

# CS 458, Data Structures and Algorithms II

## Homework 1

April 16, 2020

**Due on 11:59am April 22, 2020. Collaboration is not allowed**

For all questions you need to prove your statements or at least provide a justification for the correctness of your algorithm and the calculation of their runtime.

### Problem 1 (50 points)

In the first recitation session we designed an algorithm with  $O(\log k \cdot n^3)$  running time that given a graph  $G = (V, E)$  computes all-pairs shortest paths where each shortest path uses **exactly**  $k$  edges, denoted by  $D_{i,j}^{=k}$ . Note these paths may form a circle as shown in the example below. Particularly, we modified the recursive formula with

$$\hat{d}^k(v_i, v_j) = \min_{v_l \in V \setminus v_j} \{d^{k-1}(v_i, v_l) + w(v_l, v_j)\}$$

and defined a new Extend Shortest Path procedure and applied the repeated squaring. Consider now the similar problem that for a given  $k$  we want to compute a shortest path that uses **at least**  $k$  edges denoted by  $D_{i,j}^{\geq k}$ . Note that the cost of the shortest path that uses at least  $k$  edges never larger than the cost of the shortest path that uses exactly  $k$  edges.

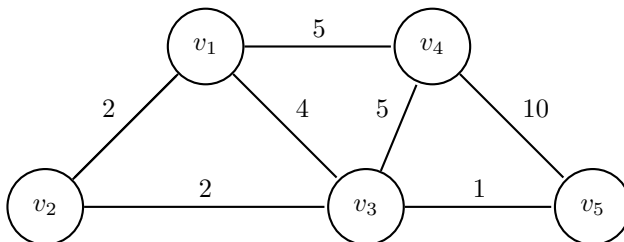


Figure 1: The shortest path with **exactly**  $k = 2$  edges from  $v_2$  to  $v_3$  is  $v_2 \rightarrow v_1 \rightarrow v_3$  for a total cost of  $D_{2,3}^{=k} = w(v_2, v_1) + w(v_1, v_3) = 2 + 4 = 6$

The shortest path with **at least**  $k = 2$  edges vertices from  $v_2$  to  $v_3$  is  $v_2 \rightarrow v_3 \rightarrow v_5 \rightarrow v_3$  for a total cost of  $D_{2,3}^{\geq k} = w(v_2, v_3) + w(v_3, v_5) + w(v_5, v_3) = 2 + 1 + 1 = 4$ .

Note that the actual shortest path from  $v_2$  to  $v_3$  is  $v_2 \rightarrow v_3$  has total cost of  $D_{i,j}^{\geq 0} = w(v_2, v_3) = 2$

- a. **(25 points)** Modify the procedure we discussed in the recitation session to compute  $D_{i,j}^{\geq k}$  for all pairs  $i, j$ . The running-time of your algorithm should be  $O((\log k + n - k) \cdot n^3)$ .
- b. **(25 points)** It turns out we can solve this problem even faster. Give an algorithm that takes as input  $D_{i,j}^{=k}$  and produces  $D_{i,j}^{\geq k}$  for all pairs  $i, j$  with running time  $O(n^3)$ . What is the total runtime?

## Problem 2 (50 points)

In Lecture 3 we showed how to calculate the longest increasing subsequence. This requirement turned out too strict and we are interested to finding the **longest close-to-increasing subsequence**.

We say that a sequence of numbers  $Z$  is close-to-increasing if the size of any decreasing substring is at most 2. So given two consecutive elements in the sequence  $Z_i$  and  $Z_{i+1}$  it must either be that

- $Z_i < Z_{i+1}$
- or if  $Z_i > Z_{i+1}$  it must be that  $Z_{i+1} < Z_{i+2}$ .

Below are some examples of close-to-increasing sequences:

$A$	$=$	11	10	18	17	25	33	30	34	27	29
$B$	$=$	8	99	33	34	40	11	13	20	23	22

As we can see  $A_2 = 10 < 11 = A_1$  but  $A_3 = 18 > 10 = A_2$  so the constrain is satisfied. Same is true in other places where monotonicity fails at  $A_7, A_{11}, A_{14}$ . Similar for sequence  $B$ .

The following are examples of sequences that are not close-to-increasing:

$C$	$=$	10	11	13	12	18	30	27	22	33	40	44
$D$	$=$	17	16	1	11	13	12	15	19	30	25	33

We have  $C_6 = 30 > 27 = C_7$  and  $C_6 = 27 > 22 = C_9$  so the constraint is violated. Similarly  $D_1 > D_2 > D_3$  so the sequence  $D$  is not close-to-increasing. For example if  $X = D$  the longest close-to-increasing subsequence is  $Z = 17, 1, 11, 13, 12, 15, 19, 30, 25, 33$  for a size of 11. You may assume all numbers in  $X$  are **unique**.

- (25 points)** Give a dynamic programming algorithm using the following optimal subproblem structure  $LCIS^+(i, j)$  : is the longest closed to increasing subsequence that ends and  $X_i X_j$ , i.e., its last element is  $X_i$  and its second to last element is  $X_j$ . Give a recursive formula, prove its correctness and write a bottom-up implementation. What is the runtime of your algorithm?
- (25 points)** Give a dynamic programming algorithm using the following optimal subproblem structure  $LCIS^+(i)$  : is the longest closed to increasing subsequence that ends and  $X_i$  and the auxiliary optimal subproblem structure  $LCIS^{+,inc}(i)$  the longest closed to increasing subsequence that ends and  $X_i$  in an increasing fashion (the second to last element is smaller than  $X_i$ ). Give a recursive formula for both optimal subproblem structures, prove their correctness and write a bottom-up implementation. What is the runtime of your algorithm?