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UCLA Computer Science Department
CS 180 Algorithms & Complexity

Final Exam

Total Time: 3 hours

December 10, 2018⁹

***** Write all algorithms in bullet form (as done in the past) *****

You need to prove EVERY answer that you provide.

There are a total of 8 pages including this page.

1. (20 points: each part has 10 points)

a. Consider a S-T network N. Prove that if f is a maxflow in the network N then there is a cut C with its capacity equal to f .

Consider flow f and consider the min cut c .

Two cases:

① $f < |C|$

② $f > |C|$ capacity of

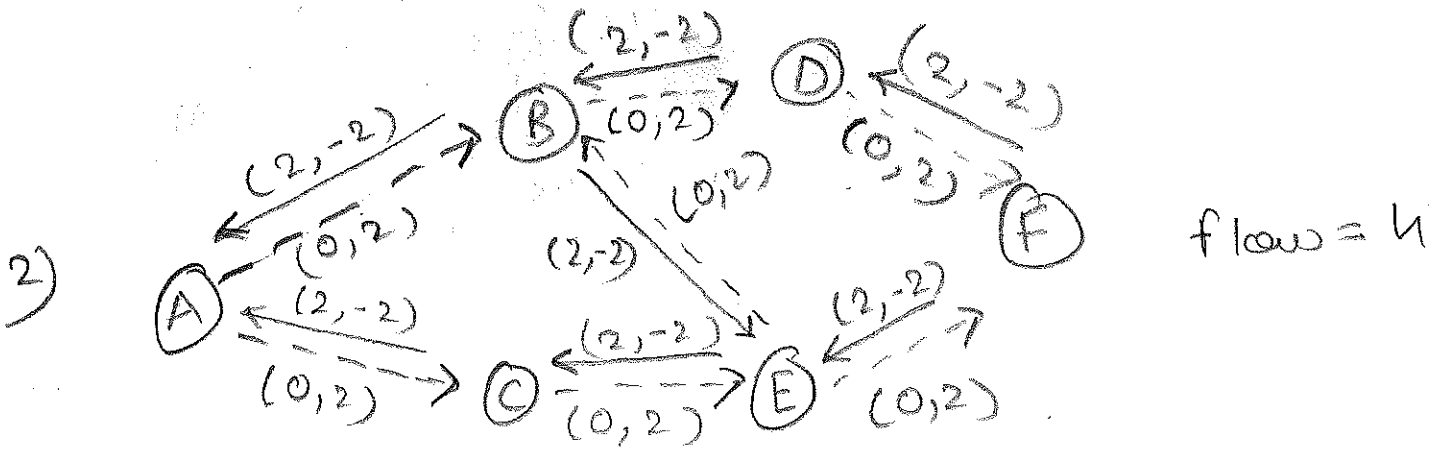
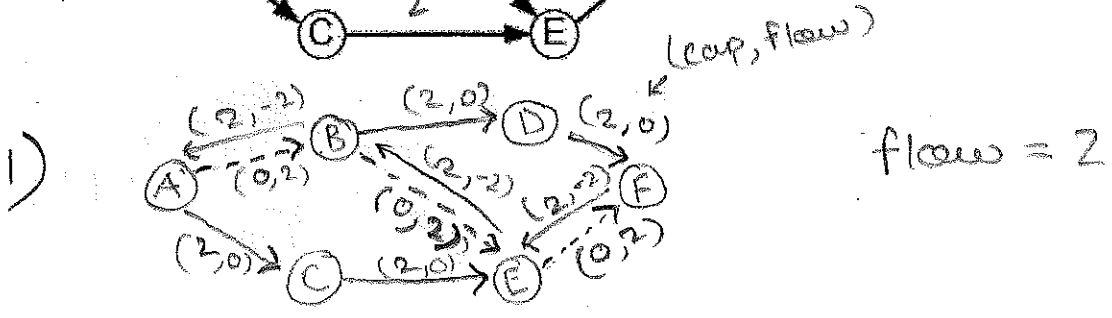
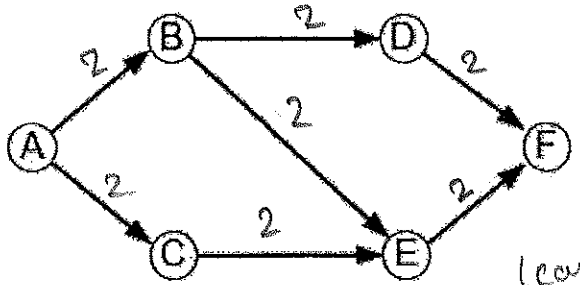
→ ① If flow is less than min cut, it means there is still some amount of flow possible from the set of vertices A that contain source S and the set of vertices B that contain sink t where $S \in A$ and $t \in B$ the two sets of vertices created by ~~the~~ partitioning along the edges of the min cut. This means we can push more flow and that f is not a max flow which is a contradiction.

