

# Geometry of Equilibria in Bimatrix Games

Bernhard von Stengel

reading material:  
Chapter 9 of “Game Theory Basics”

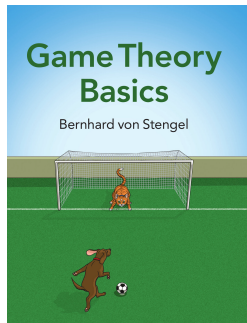
# Plan

- recall best-response condition
- upper envelope with 2 and 3 goalposts
- **labels** in best-response diagrams
- equilibria = completely labeled strategy pairs
- the Lemke–Howson algorithm
- labeled polytopes
- complementary pivoting
  - handling degeneracy
  - efficient exact arithmetic

## Some Reading Material

B. von Stengel (2021), *Game Theory Basics*.  
Cambridge University Press.

B. von Stengel (2021), Finding Nash equilibria  
of two-player games. arXiv:2102.04580.



L. S. Shapley (1974), A note on the Lemke-Howson algorithm.  
*Mathematical Programming Study 1: Pivoting and Extensions*,  
175–189.

Bimatrix Games,  
Best-Response Condition

## Nash equilibria of bimatrix games

$$A = \begin{array}{|c|c|} \hline 0 & 6 \\ \hline 2 & 5 \\ \hline 3 & 3 \\ \hline \end{array}$$
$$B = \begin{array}{|c|c|} \hline 2 & 1 \\ \hline 1 & 3 \\ \hline 4 & 3 \\ \hline \end{array}$$

Nash equilibrium =

pair of strategies  $x$ ,  $y$  with

$x$  best response to  $y$  and

$y$  best response to  $x$ .

## Mixed equilibria

$$A = \begin{bmatrix} 0 & 6 \\ 2 & 5 \\ 3 & 3 \end{bmatrix}$$

$$B = \begin{bmatrix} 2 & 1 \\ 1 & 3 \\ 4 & 3 \end{bmatrix}$$

$$x = \begin{bmatrix} 2/3 \\ 1/3 \\ 0 \end{bmatrix}$$

$$x^T B = \begin{bmatrix} 5/3 & 5/3 \end{bmatrix}$$

$$A y = \begin{bmatrix} 4 \\ 4 \\ 3 \end{bmatrix}$$

$$y^T = \begin{bmatrix} 1/3 & 2/3 \end{bmatrix}$$

only **pure best responses** can have probability  $> 0$

## Best-response condition

**Theorem** Given:  $m \times n$  bimatrix game  $(A, B)$ .

Let  $x$  be a mixed strategy of **player I** and  
let  $y$  be a mixed strategy of **player II**. Then

$x$  is a best response to  $y$

$\Leftrightarrow$  for all pure strategies  $i$  of **player I** :

$$x_i > 0 \Rightarrow (Ay)_i = u = \max\{ (Ay)_k \mid 1 \leq k \leq m \}.$$

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$(x, y)$  is a mixed equilibrium

$\Leftrightarrow$  for all pure strategies  $i$  of **player I** :

$$x_i = 0 \text{ or } (Ay)_i = u = \max\{ (Ay)_k \mid 1 \leq k \leq m \},$$

for all pure strategies  $j$  of **player II** :

$$y_j = 0 \text{ or}$$

$$(x^\top B)_j = v = \max\{ (x^\top B)_\ell \mid 1 \leq \ell \leq n \}.$$



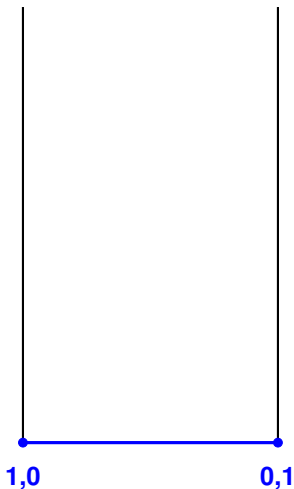
## Best responses to mixed strategy of player 2

	<b>4</b>	<b>5</b>	
<b>1</b>	<b>0</b>	<b>6</b>	<b>= A</b>
<b>2</b>	<b>2</b>	<b>5</b>	
<b>3</b>	<b>3</b>	<b>3</b>	

payoffs to  
player I



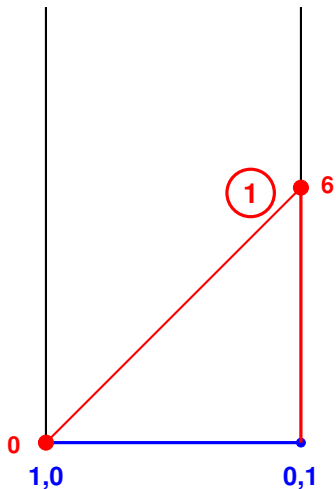
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<b>3</b>	<b>3</b>	<b>3</b>	

payoffs to  
player I

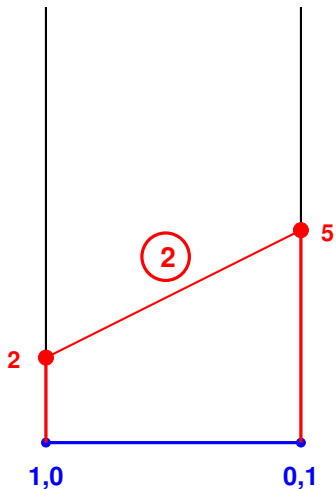
## Best responses to mixed strategy of player 2



	<b>4</b>	<b>5</b>	
<b>1</b>	0	6	= A
<b>2</b>	2	5	
<b>3</b>	3	3	

payoffs to  
player 1

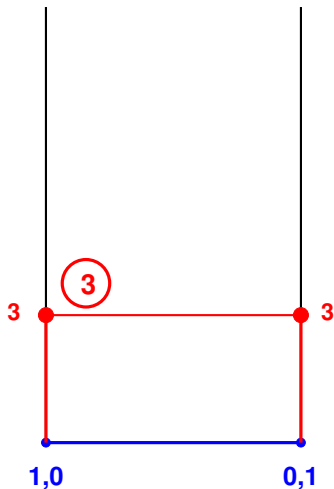
## Best responses to mixed strategy of player 2



	<b>4</b>	<b>5</b>	
<b>1</b>	0	6	
<b>2</b>	2	5	= A
<b>3</b>	3	3	

payoffs to  
player 1

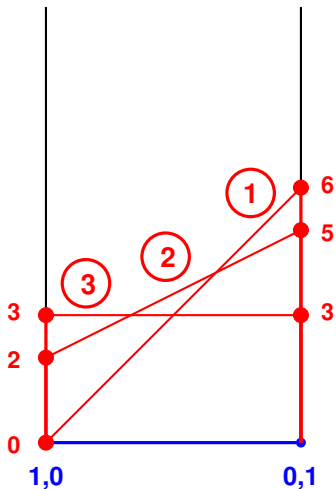
## Best responses to mixed strategy of player 2



	<b>4</b>	<b>5</b>	
<b>1</b>	0	6	
<b>2</b>	2	5	= A
<b>3</b>	3	3	

payoffs to  
player 1

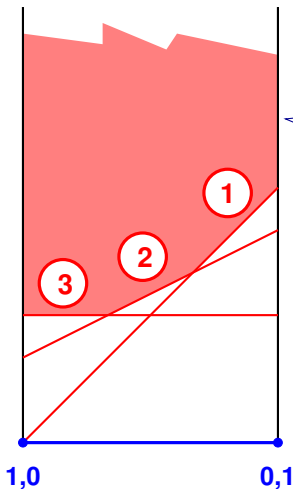
## Best responses to mixed strategy of player 2



	<b>4</b>	<b>5</b>	
<b>1</b>	0	6	= A
<b>2</b>	2	5	
<b>3</b>	3	3	

payoffs to  
player 1

## Best responses to mixed strategy of player 2

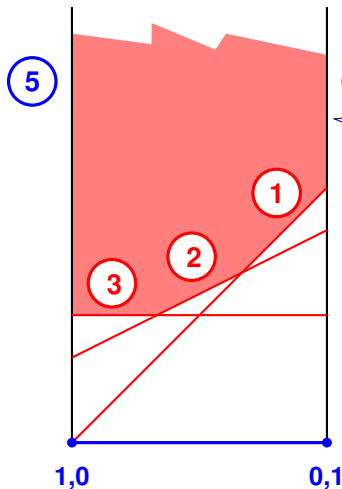


	(4)	(5)	
(1)	0	6	= A
(2)	2	5	
(3)	3	3	

payoffs to  
player I

best response polyhedron

## Best responses to mixed strategy of player 2



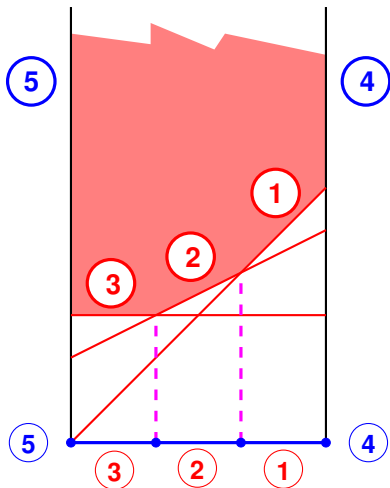
	4	5	
1	0	6	= A
2	2	5	
3	3	3	

payoffs to  
player I

**best response polyhedron  
with facet labels**



## Best responses to mixed strategy of player 2



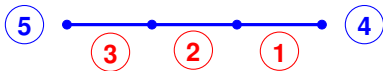
	(4)	(5)	
(1)	0	6	= A
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(3)	3	3	

payoffs to  
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## Best responses to mixed strategy of player 2

