

# Algorithms and Programation

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## Homework 4 (5/01/2020)

### Exercise 1

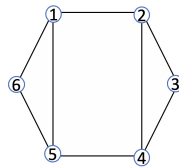
#### Solution

(1)

We want to prove the following theorem:

$G$  is  $k$ -edge-connected  $\Leftrightarrow \forall (v_i, v_j), 1 \leq i, j \leq n, i \neq j$  are joined by  $k$ -pairwise edge disjoint paths

**Note:** There is an ambiguity on the term "path" on the theorem which actually intends "at least  $k$  paths". For example the following graph is 2-edge-connected but there are 3 disjoint paths between vertex 1 and 5.



( $\Leftarrow$ ) Pair of vertices are joined by  $k$ -pairwise edge disjoint paths means that there are  $k$ -paths none of which share an edge between all pairs of vertices. So it is trivial to see that if you delete an edge from each  $k$ -paths i.e.  $k$  edges, you have disconnected the graph. If you delete less than  $k$ -edge then since you have  $k$ -paths then there is at least 1 path that is not destroyed so the graph would still be connected. Hence if every pair of vertices are joined by  $k$  pairwise edge disjoint paths then you need to delete at least  $k$  edges to disconnect the graph i.e.  $G$  is  $k$ -edge connected.

( $\Rightarrow$ ) For  $G$  to be  $k$ -edge connected at least  $k$  edges need to be cut in order to disconnect it. Let  $X$  be the set of minimum edges which if removed they disconnect the graph into two parts  $A$  and  $B$ . Consequently we have that  $\text{card}(X) = k$ . Let  $\forall e \in G, c(e) = 1$ . In other words " $G$  is  $k$ -edge connected" is equivalent to the minimum cut problem  $\forall (s, t) \in A \times B$ . So the value of the minimum cut is  $k$  since the minimum amount of edges you have to remove is  $k$ , ( $\text{card}(X)$ ). By the max flow min cut theorem we have that the value of the minimum cut is equal to the value of the maximal flow i.e.  $k$ . Now we have two case:

**case 1:** Fix  $(s, t) \in A \times B$ . We know that the maximal flow from  $s$  to  $t$  is  $k$ . Suppose that there were less than  $k$  disjoint paths then the maximal flow would be strictly lower than  $k$ . So the value of maximal flow being  $k$  forces that there must be  $k$  disjoint paths.

**case 2:** Wlog, fix  $(s, t) \in A$ . Then either there exists another  $X'$  such that the removal of it would separate the graph into a new  $A', B'$  such that  $s \in A'$  and  $t \in B'$  or there exists no such set. If that's the case then by the contrapositive of the first implication proof we have that there are more than  $k$  paths. Hence at least  $k$  paths. In all cases there are at least  $k$ -pairwise edge disjoint paths  $\forall s, t \in V$ .

By double implication this concludes the proof.

(2) The algorithm takes as input an undirected graph  $G = (V, E)$  and outputs the minimum edge cut  $X$ . We start by replacing each undirected edge in  $G$  with two oppositely directed edges both of capacity 1. Let  $G_f$  be the resulting graph. Then we choose an  $s \in V$  at random and apply the Ford-Fulkerson algorithm  $\forall t \in V \setminus \{s\}$  which will compute the maximum flow for all  $t$ . Then we take  $t_{\min}$  which has the minimal maximal flow and its residual graph  $G_r$ . By the corollary in the lecture we are sure that we can compute the min-edge-cut from  $G_r$  (in linear time). The algorithm is correct and terminates.

**Time complexity** We are iterating over  $n - 1$  vertices the Ford-Fulkerson algorithm which has complexity

$\mathcal{O}(fm) = \mathcal{O}(mn)$  since  $f \leq nC = n$  and  $C = 1$ . Finally we compute the min-edge-cut in  $\mathcal{O}(n)$  which makes the time complexity of our algorithm be  $\mathcal{O}(mn^2)$ . So our algorithm runs in poly-time.

