

HOMEWORK 1

Problem 1

Soln:

let the given array is $A[1 \dots n]$
 and $K = \text{Order Statistic}$
 If we apply black-box subroutine on A ,
 it will return $n/2$ element. If $K=n/2$, it
 will return $n/2$ th element of the array,
 i.e., $A[n/2]$, else we divide A into
 two groups; one group will have elements
 less than $A[n/2]$ named as A_1 , and
 other group will have elements greater than
 $A[n/2]$ named as A_2 .

If $K < n/2$, we will try to find
 the order statistic for the K^{th} element
 in A_1 .

If $K > n/2$, we will fetch the order
 statistic for the $(\frac{n}{2} - k)^{\text{th}}$ element in A_2 .

Below are the pseudocode using procedures:-

SELECTION (A, K)

BLACK-BOX(A)

IF $K = n/2$

then return $A[n/2]$

DIVIDE (A)

IF $K < n/2$

SELECTION (A₁, K)

ELSE SELECTION (A₂, n/2-K)

END SELECTION

Using black-box subroutine, the cost in computing the median = $O(n)$

Cost in dividing the array = $O(n)$

Suppose $R(n)$ will be the cost of computing k^{th} order statistic then

Total running time $R(n) \leq cn + R(n/2)$

$$R(n) \leq cn + R(n/2)$$

$$\leq c\left(n + \frac{n}{2} + R(\frac{n}{4})\right)$$

$$\leq c\left(n + \frac{n}{2} + \frac{n}{4} + R(\frac{n}{8})\right)$$

$$\leq C\left(n + \frac{n}{2} + \frac{n}{4} + \frac{n}{8} + \dots - R(1)\right)$$

$$\leq cn$$

$$= O(n)$$

Problem 2:

(a)

Solution: Suppose x_k is the median of n distinct elements x_1, x_2, \dots, x_n .

According to definition given, if x_k is greater than $\lfloor \frac{n+1}{2} \rfloor - 1$ other elements x_i then, sum of weights of elements less than x_k will be :

$$\sum_{x_i < x_k} w_i = \frac{1}{n} \cdot \left(\left\lfloor \frac{n+1}{2} \right\rfloor - 1 \right)$$

$$= \frac{1}{n} \cdot \left\lfloor \frac{n+1-2}{2} \right\rfloor = \frac{1}{n} \cdot \left\lfloor \frac{n-1}{2} \right\rfloor$$

$$\leq \frac{n-1}{2n}$$

$$< \frac{n}{2n}$$

$$< \frac{1}{2}$$

If x_k is lesser than exactly $n - \left\lfloor \frac{n+1}{2} \right\rfloor$ other elements then

$$\sum_{x_i > x_k} w_i = \frac{1}{n} \cdot \left(n - \left\lfloor \frac{n+1}{2} \right\rfloor \right)$$

$$= 1 - \frac{1}{n} \cdot \left\lfloor \frac{n+1}{2} \right\rfloor$$

$$\leq 1 - \left(\frac{1}{n} \right) \left(\frac{n}{2} \right)$$

$$\leq \frac{1}{2}$$

So, we can see that x_k is also the weighted median.

(b).

Soln: To compute the weighted median of n elements, we will first sort the elements, then sum up the weights of elements till the median found.

Below are the pseudo code for this:

WEIGHTED-MEDIAN(A)

$K \leftarrow 1$

$S \leftarrow 0$

~~$S =$~~ Total weight of

all $x_i < x_k$

while $S + w_k < \frac{1}{2}$

do $S \leftarrow S + w_k$

$k \leftarrow k + 1$

return x_k

S is the sum of weights of all elements less than x_k then

$$S = \sum_{x_i < x_k} w_i$$

By Induction, we can prove that this is true.

In the first iteration, $S=0$, therefore the best case is true, since the list is sorted $\forall i < k$, $x_i < x_k$.

Since in every iteration through the loop
 S increases by the weight of next element
 therefore by Induction, S is correct.

Since the sum of weights of all elements
 is 1, the loop is guaranteed to terminate.
 Now, we can prove that x_k is the
 weighted median using the definition
 of weighted median, when the loop ends.

Suppose $S' = \text{value of } S \text{ at the start of}$
 the next to last iteration of
 the loop

$$\therefore S = S' + w_{k-1}$$

Since next to last iteration is not following
 the termination condition,

$$\therefore S' + w_{k-1} < \frac{1}{2}$$

$$\sum_{x_i < x_k} w_i = S < \frac{1}{2}$$

It is still true if the loop has zero iterations
 since $S=0$ which is less than $\frac{1}{2}$.

