

## 7. Separating Hyperplane Theorems II

Daisuke Oyama

Mathematics II

May 8, 2024

# Farkas' Lemma

## Proposition 7.16 (Farkas' Lemma)

Let  $A \in \mathbb{R}^{M \times N}$  and  $b \in \mathbb{R}^N$ .

The following conditions are equivalent:

1. There exists  $x \in \mathbb{R}^M$  such that  $A^T x = b$  and  $x \geq 0$ .
2. For any  $y \in \mathbb{R}^N$ , if  $Ay \geq 0$ , then  $b^T y \geq 0$ .

For proof, we will use the following:

## Lemma 7.17

$\{A^T x \in \mathbb{R}^N \mid x \in \mathbb{R}_+^M\}$  is a closed set.

## Proof of Farkas' Lemma

▶ (1)  $\Rightarrow$  (2): Immediate.

▶ (2)  $\Rightarrow$  (1):

Suppose that (1) does not hold.

Let  $K = \{A^T x \in \mathbb{R}^N \mid x \in \mathbb{R}_+^M\}$ . Then  $b \notin K$ .

▶  $K$  is convex, and by Lemma 7.17 is closed.

▶ Then by the Separating Hyperplane Theorem, there exist  $y \in \mathbb{R}^N$  with  $y \neq 0$  and  $c \in \mathbb{R}$  such that

$$y^T b < c \leq y^T z \text{ for all } z \in K,$$

and therefore,  $y^T b < \inf_{z \in K} y^T z$ .

▶ Since  $K$  is a cone, it follows that  $\inf_{z \in K} y^T z = 0$ .  
( $\rightarrow$  Homework)

▶ Thus we have  $y^T b < 0$ , and  $y^T A^T x \geq 0$  for all  $x \geq 0$ , which implies that  $y^T A^T \geq 0^T$ .

## Proof of Lemma 7.17

Show that  $K = \{A^T x \in \mathbb{R}^N \mid x \in \mathbb{R}_+^M\}$  is closed.

- ▶ Denote the column vectors in  $A^T$  by  $a^1, \dots, a^M$ , so that  $K = \text{Cone}\{a^1, \dots, a^M\}$ .
- ▶ Let  $\{z^m\}$  be a sequence in  $K$ , and suppose that  $z^m \rightarrow \bar{z}$ . We want to show that  $\bar{z} \in K$ .
- ▶ By Carathéodory's Theorem, for each  $m$ ,  $z^m$  is written as a conic combination of a linearly independent subset of  $\{a^1, \dots, a^M\}$ .
- ▶ Since there are finitely many such subsets, there is a linearly independent subset  $\{a^{i_1}, \dots, a^{i_L}\}$  such that infinitely many elements of  $\{z^m\}$  are written as its conic combinations.
- ▶ Denote  $B = (a^{i_1} \ \dots \ a^{i_L}) \in \mathbb{R}^{N \times L}$ , and denote the corresponding subsequence again by  $\{z^m\}$ .

- ▶ Denote  $z^m = B\lambda^m$ , where  $\lambda^m \in \mathbb{R}_+^L$ .
- ▶ We have  $B^T z^m = B^T B\lambda^m$ , where  $B^T B \in \mathbb{R}^{L \times L}$  is non-singular:
  - ▶ Let  $B^T Bx = 0$ .
  - ▶ Then  $x^T B^T Bx = 0$ , where  $x^T B^T Bx = \|Bx\|^2$ .
  - ▶ Therefore,  $x^T B^T Bx = 0$  if and only if  $Bx = 0$ .
  - ▶ Since the columns of  $B$  are linearly independent, this holds if and only if  $x = 0$ .
- ▶ Therefore, we have  $\lambda^m = (B^T B)^{-1} B^T z^m$ .
- ▶ By the continuity of  $(B^T B)^{-1} B^T z$  in  $z$ ,  $\lambda^m$  converges to  $\bar{\lambda} = (B^T B)^{-1} B^T \bar{z}$ , where  $\bar{\lambda} \in \mathbb{R}_+^L$ .
- ▶ Thus, by the continuity of  $B\lambda$  in  $\lambda$ , we have  $\bar{z} = \lim_{m \rightarrow \infty} B\lambda^m = B\bar{\lambda}$ , so that  $\bar{z} \in K$ .

