

# CS170–Spring 2019 — Homework 11 Solutions

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## 1 Study Group

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## 2 Zero-Sum Battle

(a)

$$\max p$$

$$p \leq -10x_1 + 4x_2 + 6x_3 \text{ (payoff when trainer B chooses the ice Pokemon)}$$

$$p \leq 3x_1 - 1x_2 - 9x_3 \text{ (payoff when trainer B chooses the water Pokemon)}$$

$$p \leq 3x_1 - 3x_2 + 2x_3 \text{ (payoff when trainer B chooses the fire Pokemon)}$$

$$x_1 + x_2 + x_3 = 1$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$

$$x_3 \geq 0$$

The optimal strategy is  $(0.335, 0.563, 0.102)$  and the payoff is  $-0.48$ .

(b)

$$\min p$$

$$p \geq -10y_1 + 3y_2 + 3y_3 \text{ (payoff when trainer A chooses the dragon Pokemon)}$$

$$p \geq 4y_1 - 1y_2 - 3y_3 \text{ (payoff when trainer A chooses the steel Pokemon)}$$

$$p \geq 6y_1 - 9y_2 + 2y_3 \text{ (payoff when trainer A chooses the rock Pokemon)}$$

$$y_1 + y_2 + y_3 = 1$$

$$y_1 \geq 0$$

$$y_2 \geq 0$$

$$y_3 \geq 0$$

The optimal strategy is  $(0.268, 0.323, 0.409)$  and the payoff is  $-0.48$ .

### 3 Zero Sum Games

- (a)  $x_1$  is the probability that Alice will play strategy 1.  
 $x_2$  is the probability that Alice will play strategy 2.  
 $p$  is Alice's payoff

(b)

$$\max p$$

$$p \leq 4x_1 + 2x_2 \text{ (payoff when Bob take strategy 1)}$$

$$p \leq x_1 + 5x_2 \text{ (payoff when Bob take strategy 2)}$$

$$x_1 + x_2 = 1$$

$$x_1 \geq 0$$

$$x_2 \geq 0$$

$$p \geq 1$$

(c)

$$\max p$$

$$p \leq 4x_1 + 2(1 - x_1) = 2x_1 + 2$$

$$p \leq x_1 + 5(1 - x_1) = -4x_1 + 5$$

$$0 \leq x_1 \leq 1$$

$$p \geq 1$$

(d)

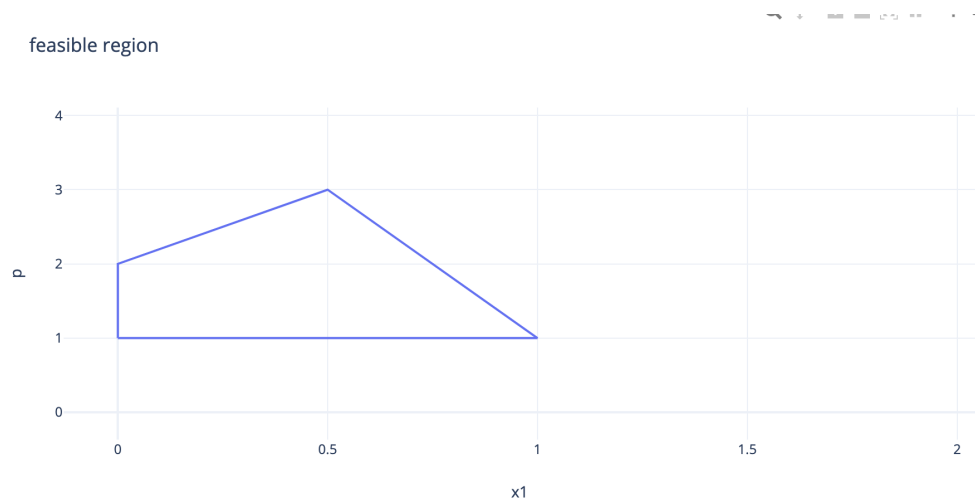


Figure 1: Feasible Region

- (e) The optimal solution is  $(\frac{1}{2}, \frac{1}{2})$  and the value of game is 3.

## 4 Repairing a Flow

### (a) Main Idea

Build the residual network based on maximum flow  $f$ . Then find a path from  $t$  to  $s$  that go through edge  $(v, u)$  in residual network. Push 1 unit flow into this path and fix the wrong capacity from  $c_{uv}$  to  $c_{uv} - 1$ . Then run the original max-flow algorithm starting from this residual network.

### (b) Proof of Correctness

Pushing 1 unit back from  $t$  to  $s$  that go through edge  $(v, u)$  will make the flow go through edge  $(u, v)$  be valid. Then run the original max-flow algorithm again from this is sure will give an optimal solution.