→ ② If f is greater than capacity of min cut, there must be $f = \text{maxflow}$ units of flow going from S to t. However if we partition the vertices into A and B where $S \in A$ and $t \in B$ along the edges of the min cut, we get a capacity $|C| < f$. There cannot be flow greater than capacity for any set of edges. Thus f cannot be ~~more~~ more than c .

→ Therefore f must be equal to c

→ Thus, if f is a maxflow in the network N, there is a cut (namely the min cut) with its capacity equal to f .

b. Consider the following network with source A and sink F. If the Ford-Fulkerson max flow algorithm initially finds the path A,B,E,F in the network below and sends 2 units of flow on it, show the residual network (also known as the augmented network) and all subsequent steps of Ford-Fulkerson algorithm on this network (all capacities are equal to 2).



2. (15 points) a. Given a $n \times n$ matrix where all numbers are distinct, design an **efficient algorithm** that finds the maximum length path (starting from any cell) such that all cells along the path are in increasing order with a difference of 1.

b. Analyze the time complexity of your algorithm

We can move in 4 directions from a given cell (i, j) , i.e., we can move to $(i+1, j)$ or $(i, j+1)$ or $(i-1, j)$ or $(i, j-1)$ with the condition that the adjacent cells have a difference of 1.

Input: $mat[][] = \begin{Bmatrix} \{1, 2, 9\} \\ \{5, 3, 8\} \\ \{4, 6, 7\} \end{Bmatrix}$

Output: 4
The longest path is 6-7-8-9.

① store all $mat[][]$ in a new 1-d array $arr[]$
Define $Opt(i)$ and set $Opt(i) = 0$ for all $i \leq 0$, for all $i > 0$

② for $i = 1$ to n^2

$a = 0, b = 0$

if $(|arr[i-1] - arr[i]| == 1)$

$a = Opt(i-1)$

if $(|arr[i-n] - arr[i]| == 1)$

$b = Opt(i-n)$

if $(|arr[i-1] - arr[i-n]| == 2)$

$Opt(i) = Opt(i) + a + b$

else $Opt(i) = Opt(i) + \max(a, b)$

endfor

- maintain a global max for $Opt()$ called m and keep updating it when doing step ②

③ Follow $Opt(m)$ backwards, considering $Opt(m-1)$ and $Opt(m-n)$ to generate path P .

- Check if $Opt(m-1) \neq Opt(m-n) = Opt(m-1)$,

- if yes add both to P

- Else add the max to P breaking ties arbitrarily

④ Output P

Proof: - In the original array, $mat[][]$, $arr[i-1]$ represents the element to the left and $arr[i-n]$ represents the element above.

- We consider both of them
- $arr[i+1]$ represents the element to the right and $arr[i+n]$ represents the element below in $mat[][]$
- Our algorithm doesn't consider these when at i but does consider them when at $i+1$ and $i+n$ respectively
- We do this until the end exhaustively checking all possibilities and choose the max.
- Our algorithm also ensures that $arr[i-1]$ and $arr[i-n]$ are in the right order with respect to $arr[i]$.

b) Time Complexity

- ① $O(n^2)$ to construct arr
- ② $O(n^2)$ to loop through $arr[]$
- ③ $O(n^2)$ at worst when all n^2 elements are a part of P .

Overall: $O(n^2)$



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1 2 3
7 8 9
4 6 10

3. (20 points: Each part has 10 points)

- a. Consider d sorted arrays of integers each containing n_1, n_2, \dots, n_d numbers. The numbers n_i 's can be very different. The total number of all elements is n (sum of all n_i 's). Design an $O(n \log d)$ algorithm that merges all arrays into one sorted list. You may wish to use a data structure that we have discussed in class. \rightarrow heap?
- for all (N)
- b. Prove a lower bound on sorting n numbers in the decision tree model (using comparison exchange).

a) Create a min heap h .

① Check the first element of each of the d arrays and ~~find the min~~ store them in the min heap h .

For the array that the min came from say d_i , merge it with n rebalancing as needed

b) Consider a tree of numbers $(1, \dots, N)$.

- There are at most 2^h leaves where h is the height of the tree
- With n numbers we have $n!$ possible leaves.

$2^h \geq n!$ - so we pick $n!$ as we want the lower bound.

- To sort such a tree of numbers we must have at least $\log_2(n!)$ considerations
- $\log_2(n!) \approx n \log n$
- Thus sorting is $\Omega(n \log n)$

4. (15 points)

Consider an array a_1, \dots, a_n of n integers, that is hidden from us. We have access to this array through an procedure $\text{knapsack}(\dots)$. For a set $S \subseteq \{1, \dots, n\}$ and an integer k , $\text{knapsack}(S, k)$ will output "yes" if there is a subset $T \subseteq S$ such that the numbers indexed in T add up to k , and it will output "no" otherwise.

Design an algorithm that calls knapsack only $O(n)$ times and outputs a set $S \subseteq \{1, \dots, n\}$ such that the numbers indexed in S add up to k , if such a set exists. You can use ONLY the knapsack function (e.g., you cannot sort the numbers or do any other operations on them).