## Variants of Farkas' Lemma

### Proposition 7.18 (Farkas' Lemma: Inequality version)

Let  $A \in \mathbb{R}^{M \times N}$  and  $b \in \mathbb{R}^M$ .

The following conditions are equivalent:

1. There exists  $x \in \mathbb{R}^N$  such that  $Ax \leq b$  and  $x \geq 0$ .
2. For any  $y \in \mathbb{R}^M$ , if  $y \geq 0$  and  $A^T y \geq 0$ , then  $b^T y \geq 0$ .

# Proof

- ▶ Condition (1) is equivalent to:

There exist  $x \in \mathbb{R}^M$  and  $z \in \mathbb{R}^N$  such that  $x \geq 0$ ,  $z \geq 0$ , and  $Ax + z = b$ ,

or  $(A^T \quad I) \begin{pmatrix} x \\ z \end{pmatrix} = b$ .

- ▶ By Farkas' Lemma, this is equivalent to:

For any  $y \in \mathbb{R}^N$ , if  $\begin{pmatrix} A \\ I \end{pmatrix} y \geq 0$ , then  $b^T y \geq 0$ ,

or, if  $y \geq 0$  and  $Ay \geq 0$ , then  $b^T y \geq 0$  (condition (2)).

# Linear Programming

Let  $A \in \mathbb{R}^{K \times N}$ ,  $f \in \mathbb{R}^N$ ,  $c \in \mathbb{R}^K$ .

Primal problem:

$$\begin{aligned} \text{(P)} \quad & \max_{x \in \mathbb{R}^N} f^T x \\ & \text{s. t. } Ax \leq c \\ & \quad x \geq 0. \end{aligned}$$

Dual problem:

$$\begin{aligned} \text{(D)} \quad & \min_{\lambda \in \mathbb{R}^K} c^T \lambda \\ & \text{s. t. } A^T \lambda \geq f \\ & \quad \lambda \geq 0. \end{aligned}$$

The Lagrangians for the two problems coincide (the nonnegativity constraints aside):

$$L(x, \lambda) = f^T x - \lambda^T (Ax - c) = c^T \lambda - x^T (A^T \lambda - f).$$



# Weak Duality

## Proposition 7.19

If  $x \in \mathbb{R}^N$  and  $\lambda \in \mathbb{R}^K$  are feasible for (P) and (D), respectively, then  $f^T x \leq c^T \lambda$ .

## Proof

► If  $x \in \mathbb{R}^N$  and  $\lambda \in \mathbb{R}^K$  are feasible for (P) and (D), then

$$f^T x \leq (A^T \lambda)^T x = \lambda^T (Ax) \leq \lambda^T c.$$

Therefore, if  $\bar{x} \in \mathbb{R}^N$  and  $\bar{\lambda} \in \mathbb{R}^K$  are feasible and if  $f^T \bar{x} = c^T \bar{\lambda}$ , then  $\bar{x}$  and  $\bar{\lambda}$  are solutions to (P) and (D), respectively.

# Strong Duality

## Proposition 7.20

*Suppose that both (P) and (D) are feasible.  
Then both (P) and (D) have solutions, and*

$$\max\{f^T x \mid Ax \leq c, x \geq 0\} = \min\{c^T \lambda \mid A^T \lambda \geq f, \lambda \geq 0\}.$$

## Proof

- ▶ Suppose that (P) and (D) are feasible.

