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CX 4140  
Assignment 2

**Problem 1:**

**a)**

- (a)  $\Theta(g)$
- (b)  $\Theta(g)$
- (c)  $\Omega(g)$
- (d)  $O(g)$
- (e)  $\Omega(g)$
- (f)  $\Theta(g)$

**b)**

Outer loop iterates from 0 through n. Time complexity  $O(n)$ .

Inner loop iterates  $\log_3 n$  times. Time complexity  $O(\log n)$ .

Total Time Complexity  $O(n \cdot \log n)$ .

The algorithm computes and prints a number and the (log base 3 of that number + 1) floored.

**Problem 2:**

Greedy Solution:

Leaks as points  $l_1$  to  $l_n$ . Strip of size s.

Place a strip on  $l_i$  starting with  $i = 1$  therefore covering all leaks  $l_i$  to  $l_j$  between  $\{l_i, l_i + s\}$ , where  $l_j \leq l_i + s$

Update  $i$  to  $j + 1$

Repeat until  $i > n$

Runtime:  $O(n)$

Correctness:

Greedy Solution  $G = \{G_1, G_2 \dots G_n\}$  where  $G_i$  is point where strip s starts and  $G_i < G_j$  for  $i < j$ .

Optimal Solution  $O = \{O_1, O_2 \dots O_x\}$ , with same notation.

By definition of greedy property  $O_1 \leq G_1$  since  $G_1 = l_1$ .

Let  $O'$  be the solution obtained by replacing  $O_1$  with  $G_1$  in  $O$ . Since  $l_1$  is the first leak and is covered by  $G_1$ ,  $G_1$  covers all the leaks that  $O_1$  covers. Now moving on to the next leak/strip  $O_2$ , a similar case can be made where  $G_2$  covers all leaks covered by  $O_2$  if we assume that our problem is to now cover all the leaks to the right of our leftmost leak  $l_1$ . By repeating this process, we can convert  $O$  into  $G$ . Therefore  $G$  is optimal.

### Problem 3:

Bag with weight limit  $L$ .

$n$  minerals with value of  $v_i$  and weight  $w_i$  for  $i^{\text{th}}$  mineral.

Maximize bag value.

Greedy Solution:

Calculate the value-per-pound  $p_i = v_i/w_i$  for  $i = 1, 2, \dots, n$  minerals.

Sort the items by decreasing  $p_i$

Let  $k$  be the current weight limit (Initially,  $k = L$ ).

In each iteration, we choose item  $i$  from the head of the unselected sorted list.

If  $k \geq w_i$ ,

$x_i = 1$  (whole Item  $i$  taken)

$k = k - w_i$  (current weight limit updated)

If  $k < w_i$ ,

$x_i = k/w_i$  (fraction of Item  $i$  taken)

terminate/break out of algorithm.

Run time is  $O(n * \log n)$  ( $\rightarrow$  Run time of sorting algorithm)

Solution Correctness:

Let the greedy solution be  $G = \langle x_1, x_2, \dots, x_n \rangle$

$x_i$  indicates fraction of item  $i$  taken (all  $x_i = 1$ , except possibly for  $i = n$ ).

Consider any optimal solution  $O = \langle y_1, y_2, \dots, y_n \rangle$

$y_i$  indicates fraction of item  $i$  taken in  $O$  (for all  $i$ ,  $0 \leq y_i \leq 1$ ).

Knapsack is full in both  $G$  and  $O$ : For all  $i$   $\sum x_i * w_i = K = \sum y_i * w_i$  for all  $i$ .

Consider the first item  $i$  where the two selections differ.

By definition, solution  $G$  takes a greater amount of item  $i$  than solution  $O$

(because the greedy solution always takes as much as it can). Let  $x = x_i - y_i$ .

Consider the following new solution  $O'$  constructed from  $O$ :

For  $j < i$ , keep  $y_j' = y_j$ .

Set  $y_i' = x_i$ .

