

# CS217 - Algorithm Design and Analysis

## Homework Assignment 5

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### 4 Linear Programs and Farkas Lemma

#### Exercise 4.2.

##### Necessity.

Obviously if  $y \in \{0, 1\}$  and the set  $C := \{u \in V \mid y_u = 1\}$  is a minimal vertex cover, then  $y$  is a feasible solution (every edge is covered by a node, namely  $\forall (u, v) \in E, y_u + y_v \geq 1$ ). To show the solution is a basic solution, we firstly show that the following lemma is true.

**Lemma 1.** *If  $C$  is a minimal vertex cover, then  $\forall u \in C, \exists v \notin C$ , such that  $(u, v) \in E$ .*

*Proof.* Otherwise if there exists  $x$  satisfied that  $\forall (u, v) \in E$  and  $v \notin C$ . Then remove  $u$  from  $C$  and we can get  $C'$ . Obviously  $C'$  is a vertex cover, which contradicts to the fact that  $C$  is a minimal vertex cover.  $\square$

Considering the nodes  $u$  satisfying  $u \notin C$ , then  $m + u \in I(y)$  holds. These nodes donate an identity matrix in  $A_{I(y)}$ . Then considering the nodes  $u$  which is in the vertex cover, then according to the lemma, we can find another adjacent vertex  $v \notin C$ , and  $y_u + y_v = 1$ , namely  $id(u, v) \in I(y)$  where  $id(u, v)$  stands for the serial number of the edge  $(u, v)$ . Hence  $A_{I(y)}$  is consisted of two kinds of rows, and one of them is  $a_{id(u, v)}$  (either  $u$  or  $v$  is in  $C$  and  $(u, v) \in E$ ) and  $a_{m+i}$  ( $i$  is not in  $C$ ). By Gauss elimination, it is obviously that  $\text{rank}(A_{I(y)}) = n$ , hence  $y$  is a basic solution.

##### Sufficiency.

Firstly we will show that given a basic and feasible solution  $y, y_u \in \{0, 1\}$  holds. We can construct a graph  $G' = (V', E')$  where  $V' = \{u \mid y_u \in (0, 1)\}$  and  $E' = \{(u, v) \mid u, v \in V', y_u + y_v = 1\}$ . Then we know that the degree of each vertex  $i$  is at least one, otherwise the  $i$ -th column of  $A_{I(y)}$  is a zero vector ( $y_i \neq 0$  and every edge  $(u, v)$  is not “tight”), contradicting to the fact that  $A_{I(y)}$  is a non-singular matrix.

**Lemma 2.** *If a graph  $G$  has  $n$  vertexes and  $n$  edges, then there must be a cycle in  $G$ .*

*Proof.* Otherwise  $G$  is a forest and has at most  $n - 1$  edges.  $\square$

According to the lemma,  $G'$  has a cycle  $c_1, c_2, \dots, c_t$ . Let  $E_c$  be the set containing the serial number of the edges in the cycle. With the rearrangement of the rows, the submatrix  $A_{E_c}$  itself become:

$$\begin{pmatrix} 1 & 1 & 0 & \cdots & 0 & 0 \\ 0 & 1 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 1 & 1 \\ 1 & 0 & 0 & \cdots & 0 & 1 \end{pmatrix}$$

And obviously the above matrix is singular (and the original matrix  $A$  is singular too), which leads the contradictions.

Then we will show that  $C := \{u \in V \mid y_u = 1\}$  is a minimal vertex cover. Showing that  $y_u \in \{0, 1\}$ , we can easily know that  $y$  is a vertex cover. If  $y$  is not a minimal vertex cover, then there must be a node  $p \in C$ , after we remove which we can obtain another vertex cover  $y'$ . Consider the  $p$ -th column of  $A$ , and according to the lemma 1, we can see that there is no adjacent node  $q$  such that  $y_p + y_q = 1$ , which means that the  $p$ -th column of  $A$  is a zero vector, which leads a contradiction.

**Exercise 4.6.**

## FLGV

According to **FLGV-2**, we have:

$$\exists \mathbf{y} \in \mathbf{R}^m : A^T \mathbf{y} \geq 0, \mathbf{b}^T \mathbf{y} < 0$$

if FLGV-1 also holds, there exists  $\mathbf{x} \geq 0$ ,  $A\mathbf{x} = \mathbf{b}$ . because  $A\mathbf{x} = \mathbf{b}$  can be written as  $\mathbf{b}^T = \mathbf{x}^T A^T$ , we have:

$$\mathbf{b}^T \mathbf{y} = \mathbf{x}^T A^T \mathbf{y}$$

Because  $A^T \mathbf{y} \geq 0$ ,  $\mathbf{x} \geq 0$ , we can get  $\mathbf{x}^T A^T \mathbf{y} \geq 0$ . But according to **FLGV-2**,  $\mathbf{b}^T \mathbf{y} < 0$ , which is a contradictory, so FLGV-2 implies that **FLGV-1** does not hold.