We want to show that there exist  $x \in \mathbb{R}^N$  and  $\lambda \in \mathbb{R}^K$  such that  $Ax \leq c$ ,  $A^T\lambda \geq f$ ,  $f^T x \geq c^T \lambda$ ,  $x \geq 0$ , and  $\lambda \geq 0$ , or

$$\begin{pmatrix} A & O \\ O & -A^T \\ -f^T & c^T \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix} \leq \begin{pmatrix} c \\ -f \\ 0 \end{pmatrix}, \quad x \geq 0, \quad \lambda \geq 0.$$

- ▶ By Farkas' Lemma (inequality version; Proposition 7.18), this is equivalent to the condition that for all  $p \in \mathbb{R}^K$ ,  $q \in \mathbb{R}^N$ , and  $r \in \mathbb{R}$ ,

$$\begin{aligned} & (p^T \quad q^T \quad r) \begin{pmatrix} A & O \\ O & -A^T \\ -f^T & c^T \end{pmatrix} \geq 0, \quad p \geq 0, \quad q \geq 0, \quad r \geq 0 \\ \Rightarrow & (p^T \quad q^T \quad r) \begin{pmatrix} c \\ -f \\ 0 \end{pmatrix} \geq 0. \end{aligned}$$

- ▶ That is,

$$(1) \quad A^T p \geq rf, \quad Aq \leq rc, \quad p \geq 0, \quad q \geq 0, \quad r \geq 0$$

implies

$$(2) \quad c^T p - f^T q \geq 0.$$

We want to show that this holds whenever (P) and (D) are feasible.

- ▶ For  $r > 0$ , (1) implies that  $q/r$  and  $p/r$  are feasible solutions to (P) and (D), so that we have  $c^T p - f^T q = r[c^T(p/r) - f^T(q/r)] \geq 0$  by Weak Duality.
- ▶ For  $r = 0$ , let  $x$  and  $\lambda$  be feasible solutions to (P) and (D). From (1), we have

$$c^T p - f^T q \geq x^T A^T p - \lambda^T Aq \geq 0.$$

# Strong Duality

## Proposition 7.21

1. *Suppose that (D) has a solution.*

*Then (P) has a solution, and*

$$\max\{f^T x \mid Ax \leq c, x \geq 0\} = \min\{c^T \lambda \mid A^T \lambda \geq f, \lambda \geq 0\}.$$

2. *Suppose that (P) has a solution.*

*Then (D) has a solution, and*

$$\max\{f^T x \mid Ax \leq c, x \geq 0\} = \min\{c^T \lambda \mid A^T \lambda \geq f, \lambda \geq 0\}.$$

## Proof

- ▶ Suppose that (D) has a solution.

In light of Proposition 7.20, it suffices to show that (P) has a feasible solution.

- ▶ Let  $\lambda^* \in \mathbb{R}^K$  be a solution to (D).

To apply Farkas' Lemma (Proposition 7.18), let  $z \in \mathbb{R}^K$  be such that  $A^T z \geq 0$  and  $z \geq 0$ .

- ▶ Then  $\lambda^* + z \geq 0$ , and  $A^T(\lambda^* + z) = A^T\lambda^* + A^Tz \geq f$ , which means that  $\lambda^* + z$  is feasible in (D).
- ▶ Therefore, by the optimality of  $\lambda^*$ , we have  $0 \leq c^T(\lambda^* + z) - c^T\lambda^* = c^Tz$ .
- ▶ By Proposition 7.18, there exists  $x \in \mathbb{R}^N$  such that  $Ax \leq c$  and  $x \geq 0$ .

# Variants of Farkas' Lemma

## Proposition 7.22 (Gale's Theorem)

Let  $A \in \mathbb{R}^{M \times N}$  and  $b \in \mathbb{R}^M$ .

The following conditions are equivalent:

1. There exists  $x \in \mathbb{R}^N$  such that  $Ax \leq b$ .
2. For any  $y \in \mathbb{R}^M$ , if  $y \geq 0$  and  $A^T y = 0$ , then  $b^T y \geq 0$ .