In  $O$ , remove items of total weight  $x * w_i$  from items  $i + 1$  to  $n$ , resetting the  $y_j'$  appropriately. The total value of solution  $O'$  is greater than or equal to the total value of solution  $O$ . Since  $O$  is largest possible solution and value of  $O'$  cannot be smaller than that of  $O$ ,  $O$  and  $O'$  must be equal. Thus solution  $O'$  is also optimal. By repeating this process, we will eventually convert  $O$  into  $G$ , without changing the total value of the selection. Therefore  $G$  is also optimal

### Problem 4:

Kruskal's Algorithm implemented.

Union Find Data structure used.

Time complexity for Union Find is  $O(\log n)$  ( $n$ = number of elements) and for  $m$  nodes ( $m$  operations) it's  $O(m \log n)$ .

$m \rightarrow$  edges;  $n \rightarrow$  nodes;

Time complexity for computeMST():  $O(m * \log n)$

Sort Edge List:  $O(m * \log n)$  from

$O(m * \log m)$  where  $O(\log m)$  belongs to  $O(\log n)$  since  $m \leq n^2$

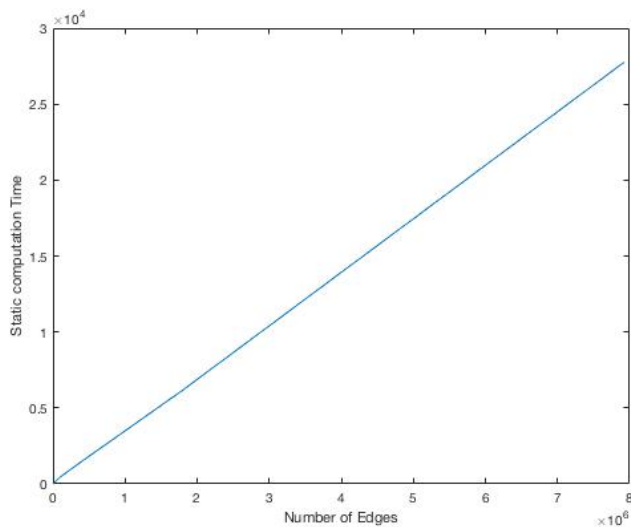
Union Find:  $O(n+m)$  from

$O(n + m * \log n)$  where  $\log n$  is  $\sim$ constant.

Time Complexity for recomputeMST():  $O(m * \log n)$

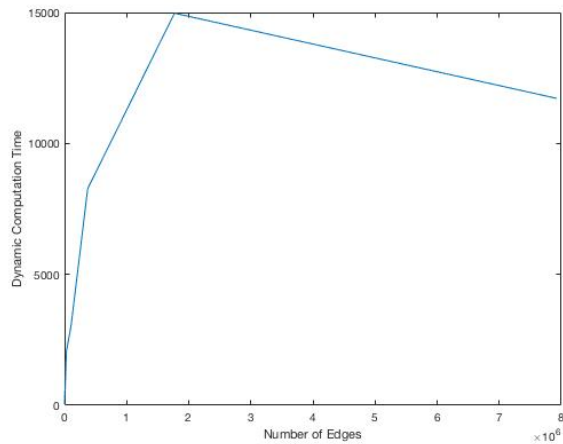
Shortest Path:  $O(m * \log n)$  (networkx implementation uses Dijkstra's)

Plots:



Static Computation

Log  $n$  close to constant therefore  $O(m * \log n) = O(m)$  = linear graph



### Dynamic computation

Slope of graph changes due to change in the ratio of initial edges to number of edges being added. Initially with a larger ratio the time complexity due to the union find is dominant ( $O(n+m \cdot \log n)$ ) but as the number of edges increases the shortest path computation ( $O(m \cdot \log n)$ ) requires a longer time.

### Reference:

Union Find Implementation: <http://www.ics.uci.edu/~eppstein/PADS/UnionFind.py>