v8,v9) = 3$$

$$(v1,v4) = 5$$

$$(v4,v8) = 6$$

$$(v9, v10) = 7$$

$$(v1, v3) = 9$$

$$(v5,v9) = 12$$

$$(v4, v5) = 13 X$$

$$(v6, v10) = 18$$

$$(v3, v4) = 19 X$$

$$(v5, v6) = 21 X$$

$$(v1, v2) = 23$$

$$(v2, v5) = 31 X$$

$$(v7, v8) = 45$$

Step 1 – connect v8 and v9. New set

Step 2 – connect v1 and v4. New set

Step 3 – connect v4 and v8. Since they came from different sets, there's no cycle created.

Step 4 – connect v9 and v10. Since they came from different sets, there's no cycle created.

Step 5 – connect v1 and v3. Since they came from different sets, there's no cycle created.

Step 6 – connect v5 and v9. Since they came from different sets, there's no cycle created.

Step 7 – discard (v4, v5) because a cycle would be created

Step 8 – connect v6 and v10. Since they came from different sets, there's no cycle created.

Step 9 – discard (v3, v4), (v5, v6) because a cycle would be created.

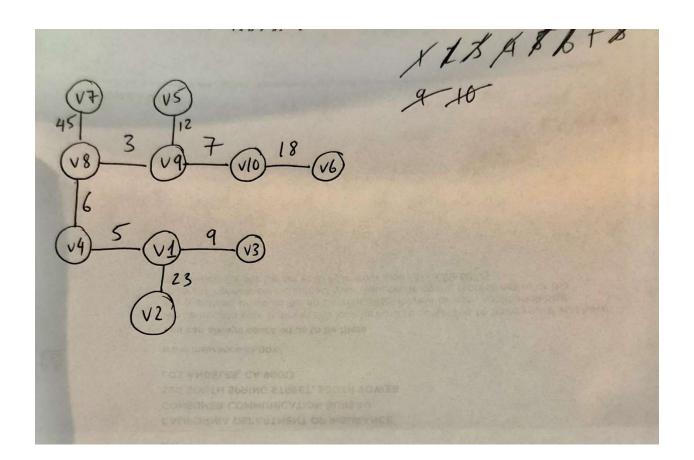
Step 10 – connect v1 and v2. Since they came from different sets, there's no cycle created.

Step 11 – discard (v2, v5) because a cycle would be created.

Step 12 – connect v7 and v8. Since they came from different sets, there's no cycle created

MST completed and as shown below with 9 edges

**Total cost** = 3+5+6+7+9+12+18+23+45 = 128



3. Imagine that the objects and their weights are as below with knapsack M = 10

Object	I <sub>1</sub>	$I_2$	I <sub>3</sub>
Weight	5	6	7
Profit	11	18	20
Profit/weight	2.2	3	2.857

The order based on nonincreasing profit/weight is I<sub>2</sub>, I<sub>3</sub>, I<sub>1</sub>.

Step 1 - We add  $I_2$  to the knapsack. Then, the knapsack only have capacity of 4 left.

Step 2 – We can't add anymore items to the bag because their weights is greater than 3. This means that the knapsack utilization rate isn't 100% if it's a 0/1 Knapsack. In this case, the profit from this 0/1 knapsack is 18.

Profit = 
$$0*11 + 1*18 + 0*20 = 18$$

However, the knapsack problem in class allow for partial/fractional inclusion of items so the knapsack utilization rate is always 100%.

The optimal solution is ordering by profit/weight of I<sub>2</sub>, I<sub>3</sub>, I<sub>1</sub> just as above.

Step 1 – We add  $I_2$  to the knapsack. Then, the knapsack only have capacity of 4 left.

Step 2 – Since 4 is less than the weight of any object, we must add a fraction of an object. The next item is  $I_3$  so we add 4/7 of  $I_3$  to the knapsack. Then, the profit would be:

Profit = 
$$0*11 + 1*18 + 4/7*20 = 29.429$$

By profit comparison, 29.429 > 18 so the proposed 0/1 knapsack with the proposed strategy is not the optimal solution. This will be the usually be the case because 0/1 knapsack doesn't always have 100% utilization rate whereas the knapsack strategy discussed in class will always have 100% utilization rate which always maximizes profit.

- 4. Let An = { a1, a2, ..., an } be a finite set of distinct coin types (e.g., a1= 50 cents, a2= 25 cents, a3= 10 cents etc.). We assume each ai is an integer and that a1 > a2 > ... > an. Each type is available in unlimited quantity. The coin changing problem is to make up an exact amount C using a minimum total number of coins. C is an integer > 0
  - a. If  $a_n \neq 1$ , then for C value ending in 1's such as  $\{1, 11, 21, 31, ...\}$ , there is no solution because  $a_n \neq 1$
  - b. The algorithm is as below #pass the amount to get changed for C, and A is the finite set of dis

#pass the amount to get changed for C, and A is the finite set of distinct coin type Def numCoin (C, A):

```
Count = 0
numOfCoin = 0
n = 1
while (C > 0):
#find greatest amount of coin possible for this type of denomination
count = C // a<sub>n</sub>
#find remaining value after
C -= (count*a<sub>n</sub>)
#update values
numOfCoin += count
n += 1
```

return numOfCoin

- c. A counter example would be to consider the amount of 55 and  $A_n = \{10,9.8.7,6,1\}$  The algorithm in part (b) would yield the number of coins is 10 coins used consisting of five 10 coins and five 1 coins. However, there's another solution consisting of four 10 coins, one 9 coin, and one 6 coin, totaling six coins used instead of 10. The algorithm in part (b) uses more coins than the counterexample so part (b) algorithm isn't an optimal solution.
- d. If  $A_n = \{K^{n-1}, K^{n-2}, K^{n-2}, K^0\}$  where n > 1, we can substitute in some value and check. Suppose n = 5 and k = 2, we have  $A_n = \{16, 8, 4, 2, 1\}$ . Then, suppose we have the amount C = 55, the algorithm would yield three 16 coins, one 4 coin, one 2 coin, and one 1 coin, totaling 6 coins.

We know this is an optimal solution because the complexity when reducing is like a binary search tree where the amount of coins or steps require is approximately O(log n). The algorithm will always yield minimum number of coins.

Lg(55) = 5.78 rounded up to 6