## Exercise 2

### Solution

First we show that the greedy algorithm terminates. Let  $S_E$  be the set of edges of the cut which initially is equal to  $\emptyset$ . Let  $|e_v|$  be the cardinality of incident edges of a vertex  $v$ . We know that  $|e_v| = |nc_v| + |c_v|$  where  $nc_v$  is the not-crossing edges and  $c_v$  is the crossing edges of  $v$ . Also by step b we have that  $|c_v| < \frac{1}{2}|e_v|$ . Assume we want to add in the  $S_E$  a new set of edges, call the new formed set  $S'_E$ . Hence we have that:

$$\begin{aligned} |S'_E| &= |S_E| + |nc_v| - |c_v| = |S_E| + |e_v| - 2|c_v| \\ |S'_E| &> |S_E| + |e_v| - 2\frac{1}{2}|e_v| = |S_E| \end{aligned}$$

So the cardinality is increasing and since  $|S'_E|$  is bounded above by  $|E|$  then that means that the algorithm terminates.

Now we show that the algorithm is a 2-approximation of Max-Cut. Since this is a maximization problem then we need to prove that  $ALG(P) \geq \rho(n)OPT(P)$  where  $\rho(n) < 1$ . Let  $MC$  be the max-cut i.e. the optimal solution and  $S_E$  the solution that the greedy algorithm outputs. We notice at step b that we add more than  $\frac{1}{2}$  of the incident edges so in the end that would result in the following inequality  $|S_E| \geq \frac{|E|}{2}$ . Hence we have the following which proves that the greedy algorithm is 2-approximation of the Max-Cut:

$$|S_E| \geq \frac{1}{2}|E| \geq \frac{1}{2}|MC|$$

Finally we show that the randomized algorithm is also a 2-approximation of the Max-Cut. We keep the same notation. Then using the linearity of expectation we have that:

$$\begin{aligned} \mathbb{E}[|S_E|] &= \sum_{(v_1, v_2) \in G.E} \mathbb{P}(v_1, v_2 \in A \times B \vee v_1, v_2 \in B \times A) = \sum_{(v_1, v_2) \in G.E} (\mathbb{P}(v_1, v_2 \in A \times B) + \mathbb{P}(v_1, v_2 \in B \times A)) \\ &= \sum_{(v_1, v_2) \in G.E} \left(\frac{1}{4} + \frac{1}{4}\right) = \frac{|E|}{2} \geq \frac{1}{2}|MC| \end{aligned}$$

## Exercise 3

### Solution

"Become infinitely rich" means that there exists a sequence of currencies  $c_1, \dots, c_k$  that when multiplying their exchange rates, the product will be bigger than 1 i.e.

$$R[i_1, i_2] \cdot R[i_2, i_3] \cdots R[i_{k-1}, i_k] \cdot R[i_k, i_1] > 1$$

We transform this sequence in the following way so we could apply the Bellman-Ford algorithm which returns False if there exists a negative cycle in the weighted graph of currencies:

$$\begin{aligned} \log(R[i_1, i_2] \cdot R[i_2, i_3] \cdots R[i_{k-1}, i_k] \cdot R[i_k, i_1]) &> \log(1) \\ \log(R[i_1, i_2]) + \log(R[i_2, i_3]) + \cdots \log(R[i_{k-1}, i_k]) + \log(R[i_k, i_1]) &> 0 \\ -\log(R[i_1, i_2]) - \log(R[i_2, i_3]) - \cdots \log(-R[i_{k-1}, i_k]) - \log(R[i_k, i_1]) &< 0 \end{aligned}$$

Hence if the algorithm returns False then there exists such a sequence in the graph, i.e. it is time to become rich. The time complexity of the Bellman-Ford algorithm is  $\mathcal{O}(VE)$ . Assume we have  $n$  vertices then in a complete digraph we would have  $\frac{n^2-1}{2}$  edges so the time complexity is  $\mathcal{O}(n^3)$ . The printing and transforming of such sequence back to its original form would take  $\mathcal{O}(n)$ . In conclusion this algorithm would be of time complexity  $\mathcal{O}(n^3)$  and space complexity  $\mathcal{O}(n^2)$  (for the storing of the table  $R$ ).