## Proof

- ▶ Condition (1) is equivalent to:

There exist  $z^1 \in \mathbb{R}^M$  and  $z^2 \in \mathbb{R}^M$  such that  $z^1 \geq 0$ ,  $z^2 \geq 0$ ,  
and  $A^T(z^1 - z^2) \leq b$ ,

$$\text{or } (A^T \quad -A^T) \begin{pmatrix} z^1 \\ z^2 \end{pmatrix} \leq b.$$

- ▶ By Farkas' Lemma (inequality version; Proposition 7.18),  
this is equivalent to:

For any  $y \in \mathbb{R}^N$ , if  $y \geq 0$  and  $\begin{pmatrix} A \\ -A \end{pmatrix} y \geq 0$ , then  $b^T y \geq 0$ ,

or, if  $y \geq 0$  and  $Ay = 0$ , then  $b^T y \geq 0$  (condition (2)).



# Variants of Farkas' Lemma

## Proposition 7.23 (Gordan's Theorem)

Let  $A \in \mathbb{R}^{M \times N}$ .

*The following conditions are equivalent:*

1. *There exists  $x \in \mathbb{R}^M$  such that  $A^T x \gg 0$ .*
2. *For any  $y \in \mathbb{R}^N$ , if  $y \geq 0$  and  $Ay = 0$ , then  $y = 0$ .*

## Proof

- ▶ Condition (1) is equivalent to:

There exists  $x \in \mathbb{R}^M$  such that  $-A^T x \leq -\mathbf{1}$ .

- ▶ By Gale's Theorem (Proposition 7.22), this is equivalent to:

For any  $y \in \mathbb{R}^N$ , if  $y \geq 0$  and  $(-A)y = 0$ , then  $(-\mathbf{1}^T)y \geq 0$ ,  
or  $y \geq 0$  and  $Ay = 0$ , then  $y = 0$  (condition (2)).

## Variants of Farkas' Lemma

### Proposition 7.24 (Ville/von Neumann-Morgenstern I)

Let  $A \in \mathbb{R}^{M \times N}$ .

*The following conditions are equivalent:*

1. *There exists  $x \in \mathbb{R}^M$  such that  $A^T x \gg 0$  and  $x \gg 0$ .*
2. *For any  $y \in \mathbb{R}^N$ , if  $y \geq 0$  and  $Ay \leq 0$ , then  $y = 0$ .*

## Variants of Farkas' Lemma

- ▶ In fact,  
“there exists  $x \in \mathbb{R}^M$  such that  $A^T x \gg 0$  and  $x \gg 0$ ”  
is equivalent to  
“there exists  $x \in \mathbb{R}^M$  such that  $A^T x \gg 0$  and  $x \geq 0$ ”.
- ▶ Given an  $x \geq 0$  in the latter, consider  $x + \varepsilon \mathbf{1}$  for sufficiently small  $\varepsilon > 0$ .

### Proposition 7.25 (Ville/von Neumann-Morgenstern II)

Let  $A \in \mathbb{R}^{M \times N}$ .

The following conditions are equivalent:

1. There exists  $x \in \mathbb{R}^M$  such that  $A^T x \gg 0$  and  $x \geq 0$ .
2. For any  $y \in \mathbb{R}^N$ , if  $y \geq 0$  and  $Ay \leq 0$ , then  $y = 0$ .

## Proof of Proposition 7.24

- ▶ Condition (1) is equivalent to:

There exists  $x \in \mathbb{R}^M$  such that  $\begin{pmatrix} A^T \\ I \end{pmatrix} x \gg 0$ .

- ▶ By Gordan's Theorem (Proposition 7.23), this is equivalent to:

For any  $y \in \mathbb{R}^N$  and  $z \in \mathbb{R}^M$ ,  
if  $y \geq 0$ ,  $z \geq 0$ , and  $(A \quad I) \begin{pmatrix} y \\ z \end{pmatrix} = 0$ , then  $\begin{pmatrix} y \\ z \end{pmatrix} = 0$ .

