

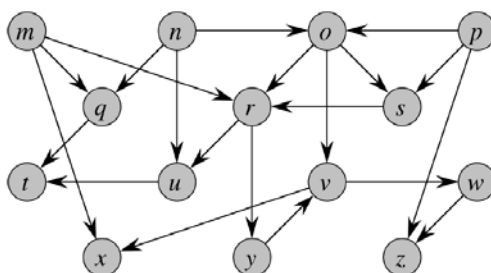
## EL9343 Homework 5

(Due August 12<sup>th</sup>, 2020)

No late assignments accepted

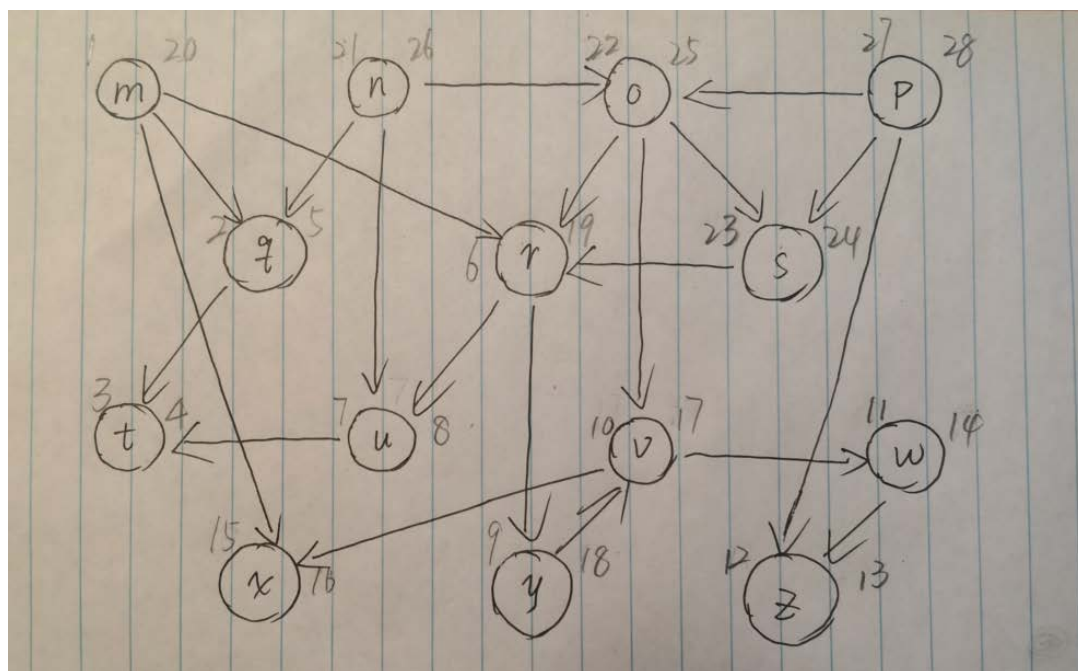
*All problem/exercise numbers are for the third edition of CLRS text book*

1. Show the ordering of vertices produced by TOPOLOGICAL-SORT when it is run on the DAG below. Assume that **for** loop of lines 5—7 of the DFS procedure (page 604 in CLRS) considers the vertices in alphabetical order, and assume the adjacency list is ordered alphabetically.



Solution:

This is the graph after we run DFS on the graph:



The left number is the discover time and the right number is the finish time.

According to the TOPOLOGICAL-SORT( $G$ ), we can get the ordering of the vertices should be:

p, n, o, s, m, r, y, v, x, w, z, u, q, t.