∴ This proves the first condition for being
 a weighted median.

Now for the second condition, the
 sum of weights of elements greater than

x_k is :

$$\sum_{x_i > x_k} w_i = 1 - \left(\sum_{x_i < x_k} w_i \right) - w_k \\ = 1 - s - w_k$$

from the loop condition,

$$\sum_{x_i < x_k} w_i = s \geq \frac{1}{2} - w_k$$

$$-s \leq -\frac{1}{2} + w_k$$

$$1 - s - w_k \leq \frac{1}{2}$$

$$\sum_{x_i > x_k} w_i \leq \frac{1}{2}$$

Therefore x_k also satisfies the second condition for being the weighted median.

i.e. x_k is the median and the algorithm is correct.

Now, the running time for this algorithm
= time to sort the array + time to find the median

$$= O(n \lg n) + O(1) \rightarrow O(n) \text{ iterations}$$

$$= O(n \lg n). \quad \text{require } O(1) \text{ time per iteration}$$

(c)

Soln:

The weighted median can be computed in $\Theta(n)$ worst-case time using a linear-time median algorithm. Like binary search, it will first compute the median and recurses on half of the input that contains the weighted median.

Below is the pseudo code for this:

```
1 LINEAR-TIME-WEIGHTED-MEDIAN (A, l)
2   n ← length[A]
3   m ← MEDIAN[A]
4   B ← φ           where  $B = \{A[i] < m\}$ 
5   C ← φ           where  $C = \{A[i] \geq m\}$ 
6    $w_B \leftarrow 0$     where  $w_B = \text{total weight of } B$ 
7   If  $\text{length}[A] = 1$ 
8     then return  $A[1]$ 
9   for  $i \leftarrow 1$  to  $n$ 
10    do if  $A[i] < m$ 
11      then  $w_B \leftarrow w_B + w_i$ 
12      Append  $A[i]$  to array  $B$ 
13    else Append  $A[i]$  to array  $C$ 
14.   If  $l + w_B > \frac{1}{2}$            where weighted median ∈
15     then LINEAR-TIME-WEIGHTED-MEDIAN (B, l)
16     else LINEAR-TIME-WEIGHTED-MEDIAN (C,  $w_B$ )
```

LINEAR-TIME WEIGHTED-MEDIAN(A, D) is the initial initial call. An array A contains median of initial input and l is the total weight of ~~the~~ elements of initial input that are less than all the elements of A. B will have all the elements ~~is~~ less than the median. C contains all elements greater than or equal to median, w_B is the total weight of elements in B.

The weighted median y of the initial A is always present in the recursive call of A and l is the total weight of all elements \leq less than all the elements of A. For ~~initial~~ initial call, this precondition is ~~is~~ partially true. By Induction, we can prove that the precondition is true for every recursive call.

First assume the case where $l + w_B > \frac{1}{2}$. Since y must be in A, y must be either in B or C (line 14). Since total weight of all elements less than any element in C is greater than $\frac{1}{2}$, then the weighted median cannot be in C, it must be in B. Also, we have ~~as~~ not discarded any element less than any element in B, so l is correct if precondition is satisfied.

Now if $l + w_B \leq \frac{1}{2}$ then γ must be in C. All elements of C are greater than all elements of B, so total weight of elements less than the elements of C is $l + w_B$ and the precondition is also satisfied.

∴ By Induction, the precondition is always true.

The size of A decreases for every recursive call, the algorithm will end. When the algorithm ends, the result is correct.

∴ the weighted median is always in A,
 \therefore in case of only one element leftover, that will become the weighted median.

Time for computing the median and dividing A into B and C = ~~$O(n)$~~ $\Theta(n)$

Each recursive call reduces the size of array from n to $\lceil n/2 \rceil$.

$$\begin{aligned}\therefore T(n) &= T(n/2) + \cancel{\Theta(n)} \Theta(n) \\ &= \Theta(n)\end{aligned}$$

(d)

Solⁿ:

$$C(P) = \sum_{i=1}^n w_i d(P, p_i)$$

Rewriting $C(P)$ as the sum of the cost contributed by points less than p and points

greater than P :

$$C(P) = \left(\sum_{p_i < P} w_i (P - p_i) \right) + \left(\sum_{p_i > P} w_i (p_i - P) \right)$$

If $P = p_k$ for some k , then the point does not contribute to cost. Because of

$\lim_{P \rightarrow \infty} C(P) = c(x) + n$, this function is continuous.