	<b>4</b>	<b>5</b>	
<b>1</b>	<b>0</b>	<b>6</b>	<b>= A</b>
<b>2</b>	<b>2</b>	<b>5</b>	
<b>3</b>	<b>3</b>	<b>3</b>	

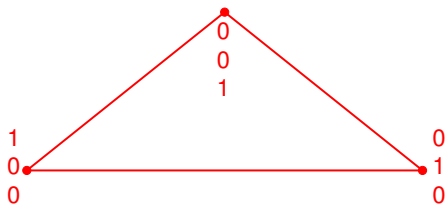
payoffs to  
player 1



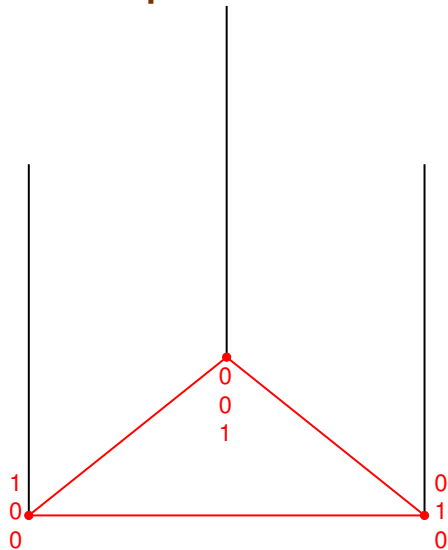
# Best responses to mixed strategy of player 1

	<b>4</b>	<b>5</b>	
<b>1</b>	2	1	= B
<b>2</b>	1	3	
<b>3</b>	4	3	

payoffs to  
player II



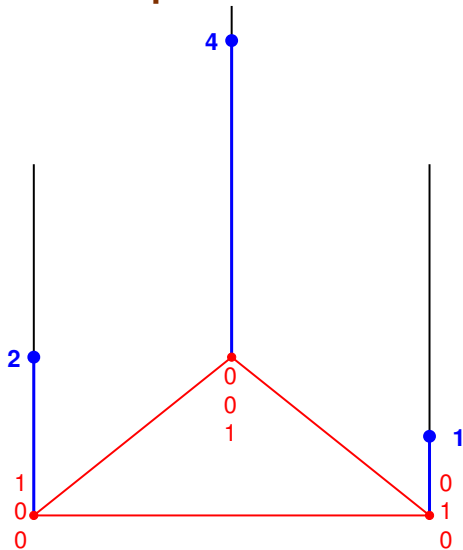
# Best responses to mixed strategy of player 1



	(4)	(5)	
(1)	2	1	= B
(2)	1	3	
(3)	4	3	

payoffs to  
player II

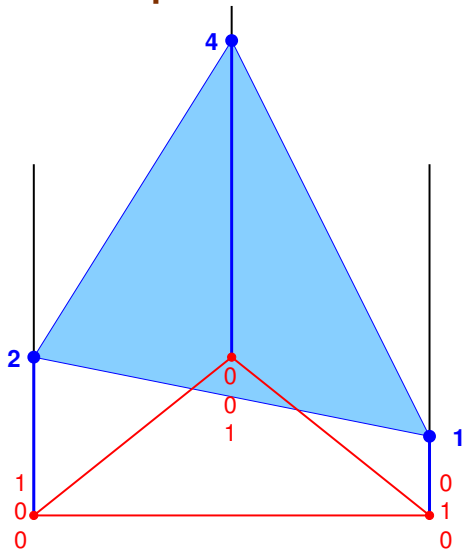
# Best responses to mixed strategy of player 1



	4	5	
1	2	1	= B
2	1	3	
3	4	3	

payoffs to  
player II

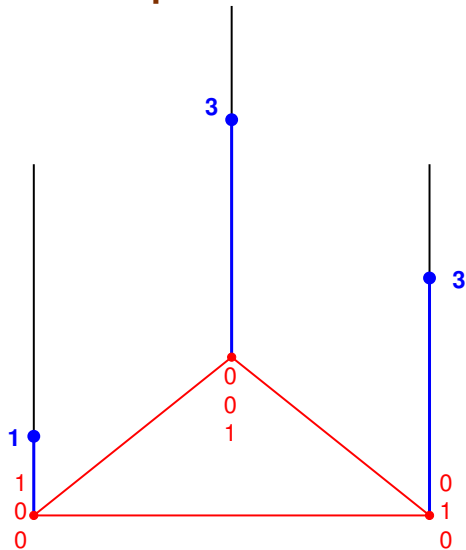
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	4	5	
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2	1	3	
3	4	3	

payoffs to  
player II

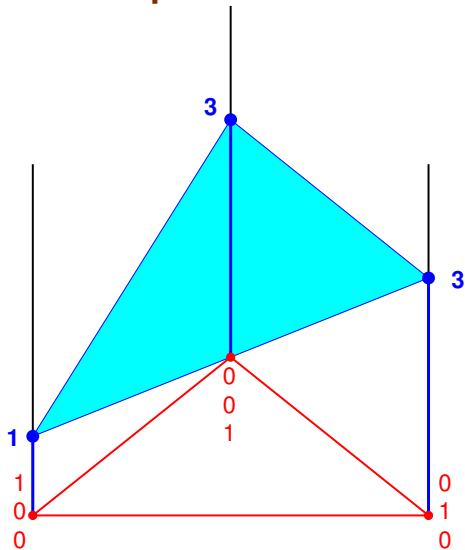
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payoffs to  
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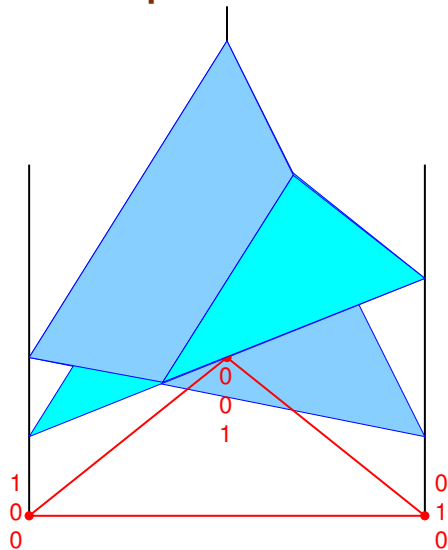


	4	5	
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2	1	3	
3	4	3	

payoffs to  
player II



# Best responses to mixed strategy of player 1

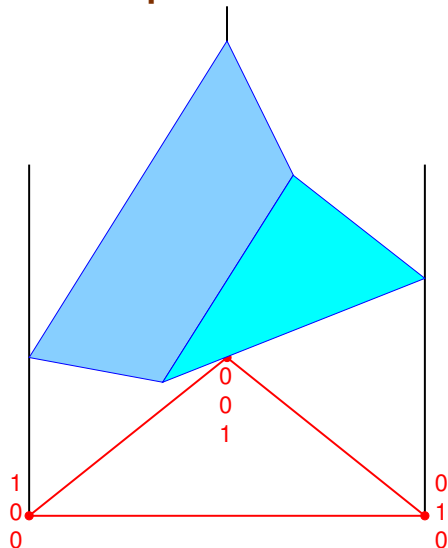


	4	5
1	2	1
2	1	3
3	4	3

= B

payoffs to  
player II

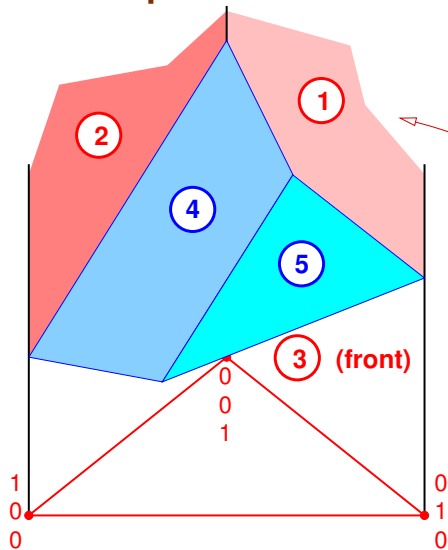
# Best responses to mixed strategy of player 1