2. Run the procedure STRONGLY-CONNECTED-COMPONENTS on the graph below. Show the:

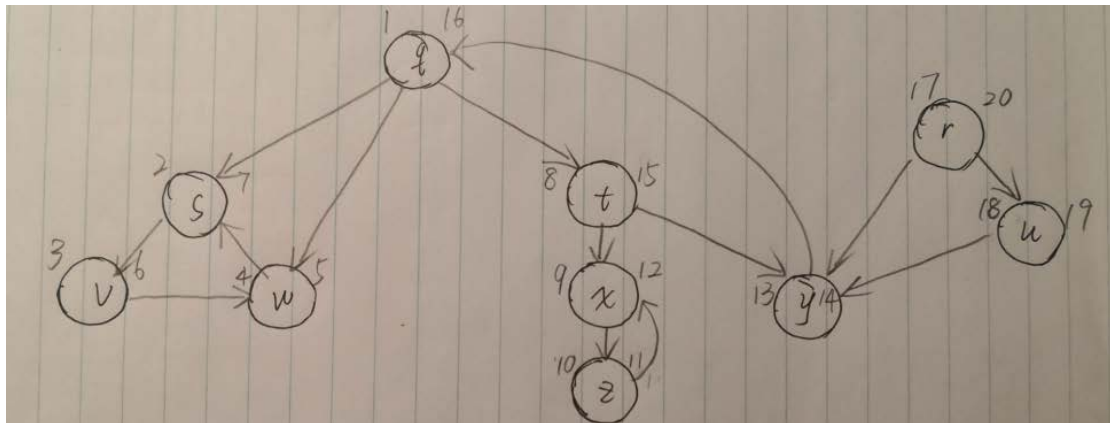
- (a) The finishing times for each node after running DFS in line 1
- (b) The DFS forest produced by line 3
- (c) The nodes of each tree in the DFS forest produced in line 3 as a separate strongly connected component.

Assume that **for** loop of lines 5—7 of the DFS procedure (page 604 in CLRS) considers the vertices in alphabetical order, and assume the adjacency list is ordered alphabetically.

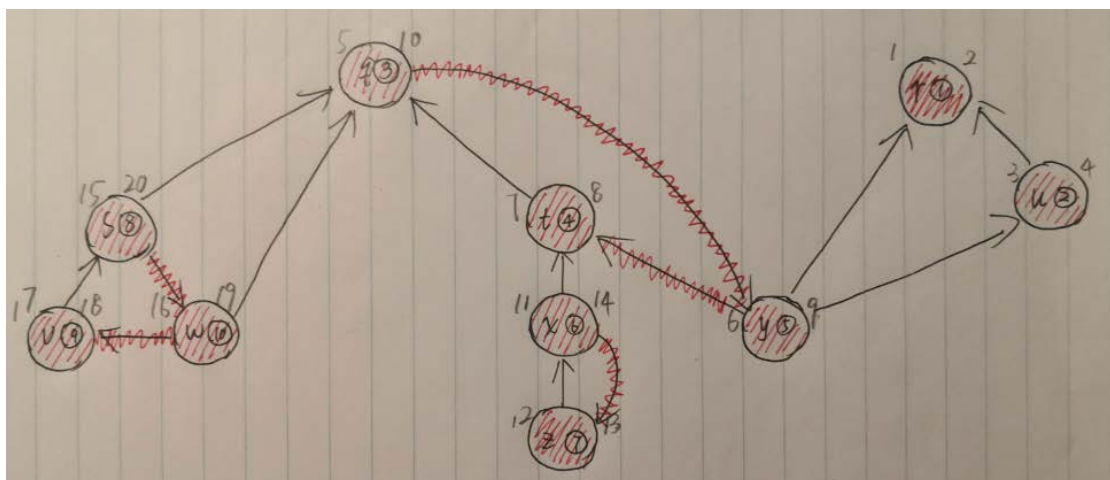
Solution:

(a)

As shown below, the discover time and finish time lie on two sides of each node, the left is the discover time, the right is the finish time.



(b)

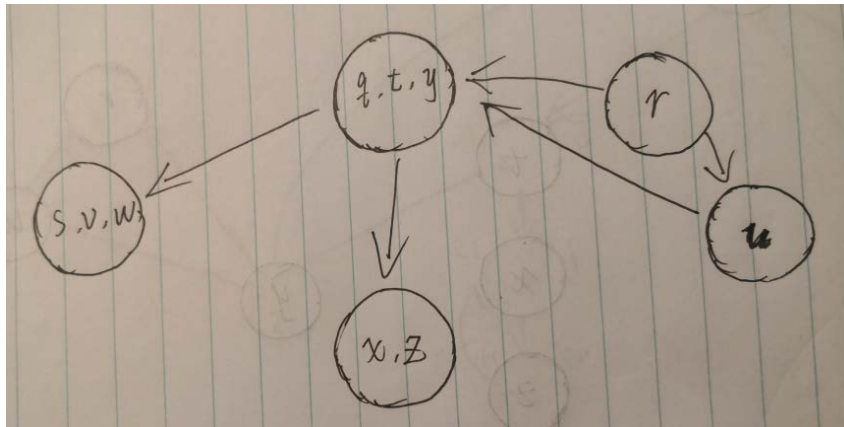


The result of DFS on  $G^T$  is shown as above. The number inside each node is the order computed

from (a). Discover time and finish time lie on two sides of each node, the left is the discover time, the right is the finish time. Every DFS tree is marked by red pen. As we can see, there are 5 distinct trees. Which means the  $G$  has 5 strongly connected component.

