

ECF 236A Shengze Ye HW4 205418959

problem 1

let us consider the primal and dual

$$\min C^T x$$

$$\text{s.t. } Ax \leq b$$

(primal)

$$\max -\lambda^T b$$

$$\text{s.t. } A^T \lambda + c = 0$$

$$\lambda \geq 0$$

(dual)

From weak duality, we have

$$-\lambda^{*T} b = d^* \leq p^* = C^T x^*$$

The dual can be written as

$$\min \lambda^T b$$

$$\text{s.t. } A^T \lambda + c = 0 \quad \text{and } \lambda \geq 0$$

Then we derive the dual of dual

$$L(\lambda, x, v) = \lambda^T b + x^T (A^T \lambda + c) - v \lambda$$

$$g(x, v) = \inf_{\lambda} L(\lambda, x, v)$$

$$= \begin{cases} C^T x & \text{if } (Ax + b = v) \\ -\infty & \text{otherwise} \end{cases}$$

\therefore The dual of dual is

$$\max C^T x$$

$$\text{s.t. } Ax + b = v$$

$$v \geq 0$$

$$\max C^T x$$

$$\text{s.t. } Ax + b \geq 0$$

if we take $x = -x$
 The dual of dual is the same of primal

$$\max -c^T x$$

$$\text{s.t. } Ax \leq b$$

From dual and dual of dual, we have

$$x^{*T} b = -d^* \geq -p^* = -c^T x^*$$

$$\therefore -d^* \geq -p^*$$

We still have $p^* \geq d^*$, we cannot conclude $p^* = d^*$

\therefore This proof of strong duality is flawed.

problem 2

maximize f_{ts}

$$\text{s.t. } f_{ij} \leq c_{ij} \quad \forall i, j \in E$$

$$\sum_{j: i \in E} f_{ji} - \sum_{i, k \in E} f_{ik} \leq 0 \quad \forall i \in V$$

$$f_{ij} \geq 0 \quad (i, j) \in E$$

(max-flow)

min $c^T d$

$$\text{s.t. } p_s - p_t \geq 1$$

$$d_{ij} \geq p_i - p_j$$

$$d_{ij} \in \{0, 1\}$$

$$p_i \in \{0, 1\}$$

(min-cut)

(a) Prove that the constraint matrix of the max-flow is TUM

As we did in the class, we should first write the max-flow LP in matrix format

$$\begin{aligned}
 \min \quad & -e_n^T f \\
 \text{s.t.} \quad & (I \ 0) f \leq c \\
 & M^T f \leq 0 \\
 & -f \leq 0
 \end{aligned}
 \quad f \in \mathbb{R}^n$$

M is the edge-vertex adjacency matrix

$$M: |E|+1 \times |V|$$

(1 for f_{es})

$$\Rightarrow \min -e_n^T f$$

$$\text{s.t.} \quad \underbrace{\begin{bmatrix} I_{|E|} & 0 \\ -I_{|E|+1} \\ M^T \end{bmatrix}}_A f \leq \begin{bmatrix} c \\ 0_{|E|+1+|V|} \end{bmatrix}$$

we have known that M is TUM

Now consider the sub-matrix of A

if all rows belong to M , then $\det = 0$ (M is TUM)

if all rows belong to the above part $\begin{pmatrix} I_{|E|} & 0 \\ -I_{|E|+1} \end{pmatrix}$

we know that $\det = 0$, and only contains $\{1, -1, 0\}$

if some rows belong to above and some come from M , we can expand along the column that contains the 1 / -1, and check the det of the small sub-matrix

\therefore matrix A is TUM, the constraint matrix is TUM

(b) Prove that the dual of the max-flow LP gives the same solution

The dual LP:

$$\min c^T d$$

$$\text{s.t. } p_s - p_t \geq 1$$

$$d_{ij} \geq p_i - p_j$$

$$d_{ij} \geq 0 \quad p_i \geq 0$$

(1)