	4	5	
1	2	1	= B
2	1	3	
3	4	3	

payoffs to  
player II

# Best responses to mixed strategy of player 1



	4	5
1	2	1
2	1	3
3	4	3

= B

payoffs to  
player II

**best response  
polyhedron  
with facet labels**

# Alternative view



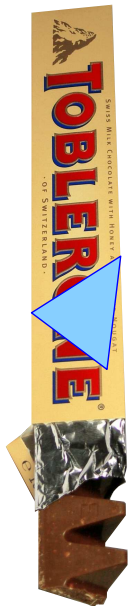
# Chop off Toblerone prism



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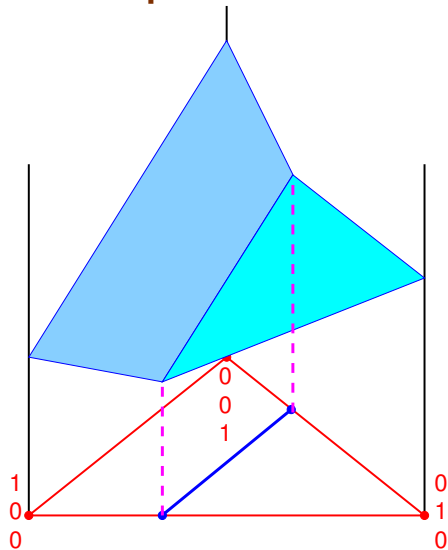




# Chop off Toblerone prism



# Best responses to mixed strategy of player 1



	4	5
1	2	1
2	1	3
3	4	3

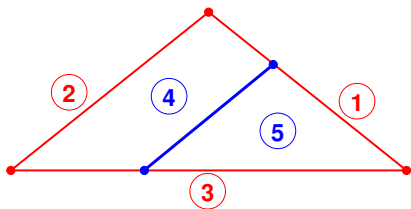
= B

payoffs to  
player II

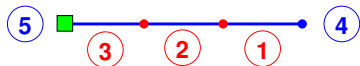
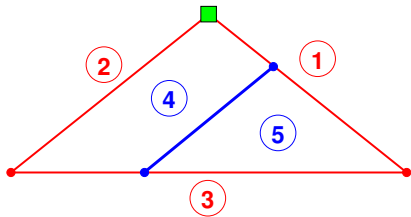
# Best responses to mixed strategy of player 1

	<b>4</b>	<b>5</b>	
<b>1</b>	2	1	= <b>B</b>
<b>2</b>	1	3	
<b>3</b>	4	3	

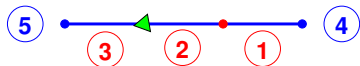
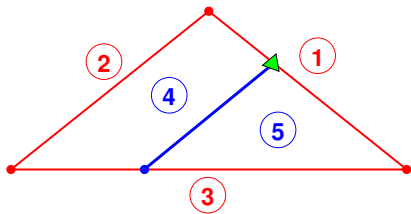
payoffs to  
player II



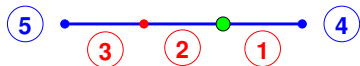
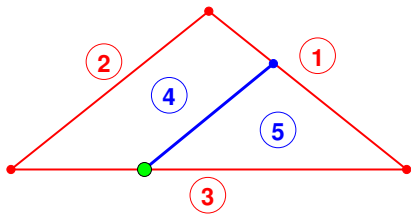
# Equilibrium = completely labeled strategy pair



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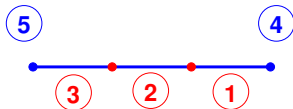
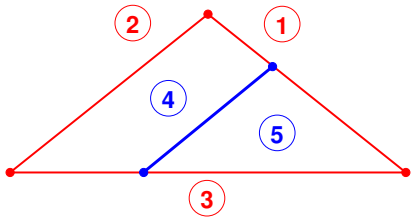


# Equilibrium = completely labeled strategy pair



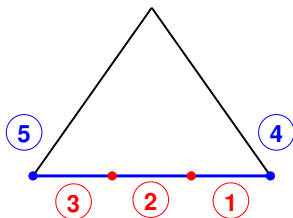
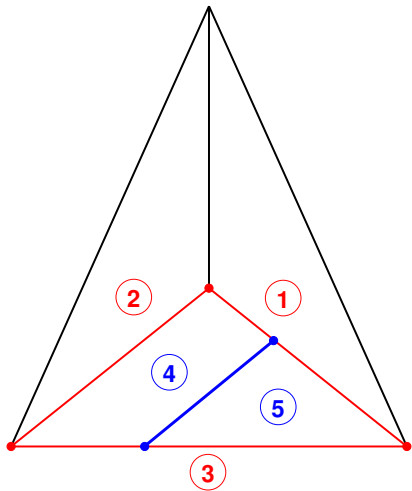
# The Lemke-Howson algorithm

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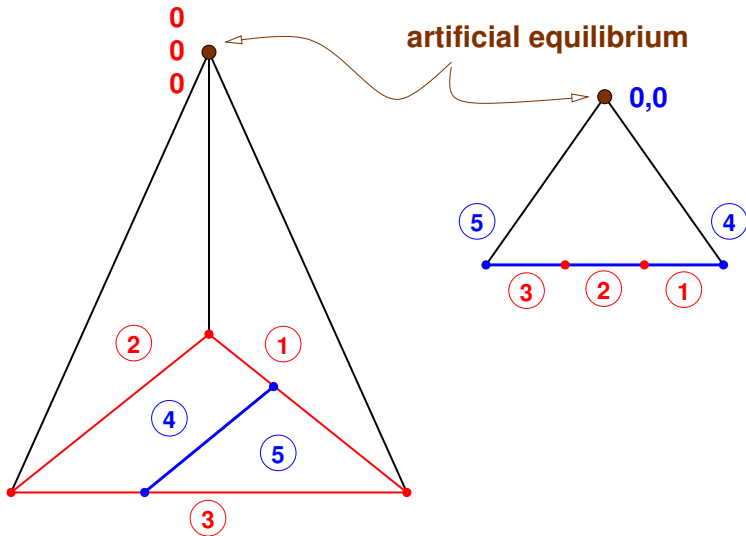




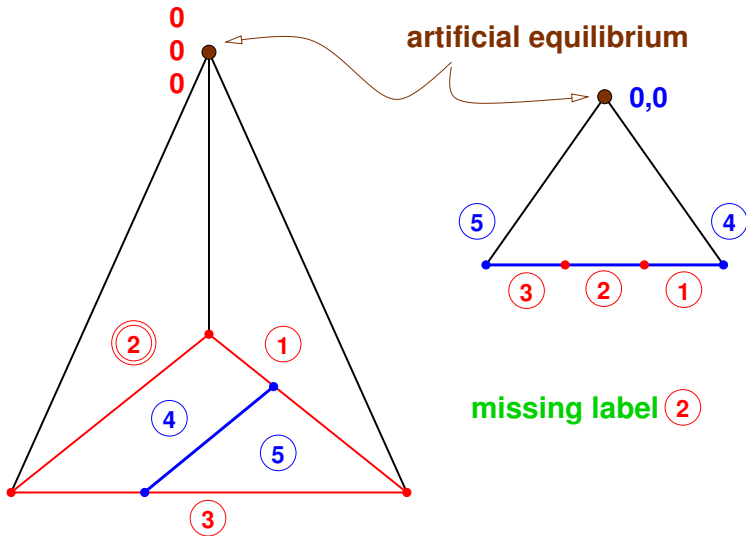
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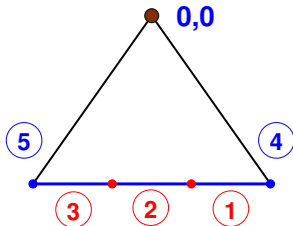
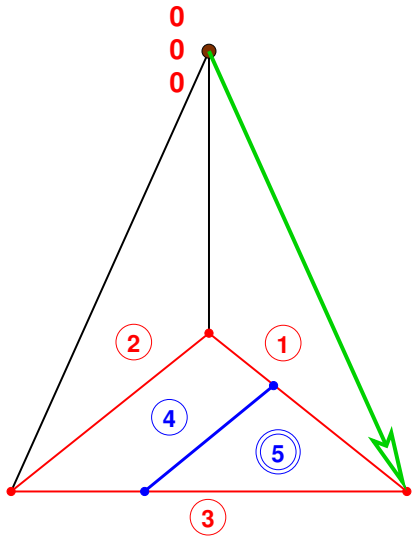
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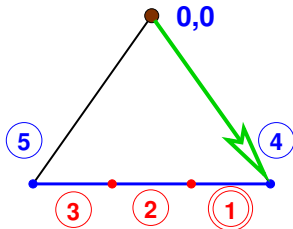
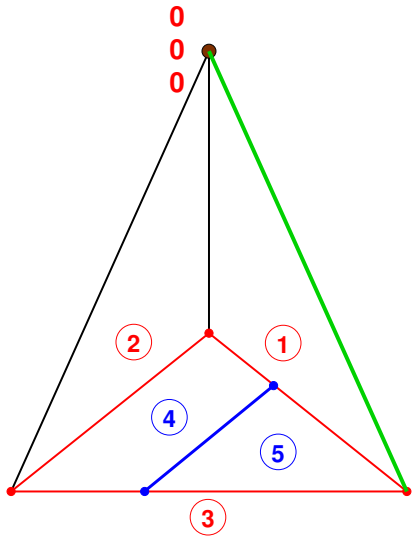