(c)

Converge the nodes in one tree in (b) to one node and link them:



3. The holidays are here! You decide to travel to see your friends in California.

Since you are on a budget, so you decide to drive to visit them. You choose your route so you pass by the most spectacular vistas on your route. Along your route are  $n$  hotels, at miles  $m_1 < m_2 < \dots < m_n$  from your starting location. The cost of the  $i^{\text{th}}$  hotel is  $h_i$ . Your friends live at mile marker  $m_{n+1}$  where  $m_n < m_{n+1}$ . The starting location is  $m_0$ .

You won't drive more than 400 miles per day for safety concerns. (The hotels on your route are all within 400 miles of each other)

You want to minimize the cost of your travel.

Provide a recurrence formula that would calculate the answer.

Using your recurrence formula, design an algorithm to find the minimum cost.

Solution:

Let  $c[0..n]$  is an array.  $c[i]$  denotes that the maximal total resource consumption when the last rest points of the whole venture is the  $i^{\text{th}}$  rest point. Once he arrives this point, his total resource consumption is determined. So,  $c[i]$  only depends on those rest points are before the  $i^{\text{th}}$  point, but the distance of every one of them must be larger than or equal to  $p$  miles to the  $i^{\text{th}}$  point. We can have the following formula: Obviously,  $c[0] = 0$ . For every  $0 < i \leq n$ , we have:

$$c[i] = \max_{0 \leq j < i} (c[j] + r_i) \text{ When } d_i - d_j \geq p.$$

After we compute every  $c[i]$ , we need to compare all of them and pick the largest one.

So, we can get our algorithm directly from the formula above.

MAX\_CONSUM( $n, d[0..n], r[1..n]$ )

Let  $c[0..n]$  be new array

$c[0] = 0$

for  $i$  from 1 to  $n$

temp =  $-\infty$

for  $j$  from 0 to  $i - 1$  // We want to find the first one that is less than  $p$  miles away from

$i^{\text{th}}$  point and then no need to find those points which is behind that point.

if  $d[i] - d[j] \geq p$  // larger than  $p$  miles

if  $c[j] + r[i] > \text{temp}$

temp =  $c[j] + r[i]$

else // within  $p$  miles, we do not need to consider these points

break

max\_consum = 0

for  $i$  from 1 to  $n$

if  $c[i] > \text{max\_consum}$

max\_consum =  $c[i]$

return max\_consum

Because we have an outer loop and an inner loop, and all operations in a loop is in  $O(1)$  time.

Running time for it is  $O(n^2)$ . After we compute every element in  $c[0..n]$ , we need a loop to pick the largest one, which is in  $O(n)$  time. Totally we will need  $O(n^2)$  time to complete this task.

4. A critical water treatment plant is composed of  $n$  subsystems. Each subsystem's chance for failure is independent of all the other subsystems, therefore the probability that the plant functions correctly on a given day ( $r_1 \cdot r_2 \cdot \dots \cdot r_n$ ). Given that the plant is critical for water distribution in the state, each subsystem has a number of backup subsystems, denoted as  $b_i - 1$ . So, you can see that the probability of the subsystem failing and not being able to be replaced

is  $(1 - r_i)^{b_i}$ . Therefore the overall chance of failure is  $1 - \prod_{i=1}^n (1 - (1 - r_i)^{b_i})$ .

Given a budget  $B$ , subsystem costs  $c_1, \dots, c_n$  and a new plant to build (in a different state), that is the optimal number of backup  $(b_1 \dots b_n)$  subsystems for your new plant.