- ▶ This is equivalent to:

For any  $y \in \mathbb{R}^N$ ,  
if  $y \geq 0$  and  $Ay \leq 0$ , then  $y = 0$  (condition (2)).

# Variants of Farkas' Lemma

## Proposition 7.26

Let  $A \in \mathbb{R}^{M \times N}$ .

*The following conditions are equivalent:*

1. *There exists  $x \in \mathbb{R}^M$  such that  $A^T x \leq 0$  and  $x \gg 0$ .*
2. *For any  $y \in \mathbb{R}^N$ , if  $y \geq 0$  and  $Ay \geq 0$ , then  $Ay = 0$ .*

## Proof

- ▶ Condition (1) is equivalent to:

There exists  $x \in \mathbb{R}^M$  such that  $\begin{pmatrix} A^T \\ -I \end{pmatrix} x \leq \begin{pmatrix} 0 \\ -\mathbf{1} \end{pmatrix}$ .

- ▶ By Gale's Theorem (Proposition 7.22), this is equivalent to:

For any  $y \in \mathbb{R}^N$  and  $z \in \mathbb{R}^M$ ,

if  $y \geq 0$ ,  $z \geq 0$ , and  $(A \quad -I) \begin{pmatrix} y \\ z \end{pmatrix} = 0$ , then

$$(0 \quad -\mathbf{1}^T) \begin{pmatrix} y \\ z \end{pmatrix} \geq 0.$$

- ▶ This is equivalent to:

For any  $y \in \mathbb{R}^N$ ,

if  $y \geq 0$  and  $Ay \geq 0$ , then  $Ay = 0$  (condition (2)).

# Variants of Farkas' Lemma

## Proposition 7.27 (Stiemke's Lemma)

Let  $A \in \mathbb{R}^{M \times N}$ .

*The following conditions are equivalent:*

1. *There exists  $x \in \mathbb{R}^M$  such that  $A^T x = 0$  and  $x \gg 0$ .*
2. *For any  $y \in \mathbb{R}^N$ , if  $Ay \geq 0$ , then  $Ay = 0$ .*



## Proof

- ▶ Condition (1) is equivalent to:

There exists  $x \in \mathbb{R}^M$  such that  $x \gg 0$  and  $\begin{pmatrix} A^T \\ -A^T \end{pmatrix} x \leq 0$ .

- ▶ By Proposition 7.26, this is equivalent to:

For any  $y \in \mathbb{R}^N$  and  $z \in \mathbb{R}^M$ ,

if  $y \geq 0$ ,  $z \geq 0$ , and  $(A \quad -A) \begin{pmatrix} y \\ z \end{pmatrix} \geq 0$ , then

$$(A \quad -A) \begin{pmatrix} y \\ z \end{pmatrix} = 0.$$

- ▶ This is equivalent to:

For any  $y \in \mathbb{R}^N$ , if  $Ay \geq 0$ , then  $Ay = 0$  (condition (2)).

# Variants of Farkas' Lemma

## Proposition 7.28 (Motzkin's Theorem)

Let  $B \in \mathbb{R}^{M \times N}$ ,  $C \in \mathbb{R}^{M \times K}$ ,  $D \in \mathbb{R}^{M \times L}$ .

The following conditions are equivalent:

1. There exists no  $x \in \mathbb{R}^M$  such that  $B^T x \gg 0$ ,  $C^T x \geq 0$ , and  $D^T x = 0$ .
2. There exist  $y_1 \in \mathbb{R}^N$ ,  $y_2 \in \mathbb{R}^K$ , and  $y_3 \in \mathbb{R}^L$  such that  $By_1 + Cy_2 + Dy_3 = 0$ ,  $y_1 \geq 0$ ,  $y_1 \neq 0$ , and  $y_2 \geq 0$ .