# The Lemke–Howson algorithm



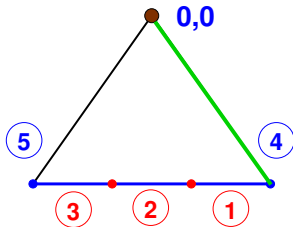
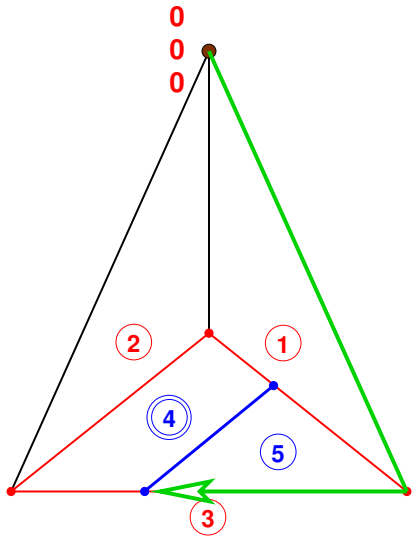
missing label (2)

# The Lemke–Howson algorithm



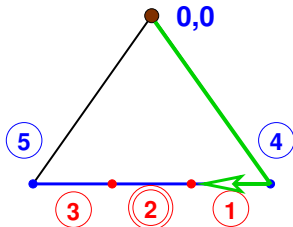
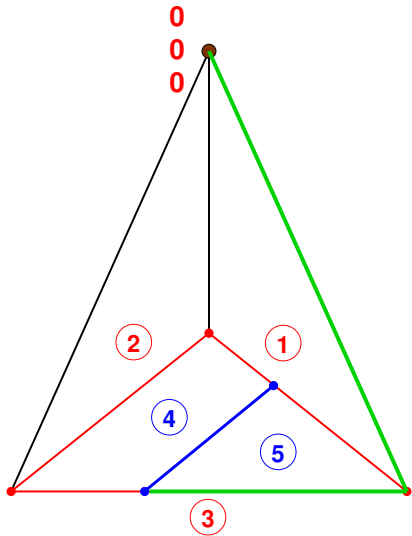
missing label (2)

# The Lemke–Howson algorithm



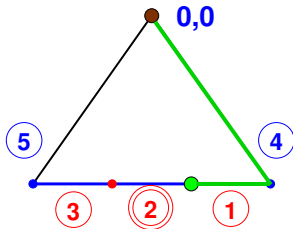
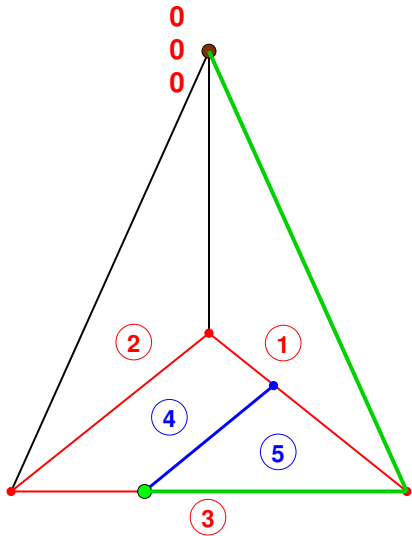
missing label 2

# The Lemke–Howson algorithm



missing label (2)

# The Lemke–Howson algorithm



found label (2)



## Why Lemke-Howson works

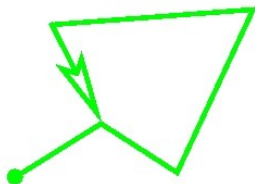
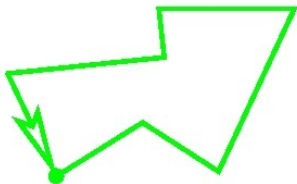
LH finds at least one Nash equilibrium because

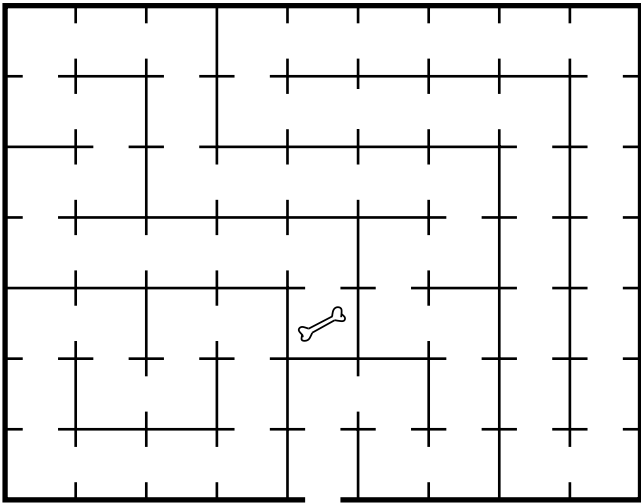
- **finitely many** "vertices"

for nondegenerate (generic) games:

- **unique** starting edge given missing label
- **unique** continuation

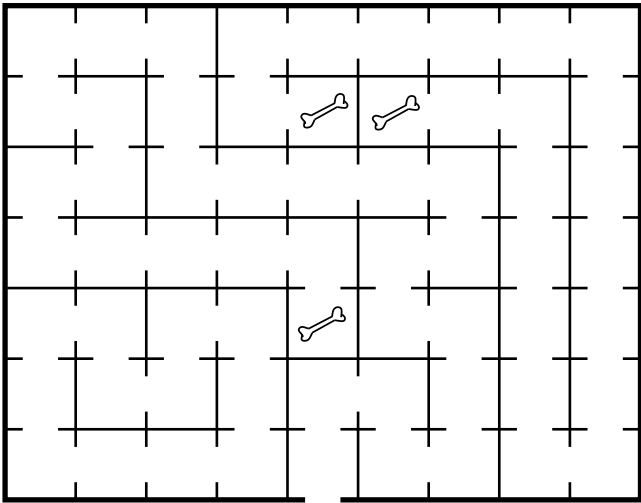
⇒ precludes "coming back" like here:









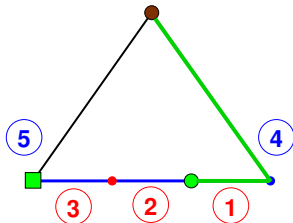
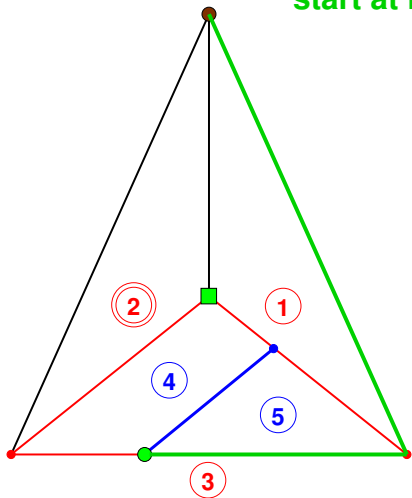






# The Lemke–Howson algorithm

start at Nash equilibrium ■

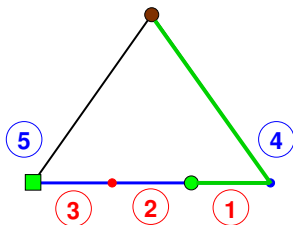
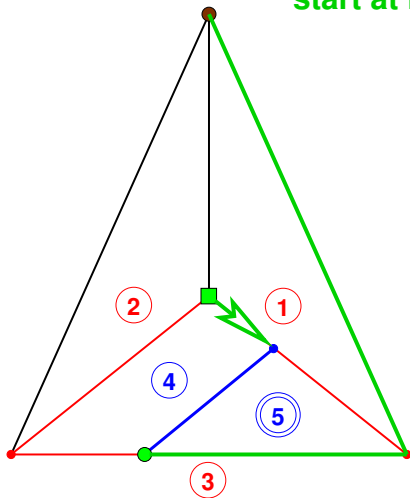


missing label (2)