Instead of providing an algorithm:

- Provide the recurrence formula
- Prove that the problem has optimal substructure

Solution:

We want the value of overall chance of the failure,  $1 - \prod_{i=1}^n (1 - (1 - r_i)^{b_i})$ , to be as small as possible. Which is equivalent to make  $\prod_{i=1}^n (1 - (1 - r_i)^{b_i})$  as large as possible. So, this question is to find the optimal combination of the number of backup subsystems for every subsystem, and make their cost is less or equal to  $B$ . Let  $dp[0..B]$  is an array,  $dp[i]$  denote that the maximal value of  $\prod_{i=1}^n (1 - (1 - r_i)^{b_i})$  when our budget is  $i$ . Then we can get the following recurrence formula:

$$dp[i] = \max_{1 \leq k \leq n} (dp[i - c_k] \frac{1 - (1 - r_k)^{b_k + 1}}{1 - (1 - r_k)^{b_k}}) \quad \text{When } i - c_k \geq 0$$

In order to compute this value, we need to store the selection method of every budget  $i$ . That is, for every  $dp[i]$ , we will have a table  $T_i[1..n]$  to store the number of every subsystem. That's where we can get our  $b_k$  on the formula above ( $b_k = T_i[k]$ ). Obviously, for  $i = 0$ ,  $T_0[1..n] = [0, 0, \dots, 0]$  and  $dp[0] = \prod_{i=1}^n (1 - (1 - r_i)^0) = 0$ . So, we have initialized our base problem. We can then compute  $dp[1], dp[2], \dots, dp[B]$ . Of course, we must also keep track of the value of  $T_i[1..n]$ . After we compute the  $dp[B]$ . The  $T_B[1..n]$  will record the number of every subsystem that will get a maximal value of  $\prod_{i=1}^n (1 - (1 - r_i)^{b_i})$  (the probability of success). Total running time is  $O(Bn^2)$ , because for every  $i$ , we need  $O(n^2)$  time to compute the optimal combination and the optimal value  $dp[i]$ .

5. Design a greedy algorithm for making change consisting of quarters, dimes, nickels, and pennies. Prove your algorithm has the greedy choice property and optimal substructure.

Justify the running time of your algorithm.

Solution:

Let  $a$  is the number of quarters,  $b$  is the number of dimes,  $c$  is the number of nickels,  $d$  is the

number of pennies. This problem is equivalent to given  $n$ , make  $a + b + c + d$  is smallest such that  $25a + 10b + 5c + d = n$ . We can simply do this:

(a) Let  $n \bmod 25 = n_1$ , the remainder is  $n_1$ , which means  $n = 25a_0 + n_1$ , where  $0 \leq n_1 < 25$ .

(b) Let  $n_1 \bmod 10 = n_2$ , the remainder is  $n_2$ , which means  $n_1 = 10b_0 + n_2$ , where  $0 \leq n_2 < 10$

(c) Let  $n_2 \bmod 5 = n_3$ , the remainder is  $n_3$ , which means  $n_2 = 5c_0 + n_3$ , where  $0 \leq n_3 < 5$

(d) Let  $d_0 = n_3$ .

Pseudocode:

GREEDY\_CHANGE( $n$ )

```

    n1 = n % 25
    a0 = (n - n1) / 25
    n2 = n1 % 10
    b0 = (n1 - n2) / 10
    n3 = n2 % 5
    c0 = (n2 - n3) / 5
    d0 = n3
    return a0, b0, c0, d0

```

The  $a_0, b_0, c_0, d_0$  will be the value that satisfies the requirement of this problem, now we prove the correctness of this greedy algorithm:

(1) Optimal Substructure:

When we want to find  $a, b, c, d$  that satisfies  $25a + 10b + 5c + d = n$ , we try every possible combination of  $n' + n'' = n$ . If we find the optimal solution for  $n'$  and  $n''$ , we can find the optimal solution for  $n$ . In our greedy algorithm, every step we make our  $n' = tk_0$ , where  $t$  is 25, 10, 5 or 1 and  $k_0$  is  $a_0, b_0, c_0$  or  $d_0$  we discussed above.  $n'' = n - n'$ , in every step,  $n = n, n_1, n_2$  or  $n_3$ . As we can see, the optimal solution for sum up to  $n$  is the combination of the optimal solution for sum up to  $n'$  and  $n''$ . So, this problem has optimal substructure property.