### (c) Runtime Analysis

Build residual network will cost  $O(|E|)$  time. Find a path from  $t$  to  $s$  that go through edge  $(v, u)$  by DFS or BFS will cost  $O(|V| + |E|)$  time. Then push 1 unit into this path will cost at most  $|E|$  time. Since the max-flow solution after repairing capacity of edge  $(u, v)$  cannot be larger than the original one. At most 1 iteration is needed if run max-flow algorithm starting from this. So total runtime is still linear, which is  $O(|V| + |E|)$ .

## 5 Generalized Max Flow

### (a) Constraints

#### (i) Capacity Constraints

$$\sum_{i=1}^k f_e^{(i)} \leq c_e \quad \text{for } \forall e$$

#### (ii) Flow Conservation

$$\sum_{i=1}^k \sum_{u:(u,v) \in E} f_{(u,v)}^{(i)} = \sum_{i=1}^k \sum_{u:(v,w) \in E} f_{(v,w)}^{(i)} \quad \text{for } \forall v \text{ except } s_1, \dots, s_k, t_1, \dots, t_k$$

#### (iii) Nonnegativity

$$f_e^{(i)} \geq 0 \quad \text{for } \forall e, i = 1, \dots, k$$

#### (iv) Demand Constraints

$$\sum_{u:(u,t_i) \in E} f_{(u,t_i)}^{(i)} \geq d_i \quad \text{for } i = 1, \dots, k$$

### (b) Objective

$$\max \sum_{i=1}^k \sum_{u:(u,t_i) \in E} f_{(u,t_i)}^{(i)}$$

The sum of all flows go to destination.

## 6 Reductions Among Flows

- (a) Separate each vertex into two new vertices and denote them as  $v_{in}$  and  $v_{out}$ . Then link all incoming edges in original graph  $G$  to  $v_{in}$  and all outgoing edges to  $v_{out}$ . More formally, create a new edge  $(u, v_{in})$  in new graph if edge  $(u, v)$  is in original graph and create a new edge  $(v_{out}, u)$  in new graph if edge  $(v, u)$  is in original graph. Finally, create a new edge  $(v_{in}, v_{out})$  with capacity  $c_v$ .

Proof:

If  $F$  is a flow in  $G$  satisfying the vertex capacity constraint, set  $f_{(v_{in}, v_{out})}$  to be  $\sum_{u:(u,v) \in E} f_{uv}$  and let all other flows remain same. This is a valid flow because  $f_{(v_{in}, v_{out})} = \sum_{u:(u,v) \in E} f_{uv} \leq c_v$ . And since all other flows are unchanged, it's flow with same size.

If  $F'$  is a flow in  $G'$ , ignore all  $f_{(v_{in}, v_{out})}$  will give a valid flow  $F$  in  $G$  with same size.

- (b) Create an artificial vertex  $s$  and create edges  $(s, s_1), \dots, (s, s_k)$  with  $\infty$  capacity.

Proof:

If  $F$  is a flow in  $G$  satisfying the vertex capacity constraint, there's always a possible solution in  $G'$  with same size. Since the capacity for edges  $(s, s_1), \dots, (s, s_k)$  is infinity. It can be arbitrary large. So just assign  $f_{(s, s_i)}$  with  $\sum_{w:(s_i, w) \in E} f_{(s_i, w)}$ .

If  $F'$  is a flow in  $G'$ , ignore all  $f_{(s, s_i)}$  will give a valid flow  $F$  in  $G$  with same size.

## 7 A Flowy Metric

- (a) According to **Max-flow min-cut theorem**, the size of the maximum flow in a network equals the capacity of the smallest  $(s, t)$ -cut. Since any cut in  $G$  has capacity at least 1, the max flow from  $s$  to  $t$  is at least 1.
- (b) Suppose a flow  $f$  send 1 unit along the shortest path from  $s$  to  $t$ , the length of  $f$  is  $1 \times d(s, t) = d(s, t)$ . Therefore,  $d_{flow}(s, t)$  is smaller than or equal to  $d(s, t)$  by definition, i.e.  $d_{flow}(s, t) \leq d(s, t)$ .

In general, a flow can send a fraction of unit along different path from  $s$  to  $t$ . Suppose flow  $f$  send  $f_1, f_2, \dots, f_n$  units along different path with length  $d_1, d_2, \dots, d_n$ . And  $\sum_{i=1}^n f_i = 1$ .

$$d_{flow}(s, t) = \sum_{i=1}^n d_i f_i \geq \sum_{i=1}^n d(s, t) f_i = d(s, t) \sum_{i=1}^n f_i = d(s, t)$$

Since  $d_{flow}(s, t) \leq d(s, t)$  and  $d_{flow}(s, t) \geq d(s, t)$ ,  $d_{flow}$  must equal to  $d(s, t)$ .