# The Lemke–Howson algorithm

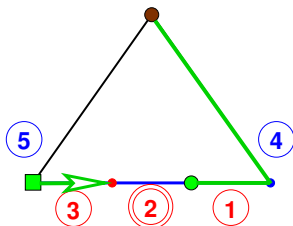
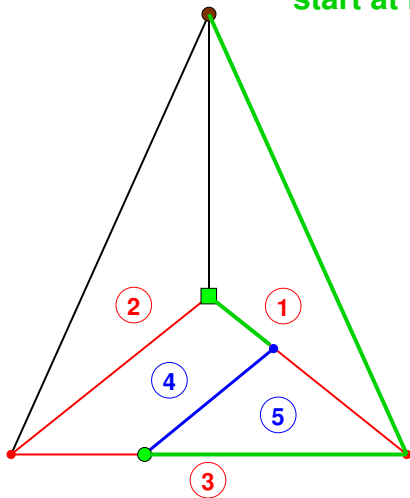
start at Nash equilibrium ■



missing label 2

# The Lemke–Howson algorithm

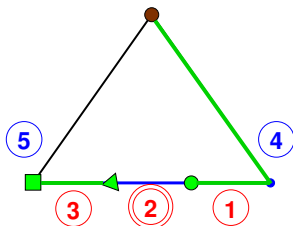
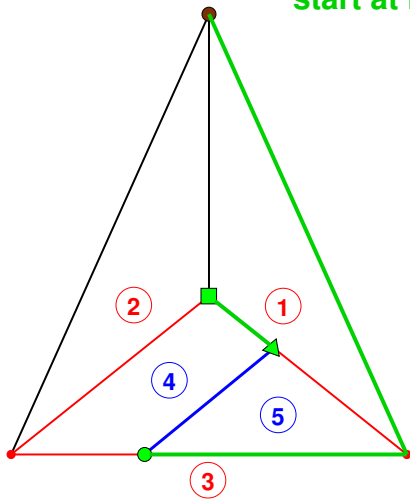
start at Nash equilibrium ■



missing label ②

# Odd number of Nash equilibria!

start at Nash equilibrium ■



found label ②

# Nondegenerate bimatrix games

Given:  $m \times n$  bimatrix game  $(A, B)$

$$X = \{ \mathbf{x} \in \mathbf{R}^m \mid \mathbf{x} \geq \mathbf{0}, x_1 + \dots + x_m = 1 \}$$

$$Y = \{ \mathbf{y} \in \mathbf{R}^n \mid \mathbf{y} \geq \mathbf{0}, y_1 + \dots + y_n = 1 \}$$

$$\text{supp}(\mathbf{x}) = \{ i \mid x_i > 0 \}$$

$$\text{supp}(\mathbf{y}) = \{ j \mid y_j > 0 \}$$

$(A, B)$  **nondegenerate**  $\Leftrightarrow \forall \mathbf{x} \in X, \mathbf{y} \in Y:$

$$| \{ j \mid j \text{ best response to } \mathbf{x} \} | \leq | \text{supp}(\mathbf{x}) |,$$

$$| \{ i \mid i \text{ best response to } \mathbf{y} \} | \leq | \text{supp}(\mathbf{y}) |.$$

## Nondegeneracy via labels

$m \times n$  bimatrix game  $(A, B)$  **nondegenerate**

$\Leftrightarrow$  no  $x \in X$  has more than  $m$  labels,  
no  $y \in Y$  has more than  $n$  labels.

E.g.  $x$  with  $> m$  labels,

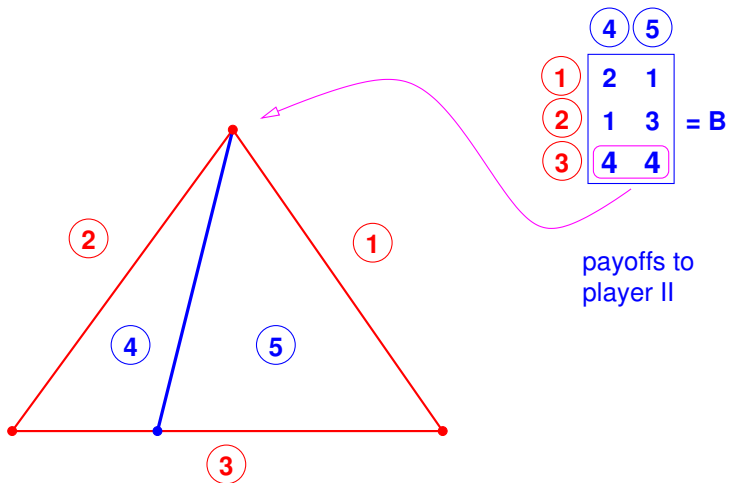
$s$  labels from  $\{1, \dots, m\}$ ,

$\Rightarrow > m-s$  labels from  $\{m+1, \dots, m+n\}$

$\Leftrightarrow > |\text{supp}(x)|$  **best responses** to  $x$ .

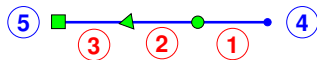
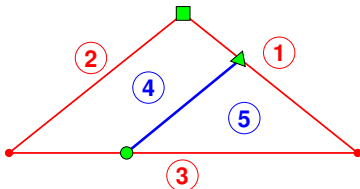
$\Rightarrow$  degenerate.

## Example of a degenerate game

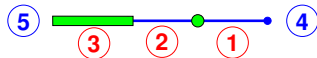
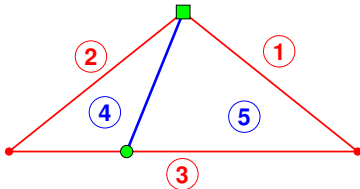


# Equilibrium components in a degenerate game

nondegenerate game:



degenerate game, same payoffs for **player I**:

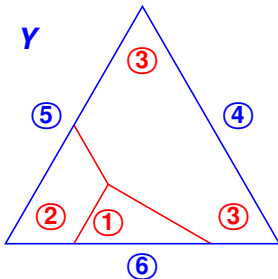


# Best-response diagrams for a $3 \times 3$ game

Consider the  $3 \times 3$  game

$$A = \begin{matrix} \textcircled{1} \\ \textcircled{2} \\ \textcircled{3} \end{matrix} \begin{bmatrix} 0 & 3 & 0 \\ 1 & 0 & 1 \\ -3 & 4 & 5 \end{bmatrix}, \quad B = \begin{matrix} \textcircled{4} & \textcircled{5} & \textcircled{6} \\ \begin{bmatrix} 0 & 1 & -2 \\ 2 & 0 & 3 \\ 2 & 1 & 0 \end{bmatrix} \end{matrix}.$$

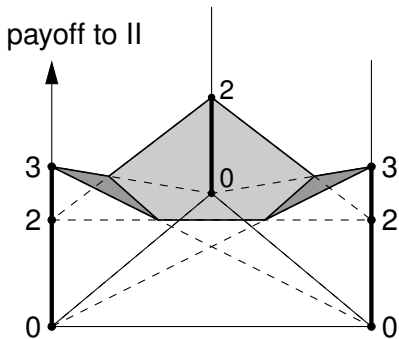
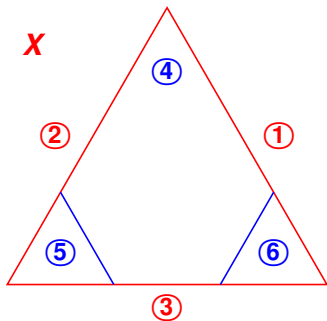
Subdivision of  $Y$  into best-response regions:





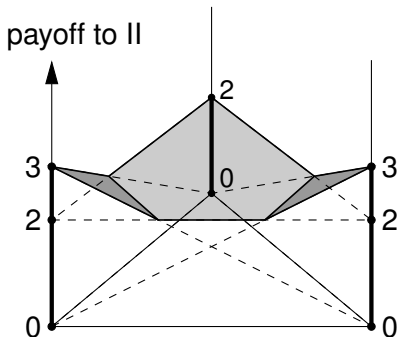
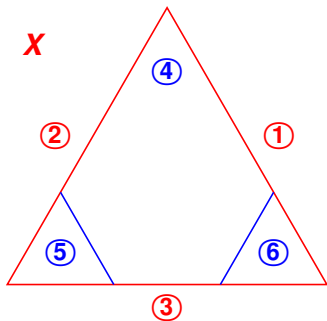
# Upper envelope – with “row shift” of $B$

$$B' = \begin{matrix} \textcircled{4} & \textcircled{5} & \textcircled{6} \\ \begin{bmatrix} 2 & 3 & 0 \\ 2 & 0 & 3 \\ 2 & 0 & 0 \end{bmatrix} \end{matrix}$$

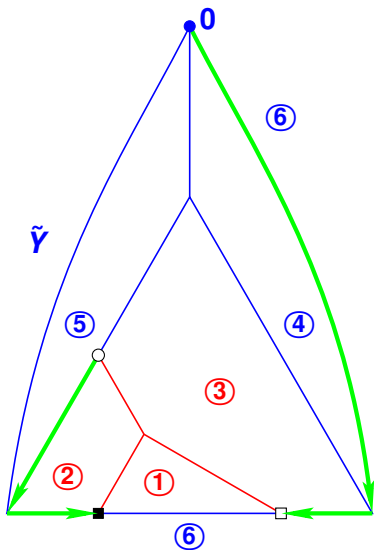
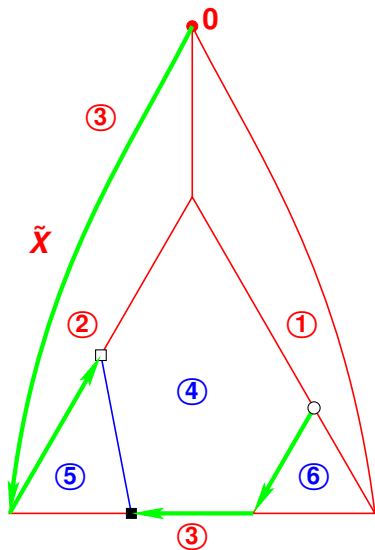