- ▶ Proved using Farkas' Lemma.
- ▶ Proposition 7.23 (Gordan's Theorem), Propositions 7.24-7.25 (Ville's Theorem), Proposition 7.26, and Proposition 7.27 (Stiemke's Lemma) are all special cases of this theorem.

# Efficient Production under Linear Technology

- ▶ For the production set  $Y \subset \mathbb{R}^N$ ,  $y \in Y$  is *efficient* if there is no  $y' \in Y$  such that  $y' \geq y$  and  $y' \neq y$ .

## Proposition 7.29

Let  $Y = \{y \in \mathbb{R}^N \mid Ay \leq b\}$  for some  $A \in \mathbb{R}^{M \times N}$  and  $b \in \mathbb{R}^M$ . Then  $\bar{y} \in Y$  is efficient if and only if there exists  $p \gg 0$  such that

$$p \cdot \bar{y} \geq p \cdot y \text{ for all } y \in Y.$$

- ▶ The “if” part holds for general set  $Y$ .

## Proof

- ▶ The “if” part:

If  $\bar{y}$  is not efficient, i.e.,  $y' - \bar{y} \geq 0, \neq 0$  for some  $y' \in Y$ , then for any  $p \gg 0$ , we have  $(y' - \bar{y})p > 0$  or  $y'p > \bar{y}^*p$ .

- ▶ The “only if” part:

Suppose that  $\bar{y} \in Y$  is efficient.

- ▶ Write  $A = \begin{pmatrix} A^1 \\ A^2 \end{pmatrix}$  and  $b = \begin{pmatrix} b^1 \\ b^2 \end{pmatrix}$  such that

$$A^1\bar{y} = b^1, \quad A^2\bar{y} \ll b^2,$$

where  $A^k \in \mathbb{R}^{M_k \times N}$ ,  $b^k \in \mathbb{R}^{M_k}$ ,  $k = 1, 2$ , and  $M_1 + M_2 = M$ .

- ▶ By the efficiency of  $\bar{y}$ ,  $M_1 \geq 1$ .

- ▶ By the efficiency of  $\bar{y}$ , there exists no  $z \in \mathbb{R}^N$  such that  $A^1 z \leq 0$ ,  $z \geq 0$ ,  $z \neq 0$ .

If there exists such  $z$ , then  $A(\bar{y} + \varepsilon z) \leq b$  for sufficiently small  $\varepsilon > 0$ , where  $\bar{y} + \varepsilon z \not\geq \bar{y}$ .

- ▶ By Proposition 7.25 (Ville's Theorem), there exists  $x \in \mathbb{R}^{M_1}$  such that  $(A^1)^T x \gg 0$  and  $x \geq 0$ .

Let  $p = (A^1)^T x (\gg 0)$ .

- ▶ Then for any  $y \in Y$  (where  $A^1 y \leq b^1$ ), we have

$$p \cdot \bar{y} = x \cdot A^1 \bar{y} = x \cdot b^1,$$

$$p \cdot y = x \cdot A^1 y \leq x \cdot b^1,$$

as desired.

# Strict Dominance and Never Best Response

Consider a two-player normal form game:

- ▶  $S_1 = \{1, \dots, M\}$ : set of pure strategies of player 1 ( $M \geq 2$ )  
 $S_2 = \{1, \dots, N\}$ : set of pure strategies of player 2 ( $N \geq 2$ )
- ▶  $\Delta(S_1) = \{x \in \mathbb{R}_+^M \mid x_1 + \dots + x_M = 1\}$ :  
set of mixed strategies of player 1  
 $\Delta(S_2) = \{y \in \mathbb{R}_+^N \mid y_1 + \dots + y_N = 1\}$ :  
set of mixed strategies of player 2
- ▶ From player 1's point of view,  $\Delta(S_2)$  is interpreted as the set of 1's *beliefs* over 2's strategies.
- ▶ Pure strategy  $m \in S_1$  is identified with  $e_m \in \Delta(S_1)$ , the  $m$ th unit vector of  $\mathbb{R}^M$ .