(2) Greedy Choice Property:

We prove this from bottom to up:

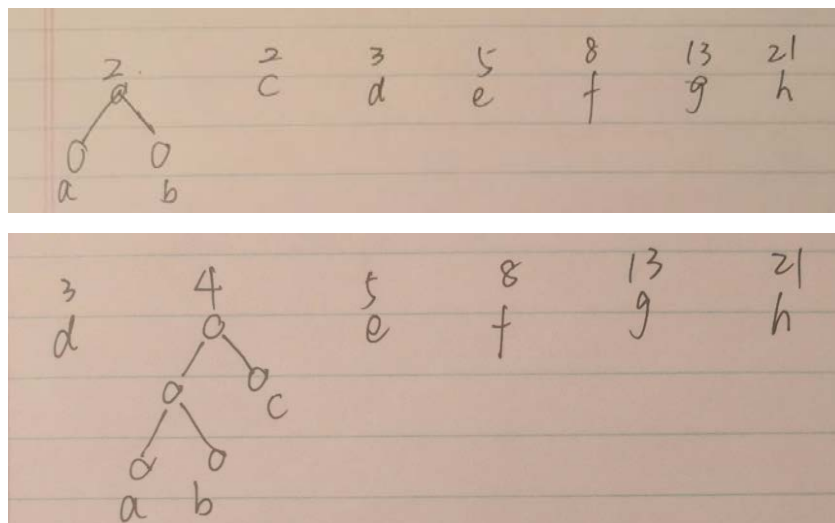
- i. For any  $n_2 = 5c + d$ , the optimal solution must be to make  $c = c_0$ . If not, which means  $c < c_0$ , WLOG we can assume  $c = c_0 - 1$ , now,  $d = d_0 + 5$ . So,  $d + c = d_0 + 5 + c_0 - 1 = d_0 + c_0 + 4 > c_0 + d_0$ . So, in the part (c) above, problem satisfies greedy choice property.
- ii. For any  $n_1 = 10b + n_2$ , the optimal solution must be to make  $b = b_0$ . If not, which means  $b < b_0$ , WLOG we can assume  $b = b_0 - 1$ , now,  $c = c_0 + 2$ ,  $d = d_0$ . So,  $b + d + c = b_0 - 1 + c_0 + 2 + d_0 = b_0 + c_0 + d_0 + 1 > b_0 + c_0 + d_0$ . So, in the part (b) above, problem satisfies greedy choice property.
- iii. For any  $n = 25a + n_1$ , the optimal solution must be to make  $a = a_0$ . If not, which means  $a < a_0$ , WLOG we can assume  $a = a_0 - 1$ , now,  $b = b_0 + 2$ ,  $c = c_0 + 1$ ,  $d = d_0$ . So,  $a + b + d + c = a_0 - 1 + b_0 + 2 + c_0 + 1 + d_0 = a_0 + b_0 + c_0 + d_0 + 2 > a_0 + b_0 + c_0 + d_0$ . So, in the part (a) above, problem satisfies greedy choice property.

