

## Final Examination

CS 561 Data Structures and Algorithms  
Fall, 2020

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- This exam lasts 4 hours. It is open book, notes and Internet. However, you are not allowed to discuss problems with any person.
  - PLEASE: Start each main problem (i.e. Problems 1-5) at the top of a new page!
  - Give yourself extra time to properly upload your solutions. Late exams may be penalized.
  - *Show your work!* You will not get full credit if we cannot figure out how you arrived at your answer.
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Question	Points	Score	Grader
1	20		
2	20		
3	20		
4	20		
5	20		
Total	100		

1. **Short Answer (2 points each)**

Answer the following questions using the *simplest possible*  $\Theta$  notation. Assume as usual, that  $f(n)$  is  $\Theta(1)$  for constant values of  $n$ .

(a) Solution to the recurrence  $T(n) = T(n - 1) + n$ ?

(b) Expected time to search for a key in a skip list storing  $n$  items?

(c) Solution to the recurrence  $T(n) = 2T(n/4) + n^2$ ?

(d) Solution to the recurrence:  $f(n) = 3f(n - 1) - 2f(n - 2)$ .

(e) In a union-find implementation with union by rank and path-compression, what is the worst case runtime of a single call to FIND-SET?



## 2. Spanning Trees and Cycles

- (a) (10 points) You are given an undirected, connected graph  $G = (V, E)$  with positive weights on all the edges. You want to find a spanning tree of  $G$  with *maximum* weight, i.e. the sum of the weight of all edges in the spanning tree is maximized over all spanning trees of  $G$ . Give an efficient algorithm to solve this problem. Give the runtime of your algorithm as a function of  $n = |V|$  and  $m = |E|$ .
- (b) (10 points) Again you are given a undirected, connected graph  $G = (V, E)$ . Call a subset of edges,  $F$ , a *cycle cover* if every cycle in  $G$  contains at least one edge in  $F$ . In other words, removing the edges of  $F$  from  $G$  results in an acyclic graph. You want to find a cycle cover,  $F$ , of  $G$  with *minimum* weight, i.e. the sum of the weight of all edges in  $F$  is minimized over all cycle covers. Give an efficient algorithm to solve this and give the runtime of your algorithm as a function of  $n = |V|$  and  $m = |E|$ .

### 3. Node Coloring

Consider a collection of  $n$  nodes numbered 1 to  $n$ . For all  $i$ ,  $1 \leq i < n$ , nodes  $i$  and  $i + 1$  are connected by an edge with weight  $w_{i,i+1} \in \mathcal{R}$ , i.e. this weight can be any real number (not just positive).

You want to color each node red or blue. If a pair  $(i, i + 1)$  of neighboring nodes are colored *the same*, the cost associated with the edge connecting those nodes is  $w_{i,i+1}$ ; if the pair are colored differently, the cost of the edge is 0. The total cost of a coloring is the sum of the costs of all edges.

- (a) (6 points) Describe a dynamic program to output the minimum cost of any coloring. Hint: Let  $c(i, r)$  be the minimum cost of coloring nodes 1 through  $i$  when node  $i$  is colored red. Let  $c(i, b)$  be the minimum cost of coloring nodes 1 through  $i$  when node  $i$  is colored blue. What is the runtime of your algorithm
- (b) (7 points) Now you are constrained so that the number of nodes colored red is no more than  $n/2$ . Describe and analyze a dynamic program to output the minimum cost of any such coloring. Hint: Add another argument to your recurrence relation.

- (c) (7 points) Now you will color nodes of a binary tree either red or blue. You are given a tree  $T = (V, E)$ , where each edge  $(u, v) \in E$  has a edge weight,  $w((u, v))$  that is a real number. If both endpoints of the edge are colored the same, the cost associated with that edge is  $w(e)$ , otherwise it is 0. The total cost of a coloring is the sum of the costs of all edges in the tree. There are no constraints on the number of red nodes. Describe and analyze a dynamic program to output the minimum cost of any coloring for  $T$ .  
Hint: For a given node  $v$ , let  $c(v, b)$  ( $c(v, r)$ ) be the minimum cost coloring of the subtree rooted at  $v$  if  $v$  is colored blue (red). It may help to define the following: for a node  $v \in V$ , let  $children(v)$  return the set of nodes that are children of  $v$ .

#### 4. Induction and NP-Hardness

A c-tree is a tree with each node colored either red, green or silver that obeys the following rules.

- Each red node has two children, exactly one of which is green.
- Each green node has exactly one child, which is not green
- Silver nodes have no children.