Integer LP:

$$\min c^T d$$

$$\text{s.t. } p_s - p_t \geq 1$$

$$d_{ij} \geq p_i - p_j$$

$$d_{ij} \in \{0, 1\}, \quad p_i \in \{0, 1\}$$

(2)

The dual LP could be written as:

$$\min c^T d$$

$$\text{s.t. } \begin{bmatrix} D_{1 \times |E|} & M^T \\ I_{|E|} & \\ -I_{|E|+|V|} & \end{bmatrix} \underbrace{\begin{bmatrix} d \\ p \end{bmatrix}}_x \geq \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_b$$

consider an extreme point in (1). let J be the set of active constraints at the extreme point.

A_J is non-singular submatrix of A , A_J is TUM from the class we know that

if A_J is non-singular submatrix of A

A_J^{-1} is an integral matrix

$$A_J \cdot x_J = b'$$

$$x_J = A_J^{-1} \cdot b'$$

b' has only single non-zero element (1)

\therefore The extreme point x_j is integral

The optimal value obtained from (1) and (2) should be same,

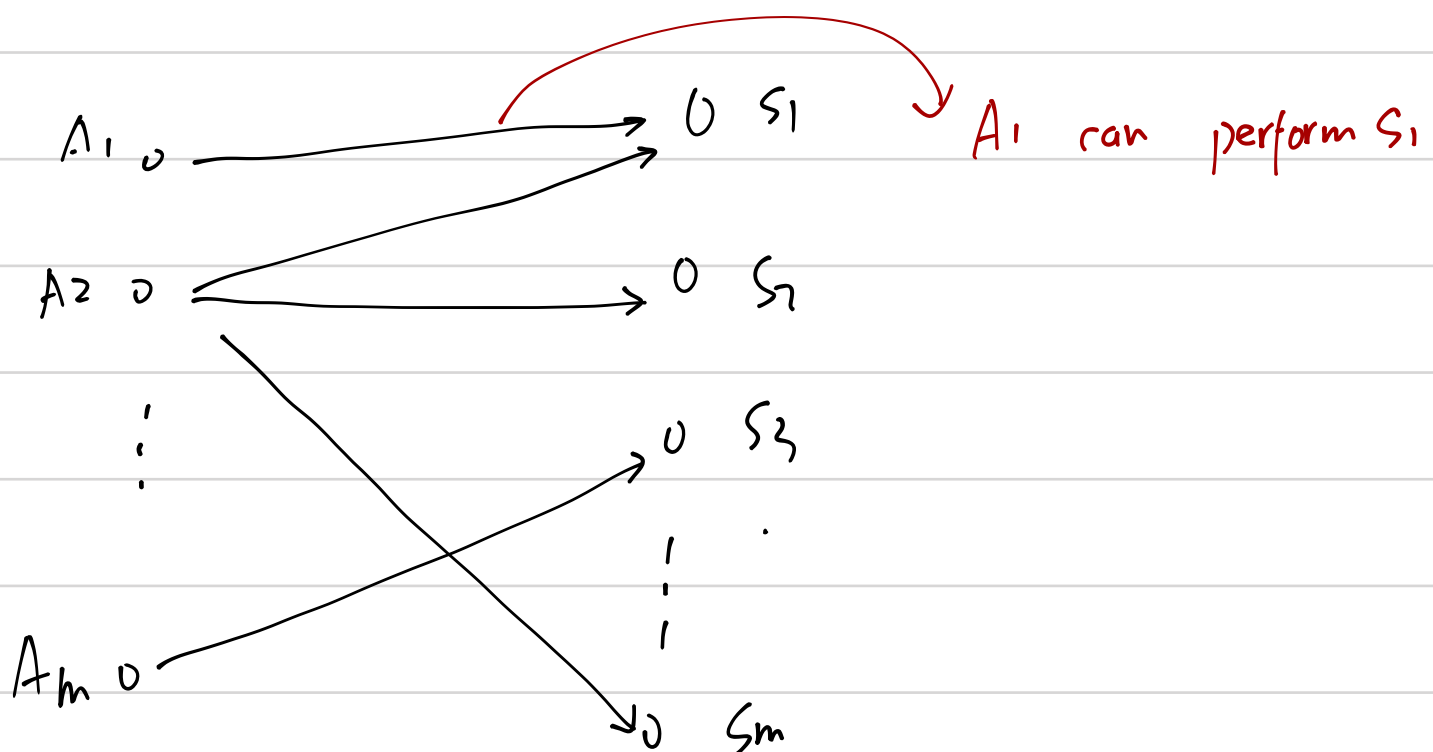
\therefore The dual of max-flow LP gives the same solution as its ILP.