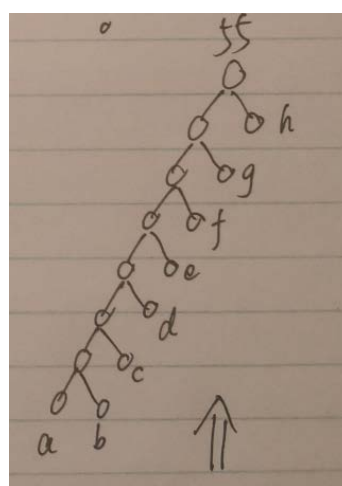
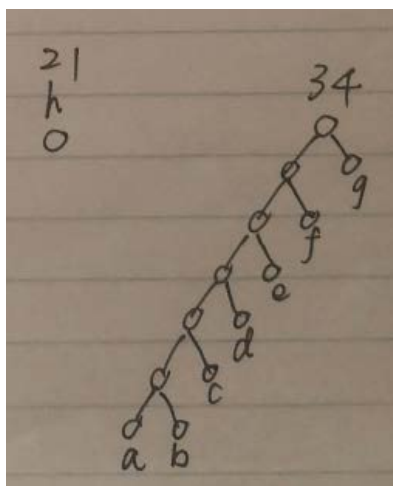
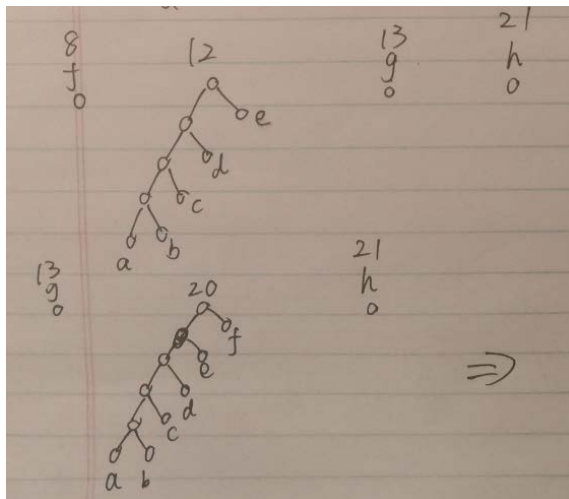
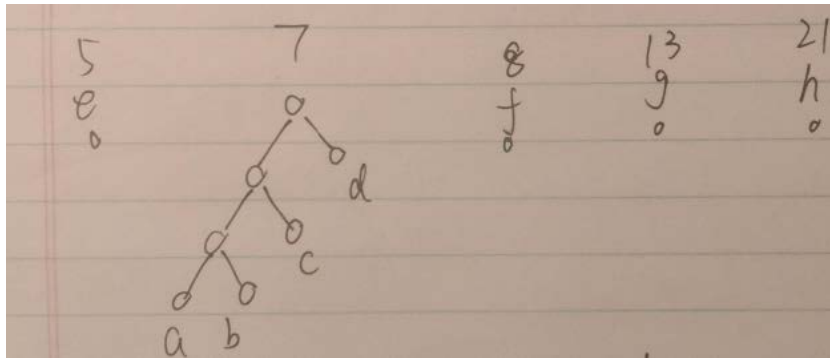
So, we have proved greedy choice property and optimal substructure for this problem, every step is a safe greedy choice, so our algorithm will work. The running time will  $O(1)$  because we only need 4 steps to find  $a$ ,  $b$ ,  $c$  and  $d$ .

6. 16.3-3 in CLRS on page 436.

Solution:

We follow the following steps to construct our Huffman encoding tree:





If left is 0, right is 1, we can get the optimal code:

a	0000000
b	0000001
c	000001
d	00001
e	0001
f	001



$g$	01
$h$	1

7. A native Australian named Alice wishes to cross a desert carrying only a single water bottle. She has a map that marks all the watering holes along the way. Assuming she can walk  $k$  miles on one bottle of water, design an efficient algorithm for determining where Alice should refill her bottle in order to make as few stops as possible. Argue why your algorithm is correct. Justify the running time of your algorithm.

Solution:

We just decide to refill our bottle at the hole that is the last one within the  $k$  miles of the hole we refill last time, and so force we will get to the destination. This refill method will give the solution with fewest stops. Let  $S[0..n+1]$  be an array,  $S[i]$  ( $i = 1, 2, \dots, n$ ) records the distance of the  $i^{\text{th}}$  watering hole from the start point (the 1<sup>st</sup> watering hole).  $S[i] = 0$ , represent the distance of the start point,  $S[n+1]$  is the distance of the destination. We can get the following pseudocode:

FEWEST\_STOP( $S[0..n+1]$ )

```

num_refill = 0
current_position = 0
last_refill_position = 0
while current_position < n + 1
    last_refill_position = current_position
    while (current_position < n + 1 and
           S[current_position + 1] - S[last_refill_position] <= k)
        current_position = current_position + 1
    if current_position < n + 1
        num_refill += 1
return num_refill

```

The running time of this algorithm is  $O(n)$ , because  $current\_position$  is increasing all the time and changes 1 by 1, and outer loop restrict it only can reach  $n + 1$ . For every  $current\_position$ , our

operation can be done in  $O(1)$  time. So, the total running time is  $(n + 1)O(1) = O(n)$ . Now we prove the correctness of this algorithm:

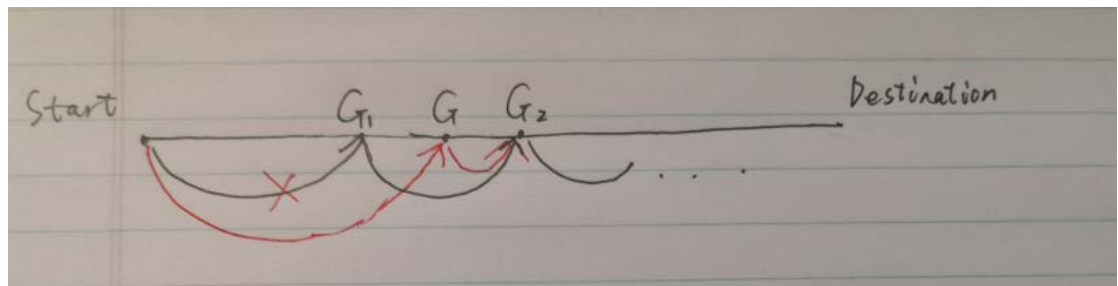
(1) Optimal Substructure:

Every time we arrive at some water hole  $i$ , the minimal number of refill time is the smallest one among all possible last stops: The last stop can be at 1<sup>st</sup> hole, 2<sup>nd</sup> hole, ...,  $(i-1)^{\text{th}}$  hole. So, this problem has optimal substructure.

(2) Greedy Choice Property:

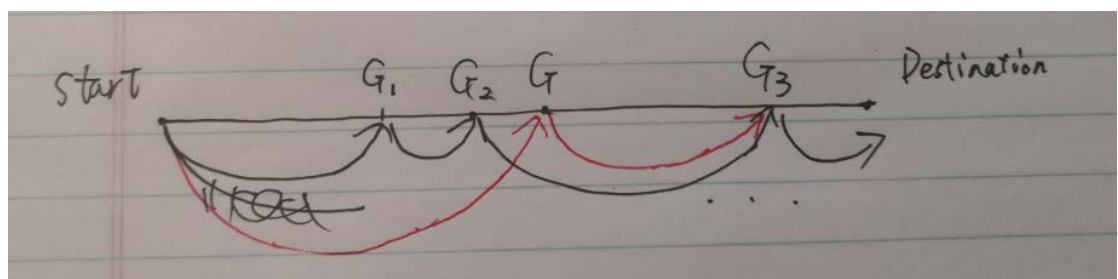
We can prove that our choice of refilling in every step will make up of the optimal solution. Let  $G$  be the hole that is the last one within  $k$  miles from our start point. If we have an optimal solution that has the fewest stops along the way and the stops are  $G_1, G_2, \dots, G_T$  (Assume that the fewest stops is  $T$ ), and  $G_1 \neq G$ , which means  $G_1$  is before  $G$ , we will have the following cases:

(a)  $G_2$  is behind  $G$



As we can see, the path consist with black directed edges is our optimal route, we can replace the first 2 black edges with 2 red edges in the picture above, the 2 red edges and the following black edges will also be an optimal route of our problem.

(b)  $G_2$  is before  $G$



As we can see, the path consist with black directed edges is our optimal route. If we replace the first 3 black edges with 2 red edges, we can get a path that has  $T - 1$  stops, it contradicts with the hypothesis that our stops are  $G_1, G_2, \dots, G_T$  is an optimal solution.

This case is impossible.

So, we only have case (a), we have proved that every time choosing the hole that is the last one within  $k$  miles of our last refill hole as our refill hole is safe. Which means this problem has greedy choice property, the algorithm above will work.

8. Suppose you've been sent back in time and have arrived at the scene of an ancient Roman battle. Moreover, suppose you have just learned that it is your job to assign  $n$  spears to  $n$  Roman soldiers, so that each man has a spear. You observe that the men and spears are of various heights, and you have been told that the army is at its best if you can minimize the total difference in heights between each man and his spear. That is, if the  $i^{\text{th}}$  man has height  $m_i$  and his spear has height  $s_i$ , then you want to minimize the

$$\sum_i^n |m_i - s_i|$$

Consider a greedy strategy of repeatedly matching the man and the spear that minimizes the difference in heights between these two. Prove or disprove that this greedy strategy results in the optimal assignment of spears to men.

Solution:

Let  $m[1..n]$  and  $s[1..n]$  be arrays,  $m[i]$  represents the height of  $i^{\text{th}}$  man and  $s[i]$  represents the height of the  $i^{\text{th}}$  spear. We firstly sort them and just assign  $s[i]$  to  $m[i]$ . Then the sum  $\sum_i^n |m[i] - s[i]|$  will be minimized. Pseudocode is shown as below:

OPT\_DIF\_SUM( $m[1..n]$ ,  $s[1..n]$ )

    Merge\_Sort( $m[1..n]$ )

    Merge\_Sort( $s[1..n]$ )

    sum = 0

    for  $i$  from 1 to  $n$

        sum += abs( $m[i] - s[i]$ ) //abs(a, b) will return the absolute value of  $a - b$

    return sum

Every merge sort will cost  $O(n \log n)$  time, and computing sum will cost  $O(n)$  time because it is a loop with  $i$  from 1 to  $n$ . The total running time will be  $O(n \log n) + O(n) = O(n \log n)$ .