If we take derivative w.r.t. P :

$$\frac{dc}{dp} = \left(\sum_{p_i < P} w_i \right) - \left(\sum_{p_i > P} w_i \right)$$

for $P = p_i$ for some i , this derivative is undefined. Also, $\frac{dc}{dp}$ is non-decreasing

function, since as P increases, the no. of points $p_i < P$ cannot decrease.

for $P < \min(p_1, p_2, \dots, p_n)$, $\frac{dc}{dp} < 0$

for $P > \max(p_1, p_2, \dots, p_n)$, $\frac{dc}{dp} > 0$

\therefore there will be some point P^* where

$\frac{dc}{dp} \leq 0$ for $P < P^*$ and

$\frac{dc}{dp} \geq 0$ for $P > P^*$ and

this point is a global minimum.

for all $p < y$ where p is not the weighted median and $p \neq p_i$ for some i ,

$$\sum_{p_i < p} w_i < \sum_{p_i > p} w_i$$

$$\Rightarrow \frac{dc}{dp} < 0$$

Likewise, for $p > y$, where p is not the weighted median and $p \neq p_i$ for some i ,

$$\sum_{p_i < p} w_i > \sum_{p_i > p} w_i$$

$$\Rightarrow \frac{dc}{dp} > 0$$

For $p = p_i$ where for some i and $p \neq y$,
where $p = p_i$, both left and right
limit of $\frac{dc}{dp}$ will have the same sign.

i. $c(p) \geq c(y)$ for all p that are
not the weighted median.

So the weighted median y is a
global median.

(e)

Solⁿ:

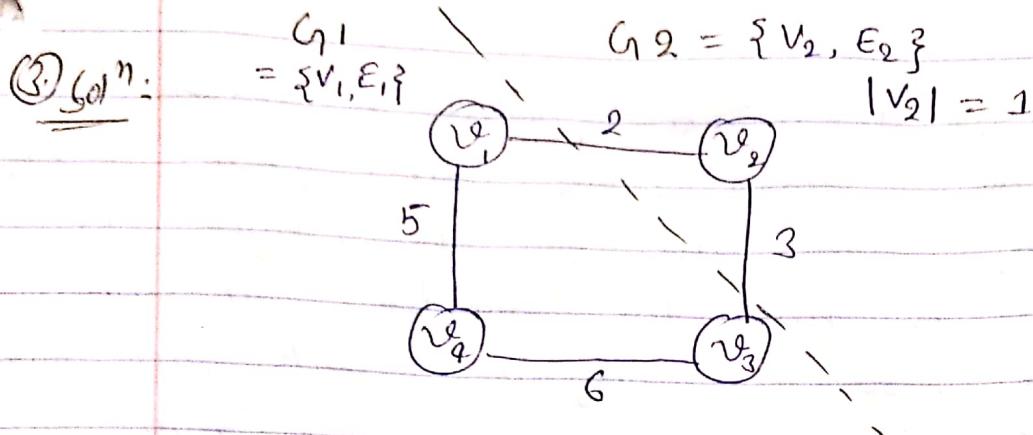
let the solⁿ be $p = (p_x, p_y)$

Writing the cost function as sum of two one-dimensional post office location cost functions :-

$$g(p) = \left(\sum_{i=1}^n w_i |x_i - p_x| \right) + \left(\sum_{i=1}^n w_i |y_i - \cancel{p_y}| \right)$$

$\frac{\partial g}{\partial p_x}$ does not depend on y-coordinates of the input points and has the same form as $\frac{dc}{dp}$ from the previous part using only the x coordinates as input. Likewise, $\frac{\partial g}{\partial p_y}$ depends only on the y-coordinates.

∴ To minimize $g(p)$, we will minimize the cost for the two dimensions independently. Let p_x be the solⁿ to the one-dimensional post-office location problem for inputs $x_1, x_2, x_3, \dots, x_n$ and p_y be the solⁿ to the one-dimensional post-office location problem for inputs $y_1, y_2, y_3, \dots, y_n$.



Consider the above graph of mst.
 In above figure, graph $G = (V, E)$ has
 4 vertices and we divided it into
 two graphs :

G_1 with $V_1 = \{v_1, v_4, v_3\}$ & E_1

& G_2 with $V_2 = \{v_2\}$

Graph G_1 contains $\{v_1, v_4, v_3\}$

Graph G_2 contains $\{v_2\}$ since $|V_2| = 1$

1. The minimum-weight edge crossing the cut (V_1, V_2) has weight 2. The Spanning tree forming by the proposed algorithm is :

$v_1 - v_2 - v_3 - v_4$ has weight 11.

$v_2 - v_3 - v_4 - v_1$ has weight 14.

$v_3 - v_4 - v_1 - v_2$ has weight 13.

$v_4 - v_1 - v_2 - v_3$ has weight 10.

