

# CS 38, PS 7

Codename: Jambul

June 1, 2023

## 1 Problem 1

*Proof.* Denote the systems

$$P_1 = \begin{cases} y^T A = c^T \\ y^T \geq 0 \end{cases} \iff A^T y = c, \quad P_2 = \begin{cases} Ax \leq 0, \\ c^T x > 0 \end{cases}.$$

Set  $B = A^T$  so we may write the systems as

$$P_1 = \begin{cases} By = c, \\ y^T \geq 0 \end{cases} \quad P_2 = \begin{cases} B^T x \leq 0, \\ c^T x > 0 \end{cases}.$$

We can now see how to apply LP strong duality in order to complete the proof. Consider the primal-dual pair

$$\mathcal{P} = \begin{cases} \min 0y \\ \text{s.t. } By = c, \\ y \geq 0 \end{cases} \quad \mathcal{D} = \begin{cases} \max c^T x \\ \text{s.t. } B^T x \leq 0 \end{cases}.$$

Taking  $x = 0$ , we can see that  $\mathcal{D}$  is trivially feasible. Suppose that  $\mathcal{P}$  is infeasible. Then it must be the case that  $\mathcal{D}$  is unbounded, since otherwise we would have that  $\mathcal{D}$  is feasible and bounded, which would imply from strong duality that  $\mathcal{P}$  and  $\mathcal{D}$  have equivalent optimal values. This would be a contradiction, since  $\mathcal{P}$  is assumed to be infeasible. However, if  $\mathcal{D}$  then must be unbounded and feasible, then clearly we can find  $x$  such that  $c^T x > 0$  and  $B^T x \leq 0$ , meaning that it cannot be the case that both  $P_1$  and  $P_2$  are infeasible.

Now suppose that  $\mathcal{P}$  is feasible. Then there exists  $y \in \mathbb{R}^n$  such that  $By = c$ ,  $y \geq 0$ . Since  $\mathcal{P}$  will then need to be bounded to satisfy its constraints ( $By = c$ ), we have that  $\mathcal{P}$  is feasible and bounded, so its optimal value must match that of  $\mathcal{D}$ . The optimal value of  $\mathcal{P}$  will be 0, since  $\mathcal{P}$  seeks to minimize  $0y$ . Thus  $\mathcal{D}$  will have to have that its optimal value is  $c^T x = 0$ . However, this would imply that  $c^T x \not> 0$ , and thus if  $P_1$  is feasible, then  $P_2$  is not.  $\square$

## 2 Problem 2

**Algorithm 1.** We will use Ford-Fulkerson algorithm for max flows to solve the problem.

1. Create a new directed graph  $G' = (V', E')$  from  $G = (V, E)$ , so  $(u, v) \in E \implies (u, v), (v, u) \in E'$ .
2. Set all edge capacities in  $G'$  to 1.
3. Pick an arbitrary node  $v \in V'$  and compute the maximum flow  $g(v, w)$  from  $v$  to each  $w \in V' \setminus \{v\}$ .
4. Return  $\min_{w \in V', w \neq v} \{g(v, w)\}$ .

*Proof of correctness.* Let  $c = \min_{w \in V', w \neq v} \{g(v, w)\}$ . We must show that  $c$  is equal to the edge connectivity (let this be  $k$ ) of  $G$ .

Let  $H$  be the subset of edges in  $E$  such that removal of  $H$ , where  $H$  has size  $k$ , from  $G$  will disconnect  $G$  into two non-empty subgraphs  $G'_1 = (V'_1, E'_1), G'_2 = (V'_2, E'_2)$ . Now take any  $v \in V'_1, w \in V'_2$ . Let  $g(v, w)$  then be the calculated max flow from  $v$  to  $w$  with all edge capacities in  $E'$  set to 1.

From the max-flow min-cut theorem, then, we know that  $g(v, w)$  will also be the size of a min cut between vertices  $v$  and  $w$ . Note furthermore that the size of a min cut between these two vertices is at most  $k$ , since if it takes at minimum  $k$  edges to get  $G'_1, G'_2$  from  $G$ , then a min cut (and thus a max flow) between  $v$  and  $w$  should be of size less than or equal to  $k$ . From this observation, we now have that  $c \leq g(v, w) \leq k$ .

However, it must also be the case that  $c \geq k$ , since otherwise there exists a cut of  $G$  with size smaller than  $k$ , which would be a contradiction. Thus we have that in fact  $c = g(v, w) = k$ , as desired.

*Complexity.*  $G'$  has  $|V|$  vertices and  $2|E|$  edges. It then solves a max-flow problem from  $v$  to every other  $w \in V' \setminus \{v\}$ , so it solves  $|V| - 1$  max-flow problems on a network with  $\mathcal{O}(|V|)$  vertices and  $\mathcal{O}(|E|)$  edges.