Now, we prove the correctness of this algorithm:

(1) Optimal Substructure:

When we split  $m[1..n]$  and  $s[1..n]$  to 2 part,  $m[1..k]$ ,  $m[k+1..n]$  and  $s[1..k]$ ,  $s[k+1..n]$ , respectively ( $k = 1, 2, \dots, n$ ), the optimal assigning of  $m[1..n]$  to  $s[1..n]$  must be combined with the optimal assigning of  $m[1..k]$  to  $s[1..k]$  and the optimal assigning of  $m[k+1..n]$  to  $s[k+1..n]$ . So, this problem has optimal substructure.

(2) Greedy Choice Property:

If we have an optimal solution that  $s[n]$  is not assigned to  $m[n]$ , WLOG, assume that  $s[n]$  is assigned to  $m[u]$ , and there must be a  $s[v]$  is assigned to  $m[n]$ . Now, we have  $s[n] \geq s[v]$  and  $m[n] \geq m[u]$ , because we sort  $m[1..n]$  and  $s[1..n]$ . If we assign  $s[n]$  to  $m[n]$  and assign  $s[v]$  to  $m[u]$ . We can also get an optimal solution:

$|s[n] - m[u]| + |s[v] - m[n]| \geq |s[n] - m[n]| + |s[v] - m[u]|$ . In order to prove this, we

can change some notation to conveniently represent our variables: When  $a < b$ ,  $c < d$ ,

$|a - d| + |b - c| \geq |a - c| + |b - d|$ .

(a)  $a < b < c < d$

$$|a - d| + |b - c| = d - a + c - b = -a - b + c + d$$

$$|a - c| + |b - d| = c - a + d - b = -a - b + c + d$$

$$|a - d| + |b - c| \geq |a - c| + |b - d| \text{ holds.}$$

(b)  $c < d < a < b$

$$|a - d| + |b - c| = a - d + b - c = a + b - c - d$$

$$|a - c| + |b - d| = a - c + b - d = a + b - c - d$$

$$|a - d| + |b - c| \geq |a - c| + |b - d| \text{ holds.}$$

(c)  $a < c < d < b$

$$|a - d| + |b - c| = d - a + b - c = -a + b - c + d$$

$$|a - c| + |b - d| = c - a + b - d = -a + b + c - d < -a + b < -a + b - c + d$$

$$|a - d| + |b - c| \geq |a - c| + |b - d| \text{ holds.}$$

(d)  $c < a < b < d$

$$|a - d| + |b - c| = d - a + b - c = -a + b - c + d$$

$$|a - c| + |b - d| = a - c + d - b = a - b - c + d < -c + d < -a + b - c + d$$

$$|a - d| + |b - c| \geq |a - c| + |b - d| \text{ holds.}$$

$$(e) \ a < c < b < d$$

$$|a - d| + |b - c| = d - a + b - c = -a + b - c + d$$

$$|a - c| + |b - d| = c - a + d - b = -a - b + c + d < -a + d < -a + b - c + d$$

$$|a - d| + |b - c| \geq |a - c| + |b - d| \text{ holds.}$$

$$(f) \ c < a < d < b$$

$$|a - d| + |b - c| = d - a + b - c = -a + b - c + d$$

$$|a - c| + |b - d| = a - c + b - d = a + b - c - d < b - c < -a + b - c + d$$

$$|a - d| + |b - c| \geq |a - c| + |b - d| \text{ holds.}$$

As we can see, in any case, if  $a < b$ ,  $c < d$ ,  $|a - c| + |b - d| \leq |a - d| + |b - c|$ . In other case,  $s[v] < s[n]$ ,  $m[u] < m[n]$ . So,  $|s[v] - m[u]| + |s[n] - m[n]| \leq |s[n] - m[u]| + |s[v] - m[n]|$ . We can see, if there is an optimal solution, we can assign  $s[n]$  to  $m[n]$  to make a solution that is also the optimal one. So, this problem has greedy choice property. It is safe for us to do the greedy algorithm above.