

# CSE 101 Homework 5

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## 1. Cart horses

Consider the following problem: We have  $n$  horses,  $Horse_1, \dots, Horse_n$ , each with a cart-pulling speed  $s_i$ . We need to pair the horses up into teams to pull carts. The faster horse will need to slow down for the slower horse, so if we pair  $Horse_i$  and  $Horse_j$ , the cart will be pulled at speed  $\min(s_i, s_j)$ . Each horse can only be in one pair. We want to maximize the sum of all the cart speeds.

Give a greedy strategy for this problem, and prove that it is correct using any of the proof templates from class.

- (a) Greedy Algorithm: When picking the next pair of horses, pick the 2 fastest horses from the remaining horses.
- (b) Proof of correctness (via exchange):

Let  $G$  be the first greedy choice. Let  $OS$  be an optimal solution

Find the 2 pairs with the 2 fastest horses,  $H_1, H_2$ :  $(H_1, H_i), (H_2, H_j)$ . These horses will be separate, otherwise  $OS$  already contains  $G$ . Exchange  $H_i$  and  $H_2$

Note:

$$\begin{aligned} H_1 &\geq H_2, H_2 \geq H_i, H_2 \geq H_j \\ \min(H_1, H_i) &= H_i, \min(H_2, H_j) = H_j \\ \implies \min(H_1, H_2) &\geq \min(H_1, H_i) \end{aligned}$$

For the second new pairing,  $(H_i, H_j)$  there are 2 possibilities

- i.  $H_i \geq H_j \implies \min(H_2, H_j) = \min(H_i, H_j) = H_j$ :
- ii.  $H_i < H_j$ :

$$\begin{aligned} &\min(H_1, H_i) + \min(H_2, H_j) - \min(H_1, H_2) - \min(H_i, H_j) \\ &= H_i + H_j - H_2 - H_i = H_j - H_2 \leq 0 \\ \implies &\text{New solution is faster} \end{aligned}$$

Thus, the net speed of the exchanged solution is always as good as the original and meets the constraints of the problem

Now, proving correctness via induction on number of horses. This proof assumes there will always be an even number of horses to make the pairings.

- i. Base Case:  $n = 0, 2$ : Trivially, the greedy solution (and all solutions) are optimal
- ii. Induction Step. Let  $OS$  be any solution for the group of horses  $I = \{H_1, \dots, H_k\}, k > 2$   
By the exchange argument above,  $\exists OS'$  such that  $|OS| \leq |OS'|$  and  $G \in OS$   
Let  $I'$  be the set of pairs of horses that have neither  $H_1$  or  $H_2$  in them.

$$\begin{aligned} OS' &= \{G\} \cup I' = \{(H_1, H_2), (H_i, H_j)\} \cup I' \\ \text{Because } ||I'|| &= k - 4 \leq k, \text{ by the induction hypothesis } |S(I')| \leq |GS(I')| \\ \implies |OS| &\leq |OS'| = |\{G\} \cup OS(I')| \leq |\{G\} \cup GS(I')| = |GS(I)| \end{aligned}$$

Thus, for every set of horses, the greedy solution is as good as the optimal solution.

(c) Time Analysis. This algorithm only needs an input of horses, sorted by their speed. Thus, the algorithm takes  $O(n \cdot \log(n))$

- $\log(n)$  to sort the horses
- $n$  to scan through the sorted list of horses and create the pairings

## 2. Sensor maximization

Suppose you are placing sensors on a one-dimensional road. You have identified  $n$  possible locations for sensors, at distances  $0 \leq d_1 \leq d_2 \leq \dots \leq d_n$  from the start of the road. Each sensor must be at most  $M$  distance from the previous one (so they can communicate reliably). The first one must be at 0 and the last at  $d_n$ . Given that, you want to minimize the number of sensors placed.

The following greedy algorithm, which places each sensor as far as possible from the previous one, will return a list  $d_{i_1} \leq d_{i_2} \leq \dots d_{i_k}$  of locations where sensors can be placed.

GreedySensorMin [ $d_1 \dots d_n, M$ ]

- (a) Initialize a list to (0).
- (b) Initialize  $I = 1$ ,  $PreviousSensor = 0$ .
- (c) While  $I \leq n$  do:
  - (d) While  $I \leq n$  and  $d_I < PreviousSensor + M$  do:  $I++$ .
  - (e) If  $I \leq n$  THEN append  $d_{I-1}$  to list;  $PreviousSensor = d_{I-1}$ ;  $I++$ .
  - (f) Append  $d_n$  to list.
  - (g) Return list

Questions:

- (a) What constraints do solutions  $d_{i_1}, \dots, d_{i_k}$  need to meet for this problem?
  - For a solution  $d_{i_1}, \dots, d_{i_k}$ , minimize  $k$
  - $d_{i_1} < d_{i_2} < \dots < d_{i_k}$
  - $d_{i_j} - d_{i_{j-1}} < M \implies d_{i_j} < d_{i_{j-1}} + M$
  - $d_{i_1} = 0, d_{i_k} = d_n$
- (b) What is the value of the objective function for a solution, of the form  $d_0, d_{i_1}, \dots, d_{i_k} = d_n$ , ?  
The value of the objective function of such a solution is  $k$ . The goal is to minimize  $k$
- (c) In using the “greedy stays ahead” proof technique to show that this is optimal, we would compare the greedy solution  $d_{g_1}, \dots, d_{g_k}$  to another solution,  $d_{j_1}, \dots, d_{j_{k'}}$ . In what sense is the greedy solution “staying ahead” of the other solution at each step  $1 \leq t \leq k'$ ?  
The  $i^{th}$  choice to place a marker in the greedy solution must always be farther or equally along the road compared to the  $i^{th}$  choice in the normal solution. ( $d_{g_i} \geq d_{j_i}$ )
- (d) Prove the claim you wrote above using induction on the step  $t$ .
  - i. Base case:  $t = 0 \implies d_{g_0} = d_0 = 0 = d_{j_0}$
  - ii. Induction Step: Assume  $d_{g_{t-1}} \geq d_{j_{t-1}}$ , show  $d_{g_t} \geq d_{j_t}$  for  $t > 0$ 
    - Let  $I$  be the set of valid choices for the greedy algorithm at step  $t$ . It contains markers between  $d_{g_{t-1}}, d_{g_{t-1}} + M$

- Let  $J$  be the set of valid choices for the regular optimal solution. It contains markers between  $d_{j_{t-1}}, d_{j_{t-1}} + M$

By the inductive hypothesis,  $d_{g_{t-1}} \geq d_{j_{t-1}} \implies d_{g_{t-1}} + M \geq d_{j_{t-1}} + M$ . Thus, any valid solution in the regular set,  $j \in J$ , is either less than all solutions in  $I$ , or contained within  $I$ . Because the Greedy algorithm picks the largest marker in  $I$ :

- If the regular solution picks  $d_{j_t} \notin I \implies d_{j_t} < d_{g_t}$
- If the regular solution picks  $d_{j_t} \in I \implies d_{j_t} \leq d_{g_t}$

Thus, the  $i^{th}$  greedy choice,  $d_{g_i}$ , is at least as far along the as the  $i^{th}$  normal choice  $d_{j_i}$

(e) In  $O$  notation, how much time does the algorithm as written take?

The above algorithm runs in  $O(n)$  time. Each one of the  $n$  markers is considered ( $O(1)$ ) exactly once.

### 3. Quests

In a role-playing game, your character will complete a series of  $k$  quests. There is a list of  $n > k$  possible quests  $Q_i$ , each with a first time it can be attempted  $k \geq f_i \geq 1$  and the amount of gold earned on the quest,  $g_i > 0$ . You cannot complete the same quest twice, and you cannot put quest  $i$  in a position before  $f_i$  in your list. You want to maximize your total gold at the end of all your quests.

- (a) Clearly state a greedy strategy that gives the most possible total gold
- (b) Prove that this strategy is correct. Hint: use a modify-the-solution proof with two cases, based on whether the next quest the greedy algorithm performs is performed at some point in  $OS$ .
- (c) Give an efficient algorithm carrying out this strategy and give a time analysis of your algorithm.

#### 4. Minimal spanning subgraph with negative edge weights

Say we allow negative edge weights into the spanning tree problem. We are given a connected undirected graph  $G = (V, E)$  with possibly negative edge weights. We wish to find the set  $E' \subset E$  so that  $G' = (V, E')$  is connected that minimizes  $\sum_{e \in E'} w(e)$ .

- (a) Give an example of a graph with some negative edge weights where the minimum cost connected subgraph is not a tree.
- (b) Which edges are always in the minimum cost subgraph?
- (c) Once the edges above are added, what is an equivalent problem?
- (d) Describe how to use the answers to the above two questions to reduce this problem to minimum spanning tree. Give a time analysis for the resulting algorithm.

## 5. Graph coloring heuristics

The graph coloring problem is, given an undirected graph, give each vertex a color, so that neighboring vertices have different colors. You want to minimize the number of colors. This is an *NP*-complete problem, so we might want to use a greedy heuristic.

One greedy heuristic would be to order the vertices, then assign each the smallest color that is different from its neighbors.

Implement this heuristic on random graphs where we start with  $|V|$  vertices and add  $5|V|$  random edges. For a variety of sizes (e.g.,  $|V| = 2^k$  for  $k = 3, \dots, 12$ ), and several graphs of each size, try the heuristic ordering the vertices by increasing degree (number of neighbors) and decreasing degree. Plot the two numbers of colors used. Do these increase with  $|V|$ ? How consistent are the results with different graphs of the same size?