## FLLPV

According to **FLLPV-2**, we have:

$$\exists \mathbf{y} \in \mathbf{R}^m : \mathbf{y} \geq 0, A^T \mathbf{y} \geq 0, \mathbf{b}^T \mathbf{y} < 0$$

if **FLLPV-1** also holds, there exists  $\mathbf{x} \geq 0$ ,  $A\mathbf{x} \leq \mathbf{b}$ . Because  $A\mathbf{x} \leq \mathbf{b}$  can be written as  $\mathbf{b}^T \geq \mathbf{x}^T A^T$ , and  $\mathbf{y}$  holds  $\mathbf{y} \geq 0$ , we have:

$$\mathbf{b}^T \mathbf{y} \geq \mathbf{x}^T A^T \mathbf{y}$$

Because  $A^T \mathbf{y} \geq 0$ ,  $\mathbf{x} \geq 0$ , we can get  $\mathbf{b}^T \mathbf{y} \geq \mathbf{x}^T A^T \mathbf{y} \geq 0$ . But according to **FLLPV-2**,  $\mathbf{b}^T \mathbf{y} < 0$ , which is a contradictory, so **FLLPV-2** implies that **FLLPV-1** does not hold.

## FLEV

According to **FLEV-2**, we have:

$$\exists \mathbf{y} \in \mathbb{R}^m : \mathbf{y} \geq 0, A^T \mathbf{y} \geq 0, \mathbf{b}^T \mathbf{y} < 0$$

If **FLEV-1** also holds, there exists  $A\mathbf{x} \leq \mathbf{b}$ . Because  $A\mathbf{x} \leq \mathbf{b}$  can be written as  $\mathbf{b}^T \geq \mathbf{x}^T A^T$ , and  $\mathbf{y}$  holds  $\mathbf{y} \geq 0$ , we have:

$$\mathbf{b}^T \mathbf{y} \geq \mathbf{x}^T A^T \mathbf{y}$$

Because  $A^T \mathbf{y} \geq 0$ ,  $\mathbf{x} \geq 0$ , we can get  $\mathbf{b}^T \mathbf{y} \geq \mathbf{x}^T A^T \mathbf{y} \geq 0$ . But according to **FLEV-2**,  $\mathbf{b}^T \mathbf{y} < 0$ , which is a contradictory, so **FLEV-2** implies that FLEV-1 does not hold.

**Exercise 4.7.**

## FLGV $\Rightarrow$ FLLPV

Construct  $\mathbf{A}' = \begin{pmatrix} \mathbf{A} & \mathbf{I} \end{pmatrix}$  where  $\mathbf{I}$  is identity matrix. Use algorithm for FLGV on  $\mathbf{A}'$  and  $\mathbf{b}$ . Then we have exact one of the following:

1.  $\exists \mathbf{x}' \in \mathbb{R}^{n+m} : \mathbf{x}' \geq 0, \mathbf{A}' \mathbf{x}' = \mathbf{b}$ .  
Let  $\mathbf{x}' = \begin{pmatrix} \mathbf{x} \\ \mathbf{x}'' \end{pmatrix}, \mathbf{x} \in \mathbb{R}^n, \mathbf{x}'' \in \mathbb{R}^m$ .  
So we have  $\mathbf{b} = \mathbf{A}' \mathbf{x}' = \mathbf{A}\mathbf{x} + \mathbf{I}\mathbf{x}'' = \mathbf{A}\mathbf{x} + \mathbf{x}'' \geq \mathbf{A}\mathbf{x}$ , thus  $\mathbf{x} \geq 0, \mathbf{A}\mathbf{x} \leq \mathbf{b}$
2.  $\exists \mathbf{y} \in \mathbb{R}^m : \mathbf{A}'^T \mathbf{y} \geq 0, \mathbf{b}^T \mathbf{y} < 0$ .  
So we have  $\begin{cases} \mathbf{A}^T \mathbf{y} \geq 0 \\ \mathbf{I}^T \mathbf{y} \geq 0 \end{cases}$ , thus  $\mathbf{y} \geq 0, \mathbf{A}^T \mathbf{y} \geq 0, \mathbf{b}^T \mathbf{y} < 0$

## FLLCP $\Rightarrow$ FLGV

Construct  $\mathbf{A}' = \begin{pmatrix} \mathbf{A} \\ -\mathbf{A} \end{pmatrix}, \mathbf{b}' = \begin{pmatrix} \mathbf{b} \\ -\mathbf{b} \end{pmatrix}$ . Use algorithm for FLLCP on  $\mathbf{A}'$  and  $\mathbf{b}'$ . Then we have exact one of the following:

1.  $\exists \mathbf{x} \in \mathbb{R}^n : \mathbf{x} \geq 0, \mathbf{A}' \mathbf{x} \leq \mathbf{b}'$ .  
So we have  $\begin{cases} \mathbf{A}\mathbf{x} \leq \mathbf{b} \\ -\mathbf{A}\mathbf{x} \leq -\mathbf{b} \end{cases}$ , thus  $\mathbf{x} \geq 0, \mathbf{A}\mathbf{x} = \mathbf{b}$
2.  $\exists \mathbf{y}' \in \mathbb{R}^{2m} : \mathbf{y}' \geq 0, \mathbf{A}'^T \mathbf{y}' \geq 0, \mathbf{b}'^T \mathbf{y}' < 0$ .  
Let  $\mathbf{y}' = \begin{pmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \end{pmatrix}$ , where  $\mathbf{y}_1 \in \mathbb{R}^m, \mathbf{y}_2 \in \mathbb{R}^m$ .  
So we have  $\mathbf{A}'^T \mathbf{y}' = \mathbf{A}^T \mathbf{y}_1 - \mathbf{A}^T \mathbf{y}_2 = \mathbf{A}^T (\mathbf{y}_1 - \mathbf{y}_2) \geq 0$ ,  $\mathbf{b}'^T \mathbf{y}' = \mathbf{b}^T \mathbf{y}_1 - \mathbf{b}^T \mathbf{y}_2 = \mathbf{b}^T (\mathbf{y}_1 - \mathbf{y}_2) < 0$ .  
Let  $\mathbf{y} \in \mathbb{R}^m, \mathbf{y} = \mathbf{y}_1 - \mathbf{y}_2$ , thus  $\mathbf{A}^T \mathbf{y} \geq 0, \mathbf{b}^T \mathbf{y} < 0$

## FLEV $\Rightarrow$ FLLPV

Construct  $\mathbf{A}' = \begin{pmatrix} \mathbf{A} \\ -\mathbf{I} \end{pmatrix}$  where  $\mathbf{I}$  is identity matrix,  $\mathbf{b}' = \begin{pmatrix} \mathbf{b} \\ \mathbf{0} \end{pmatrix}$ . Use algorithm for FLEV on  $\mathbf{A}'$  and  $\mathbf{b}'$ . Then we have exact one of the following:

$$1. \exists \mathbf{x} \in \mathbb{R}^n : \mathbf{A}'\mathbf{x} \leq \mathbf{b}'.$$

$$\text{So we have } \begin{cases} \mathbf{Ax} \leq \mathbf{b} \\ -\mathbf{Ix} \leq \mathbf{0} \end{cases}, \text{ thus } \mathbf{x} \geq \mathbf{0}, \mathbf{Ax} \leq \mathbf{b}$$

$$2. \exists \mathbf{y}' \in \mathbb{R}^{n+m} : \mathbf{y}' \geq \mathbf{0}, \mathbf{A}'^T \mathbf{y}' = \mathbf{0}, \mathbf{b}'^T \mathbf{y}' < 0.$$

$$\text{Let } \mathbf{y}' = \begin{pmatrix} \mathbf{y} \\ \mathbf{y}'' \end{pmatrix}, \mathbf{y} \in \mathbb{R}^m, \mathbf{y}'' \in \mathbb{R}^n.$$

$$\text{So we have } \mathbf{A}'^T \mathbf{y}' = \mathbf{A}^T \mathbf{y} - \mathbf{I} \mathbf{y}'' = \mathbf{0}, \mathbf{b}'^T \mathbf{y}' = \mathbf{b}^T \mathbf{y} < 0, \text{ thus } \mathbf{y} \geq \mathbf{0}, \mathbf{A}^T \mathbf{y} \geq \mathbf{0}, \mathbf{b}^T \mathbf{y} < 0$$

### **FLLPV $\Rightarrow$ FLEV**

Construct  $\mathbf{A}' = (\mathbf{A} \quad -\mathbf{A})$ . Use algorithm for FLLCP on  $\mathbf{A}'$  and  $\mathbf{b}$ . Then we have exact one of the following:

$$1. \exists \mathbf{x}' \in \mathbb{R}^{2n} : \mathbf{x}' \geq \mathbf{0}, \mathbf{A}'\mathbf{x}' \leq \mathbf{b}.$$

$$\text{Let } \mathbf{x}' = \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix}, \text{ where } \mathbf{x}_1 \in \mathbb{R}^n, \mathbf{x}_2 \in \mathbb{R}^n.$$

$$\text{So we have } \mathbf{A}'\mathbf{x}' = \mathbf{Ax}_1 - \mathbf{Ax}_2 = \mathbf{A}(\mathbf{x}_1 - \mathbf{x}_2) \leq \mathbf{b}.$$

$$\text{Let } \mathbf{x} \in \mathbb{R}^n, \mathbf{x} = \mathbf{x}_1 - \mathbf{x}_2, \text{ thus } \mathbf{Ax} \leq \mathbf{b}$$

$$2. \exists \mathbf{y} \in \mathbb{R}^m : \mathbf{y} \geq \mathbf{0}, \mathbf{A}'^T \mathbf{y} \geq \mathbf{0}, \mathbf{b}^T \mathbf{y} < 0.$$

$$\text{So we have } \begin{cases} \mathbf{A}^T \mathbf{y} \geq \mathbf{0} \\ -\mathbf{A}^T \mathbf{y} \geq \mathbf{0} \end{cases}, \text{ thus } \mathbf{A}^T \mathbf{y} = \mathbf{0}, \mathbf{b}^T \mathbf{y} < 0$$