Now, if we look into the above spanning

this point is a global minimum.
 For all trees, the first path $v_1 - v_2 - v_3 - v_4$ will give you the weight 11 which is not the ~~shortest~~ smallest weight, so the algorithm will fail here in this case. Here, the algorithm traverses through each edge in a way that it forms acyclic and includes all the vertices. Out of four ~~paths~~ paths, ~~not~~ mentioned above, only one path will be the MST which has smallest weight. In this example, path $v_4 - v_1 - v_2 - v_3$ will be the minimum Spanning tree since it has the smallest weight. The algorithm will fail to produce a minimum spanning tree, if the partition is done in worst-case manner ie, when we choose graph with highest weight ie,

$$v_2 - v_3 - v_4 - v_1 = 14$$

Q.

Ans: If the ~~partition~~ partition is always done in best-case manner, the algorithm will always produce a Minimum Spanning tree. In the above example, the graph path which will have the lowest

(Q)

or smallest weight will consider as in best-case manner and in this case, the path:

$v_4 - v_1 - v_2 - v_3$ having weight 10 will produce the minimum spanning tree. However the percentage of finding MST is really low which is 25%. For the graph proposed. Hence, it gives really a low scope of finding the correct graph.

(4.)

Solⁿ: Let $G = (V, E, c)$ be a weighted undirected graph where all the costs c_e for $e \in E$

let T be a minimum spanning tree in

$$G = (V, E, c).$$

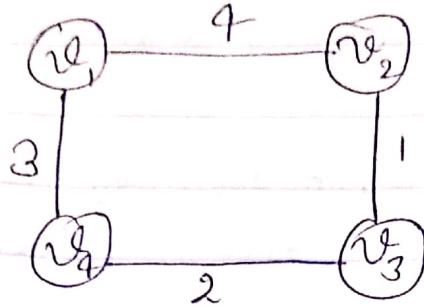
Now let replace the cost of each edge $e \in E$ by $c'_e = c_e^2$ for graph $G' = (V, E, c')$

of graph G

So, if T is the MST of G then T will also be the MST of graph G' .

If we have graph G with n vertices & $(n-1)$ edges, then incrementing the cost by square of edges by square will increase the cost of every spanning tree in the same order of square.

example:



For the above graph, the MST
~~with~~ will be $v_2 \rightarrow v_3 \rightarrow v_4 \rightarrow v_1$,

The edges covered will be $\{v_2, v_3\}, \{v_3, v_4\}, \{v_4, v_1\}$ having cost $= 1 + 2 + 3 = 6$.

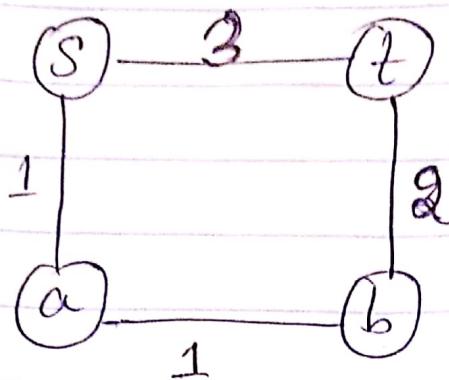
Now, when we square the cost of edges, the cost will change to

$$1^2 + 2^2 + 3^2 = 1 + 4 + 9 = 14$$

So, ~~the~~ if we find the MST of a graph G , it will remain the MST of the graph G' even if we square the edges ~~in~~ in G' . (cost of MST of G will always be smaller if we increase the cost of each edge by square).

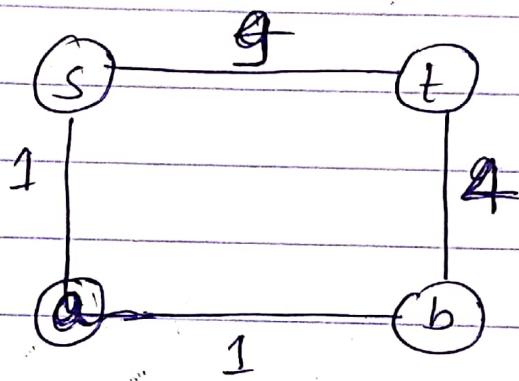
Part II →

If P is the shortest ~~path~~ $s-t$ path in G , it is not necessary that it will be the shortest path in G' . Let's consider the below graph:



Here, we can easily see that the shortest path will be $s-t$ with cost 3.

But, ~~as~~ when we square the cost of edges, we will get the below graph



Now, the shortest path will change if it will become $s-a-b-t$ having cost 6.

Hence, it is proved that the shortest path will change in ~~G~~ G' after squaring the edges in G .