# Upper envelope – with “row shift” of $B$

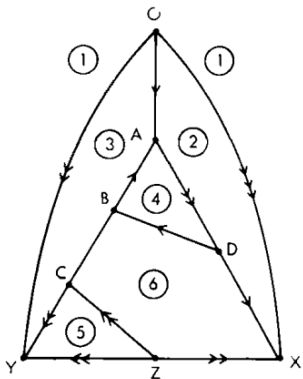
$$\begin{matrix} \textcircled{4} & \textcircled{5} & \textcircled{6} \\ B' = \begin{bmatrix} 2 & 3 & 0 \\ 2 & 0 & 3 \\ 2 & 0 & 0 \end{bmatrix}, & B = \begin{bmatrix} 0 & 1 & -2 \\ 2 & 0 & 3 \\ 2 & 0 & 0 \end{bmatrix}, & \mathbf{x}^\top B' = \mathbf{x}^\top B + x_1 [2 \ 2 \ 2]
 \end{matrix}$$



# Best-response diagrams $X$ and $Y$ and Lemke-Howson

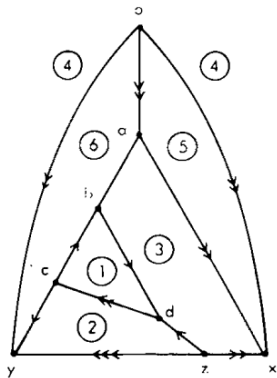


# Diagrams from Shapley (1974)



Key:

$\rho^1$   $O_oA_oD_xX$   $YyCzZ$   
 $\rho^2$   $O_oYy$   $XxZz$   
 $\rho^3$   $O_oXx$   $YyZz$   
 $\rho^4$   $oO_aXx$   $yY_cCdZz$   
 $\rho^5$   $oO_yY$   $xXzZ$  ...  $bAcBdDb$  ...  
 $\rho^6$   $oO_xX$   $yYzZ$

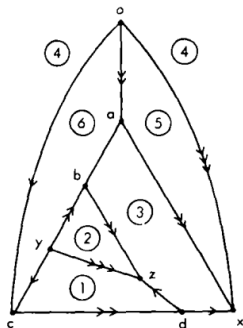
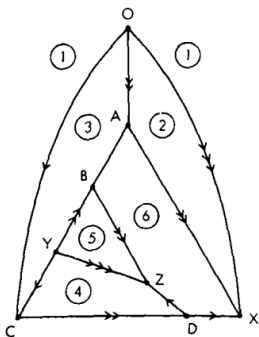


Payoffs:

2	2	0	3	0	?
0	3	0	0	3	2
3	0	1	0	0	1

# from Robert Wilson, in Shapley (1974)

Lemke-Howson may only find **some** equilibria:



Key:

$\rho^1$ OoAxX	yYbZz
$\rho^2$ OoCaDxx	yYcBdZz
$\rho^3$ OoXx	YyZz
$\rho^4$ oOaXx	YyBzZ
$\rho^5$ oOcAdXx	YyCbDzZ
$\rho^6$ oOxX	yYzZ

Payoffs:

0	3	0	0	2	3
2	2	0	3	2	0
3	0	1	0	0	1

## Running time of Lemke-Howson

The running time of Lemke-Howson may be **exponential** in the size of the game:

R. Savani and B. von Stengel (2004), Exponentially many steps for finding a Nash equilibrium in a bimatrix game. In: *Proc. 45th Annual IEEE Symposium on Foundations of Computer Science (FOCS 2004)*, 258–267.

R. Savani and B. von Stengel (2006), Hard-to-solve bimatrix games. *Econometrica* 74, 397–429.

R. Savani and B. von Stengel (2016), Unit vector games. *International Journal of Economic Theory* 12, 7–27.

# Questions

- how to implement Lemke-Howson as an algorithm
  - use labeled polytopes
  - complementary pivoting
- handling degenerate games
- finding one vs. all Nash equilibria
  - possibly exponentially many NE
  - uniqueness is co-NP-complete
- running time of Lemke-Howson
  - worst-case: exponential
  - average case?
  - smoothed analysis?

Labeled polytopes and  
completely labeled vertex pairs

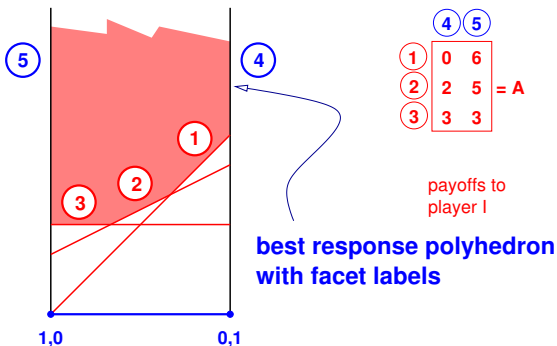


# Best-response polyhedra and polytopes

best-response **polyhedra**:

$$\bar{P} = \{ (\mathbf{x}, v) \in \mathbf{X} \times \mathbb{R} \mid \mathbf{B}^T \mathbf{x} \leq \mathbf{1}v \}$$

$$\bar{Q} = \{ (\mathbf{y}, u) \in \mathbf{Y} \times \mathbb{R} \mid \mathbf{A} \mathbf{y} \leq \mathbf{1}u \}$$



# Best-response polyhedra and polytopes

best-response **polyhedra**:

$$\bar{P} = \{ (\mathbf{x}, \mathbf{v}) \in \mathbf{X} \times \mathbb{R} \mid \mathbf{B}^T \mathbf{x} \leq \mathbf{1} \mathbf{v} \}$$

$$\bar{Q} = \{ (\mathbf{y}, \mathbf{u}) \in \mathbf{Y} \times \mathbb{R} \mid \mathbf{A} \mathbf{y} \leq \mathbf{1} \mathbf{u} \}$$

best-response **polytopes**:

$$P = \{ \mathbf{x} \in \mathbb{R}^m \mid \mathbf{x} \geq \mathbf{0}, \mathbf{B}^T \mathbf{x} \leq \mathbf{1} \}$$

$$Q = \{ \mathbf{y} \in \mathbb{R}^n \mid \mathbf{A} \mathbf{y} \leq \mathbf{1}, \mathbf{y} \geq \mathbf{0} \}$$

obtained from  $\bar{P}$ ,  $\bar{Q}$  via  $\mathbf{x} \mapsto \mathbf{x} \frac{1}{\mathbf{v}}$ ,  $\mathbf{y} \mapsto \mathbf{y} \frac{1}{\mathbf{u}}$

(requires  $\mathbf{u}, \mathbf{v} > \mathbf{0}$ , if needed via adding constants to  $\mathbf{A}, \mathbf{B}$ )

re-normalized to  $\mathbf{X}, \mathbf{Y}$  via  $\mathbf{x} \mapsto \mathbf{x} \frac{1}{\mathbf{1}^T \mathbf{x}}$ ,  $\mathbf{y} \mapsto \mathbf{y} \frac{1}{\mathbf{1}^T \mathbf{y}}$

## Labeled polytopes

$$P = \{ \mathbf{x} \in \mathbb{R}^m \mid \mathbf{x} \geq \mathbf{0}, \mathbf{B}^\top \mathbf{x} \leq \mathbf{1} \}$$

$$Q = \{ \mathbf{y} \in \mathbb{R}^n \mid \mathbf{A}\mathbf{y} \leq \mathbf{1}, \mathbf{y} \geq \mathbf{0} \}$$

$(\mathbf{x}, \mathbf{y}) \in P \times Q$  (re-normalized in  $\mathbf{X} \times \mathbf{Y}$ ) equilibrium of  $(\mathbf{A}, \mathbf{B})$

$\Leftrightarrow$

$$\mathbf{x} \geq \mathbf{0} \quad \perp \quad \mathbf{A}\mathbf{y} \leq \mathbf{1} \quad (\text{labels } 1, \dots, m)$$

$$\mathbf{y} \geq \mathbf{0} \quad \perp \quad \mathbf{B}^\top \mathbf{x} \leq \mathbf{1} \quad (\text{labels } m+1, \dots, m+n)$$

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$$\mathbf{y} \geq \mathbf{0} \quad \perp \quad \mathbf{B}^\top \mathbf{x} \leq \mathbf{1} \quad (\text{labels } m+1, \dots, m+n)$$

artificial equilibrium  $(\mathbf{x}, \mathbf{y}) = (\mathbf{0}, \mathbf{0})$ , not in  $\mathbf{X} \times \mathbf{Y}$ , not NE.