In the following, let  $R$ ,  $G$ , and  $S$  respectively denote the number of red, green, and silver nodes, and  $n$  be the total number of nodes.

- (a) (10 points) Prove by induction that in any c-tree with  $n \geq 1$ ,  $S = R + 1$ .

Show that the next problem is NP-Hard via a reduction from one of the following problems: 3-SAT, SET-COVER, VERTEX-COVER, INDEPENDENT-SET, 3-COLORABLE, HAMILTONIAN-CYCLE, or CLIQUE.

- (b) (10 points) **WEIGHTED-ITEM-COVER:** You are given (1) a set  $S$  of weighted items; (2) a set  $T$  of subsets of items; and (3) a number  $W$ . You are asked: can you choose a subset  $S'$  of items in  $S$  with total weight of items in  $S'$  no more than  $W$ , such that every subset in  $T$  contains at least one item in  $S'$ ? As an example, let  $S = \{a, b, c, d\}$ ,  $w(a) = w(b) = w(c) = 1$  and  $w(d) = 2$ ;  $T = \{\{a, b, d\}, \{c, d\}, \{b, d\}, \{a, c\}\}$ ; and  $W = 3$ . Then the answer is YES since we can set  $S' = \{a, d\}$ , which has total weight 3 and also ensures that every set in  $T$  contains at least one item from  $S'$ .



## 5. Repeated Rock, Paper, Scissors

In Repeated Rock, Paper, Scissors, in each round, you output a *probability distribution* over  $\{Rock, Paper, Scissors\}$  and then your opponent chooses a value from  $\{Rock, Paper, Scissors\}$ . As usual, Rock beats Scissors; Scissors beats Paper; and Paper beats Rock. If you lose your cost is 1, if you win, your cost is  $-1$ , otherwise, your cost is 0. Your goal is to minimize your total expected cost over  $T$  rounds of this game.

In this game, the best offline cost,  $OPT$ , equals the maximum, over all probability distributions, of the expected cost over all  $T$  rounds using that probability distribution. For example, if  $T = 6$  and your opponent played  $(R, P, S, R, P, S)$ , then  $OPT = 0$ , which can be obtained using probability distribution  $(1/3, 1/3, 1/3)$ . However, if your opponent played  $(R, R, R, R, R, R)$ , then  $OPT = -6$ , which can be obtained using the probability distribution  $(0, 1, 0)$ .

You want to get an expected total cost that is not too much higher than  $OPT$ . You decide to use *online* gradient descent to do this. For round  $i$ , you let  $x_i$  be a vector of length 3 giving the probabilities,  $x_i^r, x_i^p, x_i^s$ , that you use in round  $i$  for choosing rock, paper, scissors, respectively.

(a) (6 points) What are the functions  $f_i(x_i)$  in the online gradient descent algorithm? Hint: There are 3 different functions possible, based on your opponent's choice in round  $i$ .

(b) (2 points) In one sentence, argue that these functions are convex.

(c) (2 points) In one sentence, describe both the convex search space  $\kappa$  and  $D$ , the diameter of  $\kappa$ ?

(d) (2 points) What is  $G$ , the max value of  $|\nabla f_i(x)|$ ? over all  $i$ ,  $1 \leq i \leq T$  and all  $x \in \kappa$ ?

(e) (2 points) Recall that online gradient descent ensures that for any  $x^* \in \kappa$ ,

$$\frac{1}{T} \left( \sum_{i=1}^T f_i(x_i) - \sum_{i=1}^T f_i(x^*) \right) \leq \frac{GD}{\sqrt{T}}$$

Based on this, give a bound on the expected total cost of your algorithm over all  $T$  rounds, as a function of OPT and  $T$  only.

(f) (6 points) You now want to incorporate “memory” of what your opponent played in the previous round (assume there is one practice round before the first, so that there is always a previous value for your opponent). Now OPT uses 3 probability distributions, one for each possible value of what was played in the previous round. For example, if your opponent plays  $(R, P, R, P, R, P)$ , then now  $\text{OPT} = -5$ , which is obtained by using probability distribution  $(0, 0, 1)$  when previous round value is  $R$ ; and probability distribution  $(0, 1, 0)$  when the previous round value is  $P$

Briefly describe how you would change online gradient descent to be competitive with this new OPT. Describe changes to the convex search space  $\kappa$ , and the new  $f_i$  functions. Give a bound on the expected total cost of your algorithm now as a function of OPT and  $T$ .