Problem 3.

Surgeries $S = \{s_1, s_2, \dots, s_n\}$

Surgeon $A = \{A_1, A_2, \dots, A_m\}$

Each candidate can perform several surgeries, this relation can be illustrated by a graph



A connection means that surgeon A_i could perform S_j

we can use a vector to denote that

$$p_i = \begin{bmatrix} 0 \\ 1 \\ 0 \\ i \\ 1 \\ 0 \end{bmatrix} \begin{array}{l} \rightarrow 2\text{th surgery} \\ \vdots \\ \rightarrow n-1\text{th surgery} \\ \rightarrow n\text{th} \end{array}$$

\rightarrow means that i th surgeon can perform 2th and $n-1\text{th}$ surgery

Once we have the set S and A ,

\Rightarrow we know p_1, p_2, \dots, p_m (constant)

Then we introduce the salary $\Rightarrow S_i$ be the salary required by surgeon number i

and x_i to denote whether we select surgeon number i

$$\therefore x_i \in \{0, 1\}$$

\therefore We can write the ILP

$$\min \sum_{i=1}^m x_i \cdot S_i$$

s.t

$$x_1 \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} + x_2 \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \dots + x_m \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} \geq \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

each $p_i \in \mathbb{R}^n$

(handle all the surgeries)

$$S_i \geq 0$$

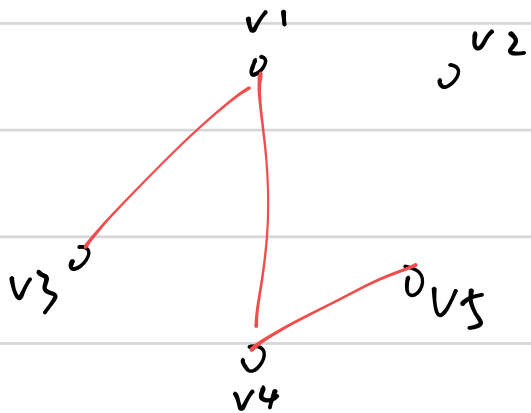
$$i = 1, 2, \dots, m$$

$$x_i \in \{0, 1\}$$

Problem 4

a) Write an ILP associated with the question

The problem can be described as a graph. we use the node v_i to denote the player i . we use the connection e_{ij} to note that match-ups between i and j .



use x_e to denote whether we use the match-up

let $N(v) \subseteq E$ be the set of edges that are adjacent to v

\therefore we can formulate the ILP

$$\begin{aligned} \max \quad & \sum_{e \in E} x_e \\ \text{s.t.} \quad & \sum_{e \in N(v)} x_e \leq 1 \quad \forall v \\ & x_e \in \{0, 1\} \end{aligned}$$

We can express it using a matrix (TUM)

let $M = \begin{matrix} & v_1 & v_2 & \dots & v_m \\ \begin{matrix} e_1 \\ e_2 \\ \vdots \\ e_n \end{matrix} & \begin{bmatrix} 1 & & & 1 \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \end{matrix}$

$|E| \cdot |V|$

each match-up will set two "1" to the corresponding player

$\rightarrow v_1 \xrightarrow{e_1} v_m$

\therefore The ILP can be written as

$$\max \sum x$$

$$\text{s.t. } \sum_{|v| \times |z|} M^T \cdot x \leq 1$$

$$x_i \in \{0, 1\}$$

(b) Coach P want to arrange matches on three days,

$\therefore \Rightarrow$ each player can play in up to 3 different matches

Write the ILP.