## Only one labeled polytope

$$P = \{ \mathbf{x} \in \mathbb{R}^m \mid \mathbf{x} \geq \mathbf{0}, \mathbf{B}^\top \mathbf{x} \leq \mathbf{1} \}$$

$$Q = \{ \mathbf{y} \in \mathbb{R}^n \mid \mathbf{A} \mathbf{y} \leq \mathbf{1}, \mathbf{y} \geq \mathbf{0} \}$$

$$R = \{ \mathbf{z} \in \mathbb{R}^k \mid \mathbf{z} \geq \mathbf{0}, \mathbf{C} \mathbf{z} \leq \mathbf{1} \}$$

$$R = P \times Q,$$

$$k = m + n,$$

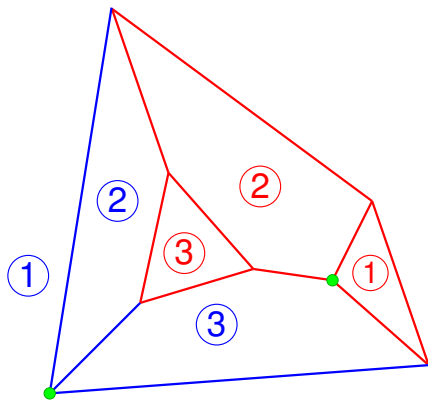
$$\mathbf{C} = \begin{pmatrix} \mathbf{0} & \mathbf{A} \\ \mathbf{B}^\top & \mathbf{0} \end{pmatrix} \in \mathbb{R}^{k \times k}, \quad \mathbf{z} = (\mathbf{x}, \mathbf{y})$$

equilibrium  $\mathbf{z} \Leftrightarrow \mathbf{z} \geq \mathbf{0} \perp \mathbf{C} \mathbf{z} \leq \mathbf{1}$  (labels  $1, \dots, k$ )

artificial equilibrium  $\mathbf{z} = \mathbf{0}$ , any other  $\mathbf{z} = (\mathbf{x}, \mathbf{y})$  with  $\mathbf{x}$  re-normalized in  $\mathbf{X}$  and  $\mathbf{y}$  in  $\mathbf{Y}$  is NE of  $(\mathbf{A}, \mathbf{B})$

# Path of “almost completely labeled” edges

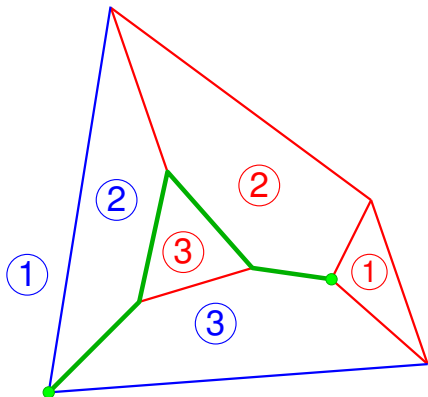
$$R = \{z \in \mathbb{R}^k \mid z \geq 0, Cz \leq 1\}$$



# Path of “almost completely labeled” edges

$$R = \{z \in \mathbb{R}^k \mid z \geq 0, Cz \leq 1\}$$

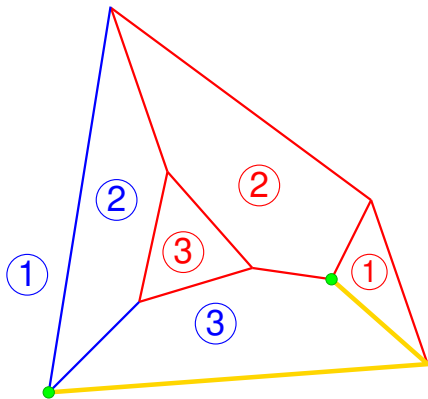
missing label ① :



# Path of “almost completely labeled” edges

$$R = \{z \in \mathbb{R}^k \mid z \geq 0, Cz \leq 1\}$$

missing label ② :

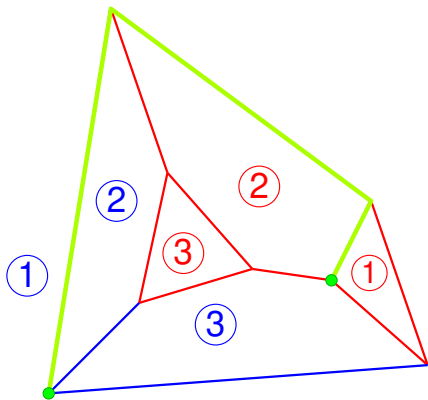




# Path of “almost completely labeled” edges

$$R = \{z \in \mathbb{R}^k \mid z \geq 0, Cz \leq 1\}$$

missing label ③ :



# Algebraic implementation by pivoting

## Complementary pivoting

$$\mathbf{z} \geq \mathbf{0} \perp \mathbf{Cz} \leq \mathbf{1}$$

$$\Leftrightarrow \mathbf{z} \geq \mathbf{0} \perp \mathbf{s} \geq \mathbf{0}, \quad \boxed{\mathbf{Cz} + \mathbf{s} = \mathbf{1}}$$

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$\mathbf{z} \geq \mathbf{0}, \mathbf{s} \geq \mathbf{0}$   $\ell$ -almost complementary (missing label  $\ell$ )

$$\Leftrightarrow \mathbf{Cz} + \mathbf{s} = \mathbf{1}, \quad \boxed{\mathbf{z}_i \mathbf{s}_i = 0} \quad \text{for } i = 1, \dots, k, \quad \boxed{i \neq \ell}$$

## Complementary pivoting

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**complementary pivoting** = algebraic traversal of  $\ell$ -almost complementary edges of  $\{\mathbf{z} \in \mathbb{R}^k \mid \mathbf{z} \geq \mathbf{0}, \mathbf{Cz} \leq \mathbf{1}\}$

starting with  $\mathbf{z} = \mathbf{0}, \quad \mathbf{s} = \mathbf{1} - \mathbf{Cz}$ .

# Complementary pivoting

$$\mathbf{z} \geq \mathbf{0} \perp \mathbf{Cz} \leq \mathbf{1}$$

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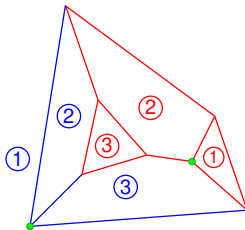
$\mathbf{z} \geq \mathbf{0}, \mathbf{s} \geq \mathbf{0}$   $\ell$ -almost complementary (missing label  $\ell$ )

$$\Leftrightarrow \mathbf{Cz} + \mathbf{s} = \mathbf{1}, \quad \boxed{\mathbf{z}_i \mathbf{s}_i = 0} \quad \text{for } i = 1, \dots, k, \quad \boxed{i \neq \ell}$$

**complementary pivoting** = algebraic traversal of  $\ell$ -almost complementary edges of  $\{\mathbf{z} \in \mathbb{R}^k \mid \mathbf{z} \geq \mathbf{0}, \mathbf{Cz} \leq \mathbf{1}\}$

starting with  $\mathbf{z} = \mathbf{0}, \quad \mathbf{s} = \mathbf{1} - \mathbf{Cz}$ .

**Example:**  $\mathbf{C} = \begin{bmatrix} 0 & 3 & 0 \\ 2 & 2 & 2 \\ 3 & 0 & 0 \end{bmatrix}$



## Almost complementary dictionaries

**dictionary** = any equivalent system to  $\mathbf{Cz} + \mathbf{s} = \mathbf{1}$

**basic** variables expressed depending on **nonbasic variables**

- nonbasic variables set to  $\mathbf{0}$  :
  - gives **basic solution** = polytope **vertex**,
  - nonbasic variables = binding inequalities = vertex **labels**
- starting dictionary:  $\mathbf{s} = \mathbf{1} - \mathbf{Cz}$

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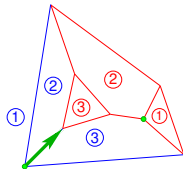
choose **entering column** = entering nonbasic variable  $\mathbf{z}_l$

identify *the* **leaving row** = leaving basic variable, here  $\mathbf{s}_3$

$$\mathbf{s}_1 = \mathbf{1} - 3\mathbf{z}_2$$

$$\mathbf{s}_2 = \mathbf{1} - 2\mathbf{z}_1 - 2\mathbf{z}_2 - 2\mathbf{z}_3$$

$$\mathbf{s}_3 = \mathbf{1} - 3\mathbf{z}_1$$





## Complementary variables

$$s_1 = 1 - 3z_2$$

$$s_2 = 1 - 2z_1 - 2z_2 - 2z_3$$

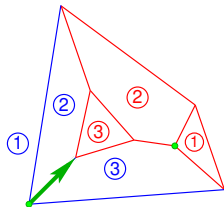
$$s_3 = 1 - 3z_1$$

$z_1$  enters,  $s_3$  leaves:

$$s_1 = 1 - 3z_2$$

$$s_2 = \frac{1}{3} + \frac{2}{3}s_3 - 2z_2 - 2z_3$$

$$z_1 = \frac{1}{3} - \frac{1}{3}s_3$$



## Complementary variables

$$s_1 = 1 - 3z_2$$

$$s_2 = 1 - 2z_1 - 2z_2 - 2z_3$$

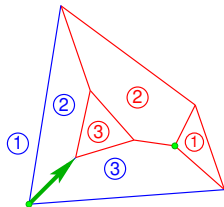
$$s_3 = 1 - 3z_1$$

$z_1$  enters,  $s_3$  leaves:

$$s_1 = 1 - 3z_2$$

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$$z_1 = \frac{1}{3} - \frac{1}{3}s_3$$

