## Tutorial 8: Applications of Linear Programming

**Problem 1.** (Densest Subgraph) Consider a graph G = (V, E). For a non-empty subset of vertices  $S \subseteq V$ , let  $E(S) = \{(u, v) \in E : u, v \in S\}$  be the set of edges within S. Define the density of S as d(S) = |E(S)|/|S|. The goal is to find a non-empty subset S of maximum density. To solve this problem, consider the following LP for the problem.

LP variables:  $x_u$  for every vertex u and  $y_e$  for every edge e.

$$\mathbf{maximize} \quad \sum_{e \in E} y_e$$

s.t.

$$\sum_{u \in V} x_u \le 1$$
 
$$y_{(u,v)} \le x_u \qquad \text{for every } u,v \in V$$
 
$$x_u,y_e \ge 0 \qquad \text{for every } u \in V \text{ and } v \in V$$

Let OPT be the optimal value of the problem and LP be the value of this linear program.

- 1. Prove that  $LP \geq OPT$ . To this end, consider an optimal solution  $S^*$  of value OPT and define a corresponding LP solution of the same value.
- 2. Define set  $S_t = \{u : x_u \ge t\}$ . Let  $T = \max_{u \in V} x_u$ .
- 3. Choose t uniformly at random from [0,T]. What is the probability that edge (u,v) belongs to  $E(S_t)$ .
- 4. What is the expected number of vertices in  $S_t$ ?
- 5. What is the expected number of edges in  $E(S_t)$ ?
- 6. Prove that  $\mathbf{E}_t[|E(S_t)| LP \cdot |S_t|] \ge 0$ .
- 7. Propose an LP-based algorithm for solving the Densest Subgraph Problem.

## Solution

1. For each  $u \in V$  and  $(u, v) \in E$ , set

$$x_u = \begin{cases} 1/|S^*| & \text{if } u \in S^*, \\ 0 & \text{otherwise;} \end{cases} \quad y_{(u,v)} = \begin{cases} 1/|S^*| & \text{if both } u, v \in S^*, \\ 0 & \text{otherwise.} \end{cases}$$

We need to show that the variables  $x_u$  and  $y_{(u,v)}$  make up a feasible solution whose value is  $S^*$ . For the latter, notice that

$$\sum_{e \in E} y_e = \sum_{(u,v) \in E \& u,v \in S^*} y_{(u,v)} = \frac{|E(S^*)|}{|S^*|} = d(S^*).$$

Regarding feasibility, notice that we give non-negative values to the variables, and  $y_{(u,v)} > 0$  means that  $u \in S$ , hence  $x_u = y_{(u,v)}$ . Finally, we have that

$$\sum_{u \in V} x_u = \sum_{u \in S^*} 1/|S^*| = 1.$$

- 3. The probability that the edge (u, v) belongs to  $E(S_t)$  is the probability that both  $x_u \ge t$  and  $x_v \ge t$ , which is equal to  $\min\{x_u, x_v\}/T$ .
- 4. To compute the expected number  $\mathbf{E}[|S_t|]$  of vertices in  $S_t$ , consider the random variables  $X_u$  for every  $u \in V$ , defined as

$$X_u = \begin{cases} 1 & \text{if } x_u \ge t, \\ 0 & \text{otherwise.} \end{cases}$$

We have that  $\mathbf{E}[X_u] = P[x_u \ge t] = x_u/T$ , and hence

$$\mathbf{E}[|S_t|] = \mathbf{E}\left[\sum_{u \in V} X_u\right] = \sum_{u \in V} \mathbf{E}[X_u] = \frac{1}{T} \sum_{u \in V} x_u.$$

5. Similarly, to compute the expected number  $\mathbf{E}[|E(S_t)|]$  of edges in  $S_t$ , consider the random variables  $Y_{(u,v)}$  for every  $(u,v) \in E$ , defined as

$$Y_{(u,v)} = \begin{cases} 1 & \text{if } x_u, x_v \ge t, \\ 0 & \text{otherwise.} \end{cases}$$

We have  $\mathbf{E}[Y_{(u,v)}] = P[x_u, x_v \ge t] = \min\{x_u, x_v\}/T$ , and

$$\begin{split} \mathbf{E}\left[|E(S_t)|\right] &= \mathbf{E}\left[\sum_{(u,v)\in E} Y_{(u,v)}\right] = \sum_{(u,v)\in E} \mathbf{E}[Y_{(u,v)}] \\ &= \frac{1}{T}\sum_{(u,v)\in E} \min\{x_u,x_v\} \ge \frac{1}{T}\sum_{(u,v)\in E} y_{(u,v)} = \frac{LP}{T}. \end{split}$$

6. We have that

$$LP \cdot \mathbf{E}[|S_t|] = \frac{LP}{T} \sum_{u \in V} x_u \le \frac{LP}{T} \le \mathbf{E}[|E(S_t)|],$$

therefore

$$\mathbf{E}[|E(S_t)| - LP \cdot |S_t|] \ge 0.$$

7. Since  $\mathbf{E}[|E(S_t)| - LP \cdot |S_t|] \ge 0$ , there must be a  $t^*$  for which  $|E(S_{t^*})| - LP \cdot |S_{t^*}| \ge 0$ . For that  $t^*$ ,

$$d(S_{t^*}) = \frac{|E(S_{t^*})|}{|S_{t^*}|} \ge LP,\tag{1}$$

and therefore  $OPT \geq LP$ . We already saw that  $OPT \leq LP$ , thus OPT = LP. That is, running our linear program, the value we get is an optimal value for our problem. Moreover, (1) guarantees that such a value will be achieved by some  $S_{t^*}$ . How to find such a  $t^* \in [0, T]$ ? Well, there is only a linear number of values we have to try, namely, the values  $x_u$  for every  $u \in V$ . So we can try them all: After our linear program returns with some solution, we can check in linear time whether  $d(S_t) = LP$  for some  $t = x_u$ .