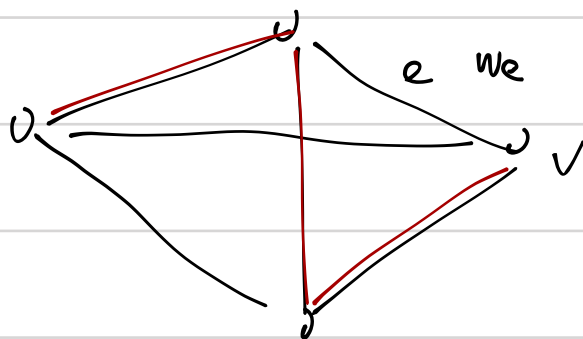
$$\therefore \Rightarrow \max \sum x$$

$$\text{s.t. } M^T \cdot x \leq 3$$

$$x_i \in \{0, 1\} \quad \text{for all } i$$

Problem 5

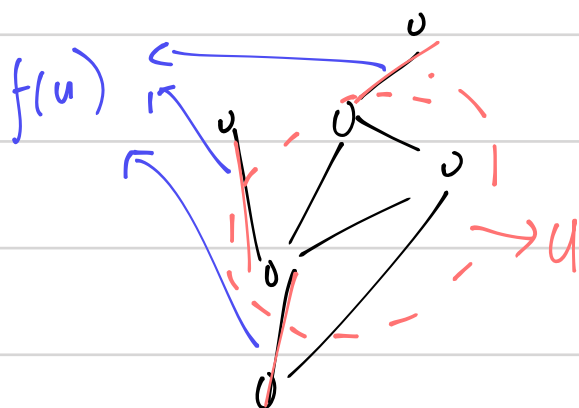
We can model the problem using an undirected graph



The nodes denote the houses, and we have edges between nodes, for each edge, we could set a penalty term, represents the distance between two nodes

We want to find a subset of E to access all the nodes with minimum sum of penalty, and do not go back to the nodes that visited.

define $f(u) = \{ (i, j) \in E \mid i \in U, j \notin U \}$



\therefore The problem can be formulated as

$$\min \sum_{e \in E} W_e x_e$$

(x_e : whether we select this edge or not)

$$\text{s.t. } \sum_{e \in f(v)} x_e = 2 \quad \forall v \in V$$

$$\textcircled{2} \sum_{e \in f(u)} x_e \geq 2 \quad \forall U \subset V \quad 2 \leq |U| \leq |V|-1$$

$$x_e \in \{0, 1\}$$

① because each node in the solution has degree 2

② because we need to exclude the situation of multiple disjoint cycles

Problem 6

$$\begin{aligned} \max \quad & \sum f_i \\ \text{s.t.} \quad & \sum_{i: e \in p_i} f_i \leq c_e \quad \forall \text{ edge } e \\ & f_i \geq 0 \end{aligned}$$

$P = \{p_i\}$ is the set of all paths from s to t .
and f_i is the flow associated with the i th path.

① prove that the LP is equivalent to the max-flow

The max-flow LP is

$$\begin{aligned} \max \quad & f_{ts} \\ \text{s.t.} \quad & f_{ij} \leq c_{ij} \quad \forall (i,j) \in E \\ & \sum_{j: (i,j) \in E} f_{ji} - \sum_{k: (i,k) \in E} f_{ik} = 0 \quad \forall i \in V \\ & f_{ij} \geq 0 \quad (i,j) \in E \end{aligned} \quad \text{①}$$