- ▶ Payoff matrix for player 1:

$$U = \begin{pmatrix} u_{11} & \cdots & u_{1N} \\ \vdots & \ddots & \vdots \\ u_{M1} & \cdots & u_{MN} \end{pmatrix} \in \mathbb{R}^{M \times N}$$

(We only consider the incentives of player 1.)

- ▶  $e_m^T U y \cdots$  payoff from  $m \in S_1$  against  $y \in \Delta(S_2)$
- ▶  $x^T U y \cdots$  payoff from  $x \in \Delta(S_1)$  against  $y \in \Delta(S_2)$

- ▶  $m \in S_1$  is a best response to  $y \in \Delta(S_2)$  if  $e_m^T U y \geq e_\ell^T U y$  for all  $\ell \in S_1$ .
- ▶  $m \in S_1$  is a *never best response* if it is not a best response to any  $y \in \Delta(S_2)$ .
- ▶  $m \in S_1$  is *strictly dominated* if there exists  $x \in \Delta(S_1)$  such that  $e_m^T U e_n < x^T U e_n$  for all  $n \in S_2$ .



## Proposition 7.30

*In a two-player normal form game,  $m \in S_1$  is a never best response if and only if it is strictly dominated.*

- ▶ The result extends straightforwardly to (finite) games with more than two players if best response is defined with respect to *correlated* beliefs over opponents' strategies.

# Proof

- ▶ Let

$$\tilde{U} = \begin{pmatrix} u_{11} - u_{m1} & \cdots & u_{1N} - u_{mN} \\ \vdots & \ddots & \vdots \\ u_{M1} - u_{m1} & \cdots & u_{MN} - u_{mN} \end{pmatrix}.$$

- ▶  $m \in S_1$  is a never best response
  - $\iff$  there exists no  $y \geq 0$ ,  $y \neq 0$ , such that  $\tilde{U}y \leq 0$
  - $\iff$  if  $y \geq 0$  and  $\tilde{U}y \leq 0$ , then  $y = 0$ .
- ▶  $m \in S_1$  is strictly dominated
  - $\iff$  there exists  $x \geq 0$ ,  $x \neq 0$ , such that  $x^T \tilde{U} \gg 0$ .
- ▶ By Ville's Theorem (Proposition 7.25), these are equivalent.

# Weak Dominance and Never Best Response

- ▶  $m \in S_1$  is *weakly dominated* if there exists  $x \in \Delta(S_1)$  such that
  - ▶  $e_m^T U e_n \leq x^T U e_n$  for all  $n \in S_2$ , and
  - ▶  $e_m^T U e_n < x^T U e_n$  for some  $n \in S_2$ .

## Proposition 7.31

*In a two-player normal form game,  $m \in S_1$  is a best response to some totally mixed strategy  $y \in \Delta(S_2)$  if and only if it is not weakly dominated.*

# Proof

- ▶ Again let

$$\tilde{U} = \begin{pmatrix} u_{11} - u_{m1} & \cdots & u_{1N} - u_{mN} \\ \vdots & \ddots & \vdots \\ u_{M1} - u_{m1} & \cdots & u_{MN} - u_{mN} \end{pmatrix}.$$

- ▶  $m \in S_1$  is a best response to some totally mixed strategy  
 $\iff$  there exists  $y \gg 0$  such that  $\tilde{U}y \leq 0$ .
- ▶  $m \in S_1$  is not weakly dominated  
 $\iff$  there exists no  $x \geq 0$ ,  $x \neq 0$ , such that  $x^T \tilde{U} \not\geq 0$   
 $\iff$  if  $x \geq 0$  and  $x^T \tilde{U} \geq 0$ , then  $x^T \tilde{U} = 0$ .
- ▶ By Proposition 7.26, these are equivalent.