For example, suppose $a_1 = 2, a_2 = 4, a_3 = 3, a_4 = 1$, and $k = 7$. Then, $\text{knapsack}(\{1, 2, 3, 4\}, 7)$ returns "yes" and $\text{knapsack}(\{1, 3, 4\}, 7)$ returns "no". In this case your algorithm can output either of the sets $\{1, 2, 4\}$ or $\{2, 3\}$. Note that for example $\{1, 2, 4\}$ are indices of the numbers, that is, a_1, a_2 , and a_4 .

Let $a[i]$ be the array of numbers, S be the output set.

for $i = 1$ to n

flag = $\text{knapsack}(a[i] \text{ without } a[i])$

if (flag = "Yes")

remove $a[i]$ from all future considerations of $a[i]$

else

add i to S

endfor
 → check if $\text{knapsack}(S, k)$ returns yes. If yes then return S , else return empty set.

Proof: There are 2 possibilities after removing an element

① Yes: This means the element was not essential and it is safe to remove as there is a subset without it that adds up to k .

② No: This means there is no subset that adds up to k without this element so it must be present in the final set.

- We finally check if the set S obtained can add up to k to deal with the case where none of the subsets add up to k , so we would have gotten No for each iteration.

Run time: $O(n)$ - we pass through one for loop of $O(n)$.

5. (15 points)

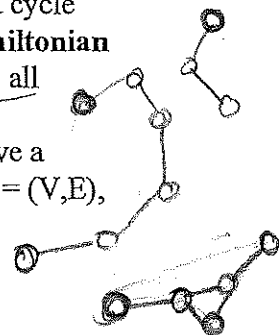
→ undirected

A **Hamiltonian cycle** in a graph with n vertices is a cycle of length n , i.e., it is a cycle that visits all vertices exactly once and returns back to the starting point. A **Hamiltonian path** in a graph with n vertices is a path of length $n-1$, i.e., it is a path that visits all vertices of the graph exactly once.

Hamil-cycle problem is defined as follows: Given a graph $G = (V, E)$, does it have a Hamiltonian cycle? Hamil-path problem is defined as follows: Given a graph $G = (V, E)$, does it have a Hamiltonian path?

Prove that Hamil-path is polynomial-time transformable to Hamil-cycle.

That is $\text{Hamil-path} \leq_P \text{Hamil-cycle}$.



- without loss of generality, take any arbitrary graph.
- Check the in-degree and out degree of each vertex.
- Pick all ^{pairs of} vertices such that in-degree \neq out-degree.
- Say the pair we have is (u, v)
- Run DFS/BFS to find if u has a path to w , construct an edge between
- Say the graph produced by these operations is G' , checking if G' has a hamil cycle is the same as checking if G has a hamil path.

Proof: - If G had a hamil path, there would be only one pair of vertices with $\text{in} \neq \text{out}$, namely the first and the last vertex on the hamil path. They would also be reachable from one another. We construct an edge between them to make a hamil cycle.

- IF G didn't have a hamil path, there would be an odd number vertex pairs with $\text{in} \neq \text{out}$. Moreover, the in degree of ^{at least} one vertex would be 1 and out degree would be zero. This vertex would not have a new edge in our transformation, thus we would have a way to enter this vertex but not leave it. This would yield no hamil cycles.

Thus $\text{Hamil-Path} \leq_P \text{Hamil-cycle}$.

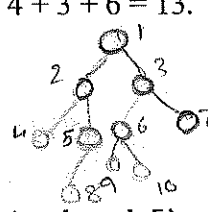
6. (15 points)

You are given a tree T where every node i has weight $w_i \geq 0$.

for any vertex i pick, I cannot pick its children, or its parents

- a. Design a polynomial time algorithm to find the weight of the largest weight independent set in T : among all independent sets one with maximum sum of the weights (an **independent set** is a subset of vertices where there are no edges between any of them).

For example, suppose in the following picture $w_1 = 3, w_2 = 1, w_3 = 4, w_4 = 3, w_5 = 6$. The maximum independent set has nodes 3,4,5 with weight $4 + 3 + 6 = 13$.



①, ④, ⑦, ⑩, ⑤

(1 is connected to 2 and 3. And 2 is connected to 4 and 5)

- b. Analyze the time complexity of your algorithm.

a) ① for each vertex v
 exclude its children and parent from consideration.

② for each remaining vertex w
 - compare its weight with the sum of the weights of its parent and its children.
 - remove the lesser set from consideration.
 - for every vertex removed from consideration bring back its parent and children into consideration and accordingly update other vertices using the same idea.

③ Calculate weight of all vertices remaining and keep a global max weight and update it when necessary.

endfor

Proof: - In a tree, a vertex can only be connected to its parent and its children.

- So we can either include a vertex or include its parent and children
- Our algorithm considers both these choices and picks the best possible outcome

PTO →

b) Run time

① $O(n)$ - we loop for each vertex,

② $O(n^2 \log n)$ - we loop for $O(n)$ vertices, each update can take $O(n)$ and we have at most $\log n$ updates

③ $O(n)$ - we must $O(n)$ vertices

Overall: $O(n^3 \log n + n^2) = O(n^3 \log n)$