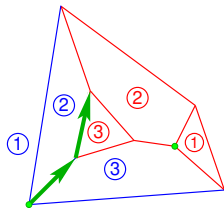


$z_3$  enters,  $s_2$  leaves:

$$s_1 = 1 - 3z_2$$

$$z_3 = \frac{1}{6} + \frac{1}{3}s_3 - z_2 - \frac{1}{2}s_2$$

$$z_1 = \frac{1}{3} - \frac{1}{3}s_3$$



## complementary pivoting, continued

...  $\mathbf{s}_2$  left the basis,  $\mathbf{z}_2$  enters:

$$\mathbf{s}_1 = 1 - 3\mathbf{z}_2$$

$$\mathbf{z}_3 = \frac{1}{6} + \frac{1}{3}\mathbf{s}_3 - \mathbf{z}_2 - \frac{1}{2}\mathbf{s}_2$$

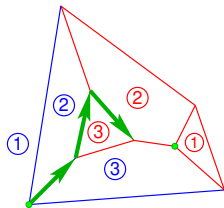
$$\mathbf{z}_1 = \frac{1}{3} - \frac{1}{3}\mathbf{s}_3$$

$\mathbf{z}_3$  leaves:

$$\mathbf{s}_1 = \frac{1}{2} - \mathbf{s}_3 + 3\mathbf{z}_3 + \frac{3}{2}\mathbf{s}_2$$

$$\mathbf{z}_2 = \frac{1}{6} + \frac{1}{3}\mathbf{s}_3 - \mathbf{z}_3 - \frac{1}{2}\mathbf{s}_2$$

$$\mathbf{z}_1 = \frac{1}{3} - \frac{1}{3}\mathbf{s}_3$$



## complementary pivoting, continued

...  $\mathbf{s}_2$  left the basis,  $\mathbf{z}_2$  enters:

$$\mathbf{s}_1 = 1 - 3\mathbf{z}_2$$

$$\mathbf{z}_3 = \frac{1}{6} + \frac{1}{3}\mathbf{s}_3 - \mathbf{z}_2 - \frac{1}{2}\mathbf{s}_2$$

$$\mathbf{z}_1 = \frac{1}{3} - \frac{1}{3}\mathbf{s}_3$$

$\mathbf{z}_3$  leaves:

$$\mathbf{s}_1 = \frac{1}{2} - \mathbf{s}_3 + 3\mathbf{z}_3 + \frac{3}{2}\mathbf{s}_2$$

$$\mathbf{z}_2 = \frac{1}{6} + \frac{1}{3}\mathbf{s}_3 - \mathbf{z}_3 - \frac{1}{2}\mathbf{s}_2$$

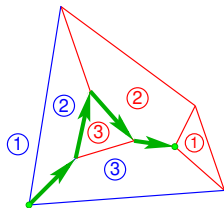
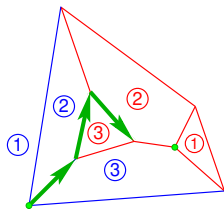
$$\mathbf{z}_1 = \frac{1}{3} - \frac{1}{3}\mathbf{s}_3$$

$\mathbf{s}_3$  enters,  $\mathbf{s}_1$  leaves, equilibrium found:

$$\mathbf{s}_3 = \frac{1}{2} - \mathbf{s}_1 + 3\mathbf{z}_3 + \frac{3}{2}\mathbf{s}_2$$

$$\mathbf{z}_2 = \frac{1}{3} - \frac{1}{3}\mathbf{s}_1$$

$$\mathbf{z}_1 = \frac{1}{6} + \frac{1}{3}\mathbf{s}_1 - \mathbf{z}_3 - \frac{1}{2}\mathbf{s}_2$$



# Labeled polytopes and bimatrix games

## Did we solve a game?

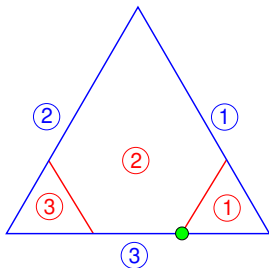
Yes!

$\mathbf{z} = (\frac{1}{6}, \frac{1}{2}, \mathbf{0})^\top$  is normalized  $\bar{\mathbf{z}} = (\frac{1}{3}, \frac{2}{3}, \mathbf{0})^\top$  and a (here unique) symmetric equilibrium  $(\bar{\mathbf{z}}, \bar{\mathbf{z}})$  of the game  $(\mathbf{C}, \mathbf{C}^\top)$  with

$$\mathbf{C} = \begin{bmatrix} 0 & 3 & 0 \\ 2 & 2 & 2 \\ 3 & 0 & 0 \end{bmatrix}, \quad \text{that is,}$$

$$\bar{\mathbf{z}} \geq \mathbf{0} \perp \mathbf{C}\bar{\mathbf{z}} \leq \mathbf{1}\mathbf{u} \quad \text{with payoff } \mathbf{u} = \mathbf{2} = \frac{\mathbf{1}}{\mathbf{1}^\top \mathbf{z}}$$

	①	②	③
①	0	3	0
②	2	2	2
③	3	0	0



## Arbitrary labeled polytopes

**simple** polytope in  $\mathbb{R}^m \Leftrightarrow$  every vertex on only  $m$  facets

**labeled** (simple) polytope in  $\mathbb{R}^m$  :

every facet has one label in  $\{1, \dots, m\}$

completely labeled vertex = its facets have all labels  $1, \dots, m$

## Arbitrary labeled polytopes

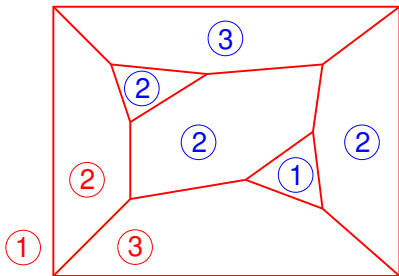
**simple** polytope in  $\mathbb{R}^m \Leftrightarrow$  every vertex on only  $m$  facets

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**Theorem** The number of completely labeled vertices is **even**.





## Arbitrary labeled polytopes

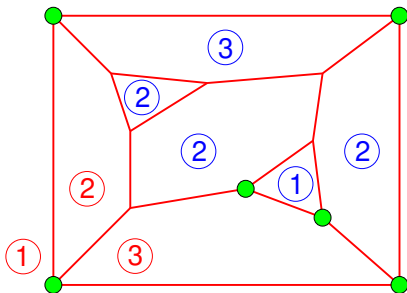
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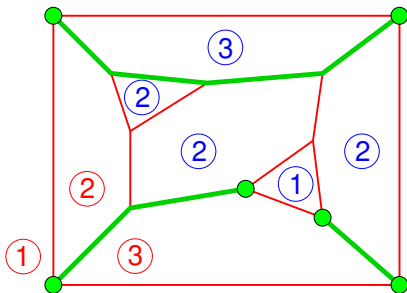
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**Theorem** The number of completely labeled vertices is **even**.



# Unit vector games

Let  $\mathbf{b}_1, \dots, \mathbf{b}_n \in \mathbb{R}^m$ ,  $\mathbf{B} = [\mathbf{b}_1 \cdots \mathbf{b}_n]$

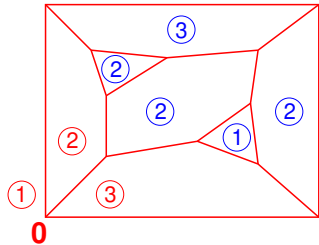
$\ell(1), \dots, \ell(n) \in \{1, \dots, m\}$  be labels

$$\mathbf{P} = \{\mathbf{x} \in \mathbb{R}^m \mid \mathbf{x} \geq \mathbf{0}, \mathbf{B}^\top \mathbf{x} \leq \mathbf{1}\}$$

with labels of  $\mathbf{P}$  for binding inequalities:

label  $i$  :  $\mathbf{x}_i \geq 0 \quad (1 \leq i \leq m)$

label  $\ell(j)$  :  $\mathbf{b}_j^\top \mathbf{x} \leq 1 \quad (1 \leq j \leq n)$



**Theorem**  $\mathbf{x} \neq \mathbf{0}$  completely labeled vertex of  $\mathbf{P} \Leftrightarrow$

$(\mathbf{x}, \mathbf{y})$  Nash equilibrium of  $(\mathbf{U}, \mathbf{B})$  where  $\mathbf{U} = [\mathbf{e}_{\ell(1)} \cdots \mathbf{e}_{\ell(n)}]$

$\mathbf{e}_i = i$ th unit vector in  $\mathbb{R}^m$

# Summary

Nash equilibria of bimatrix games

**are** completely labeled vertices of facet-labeled polytopes  $P$

(assuming there is one completely labeled vertex  $\mathbf{x} = \mathbf{0}$  of  $P$  whose incident facet inequalities can w.l.o.g. be written as  $\mathbf{x} \geq \mathbf{0}$ , which is not a NE but the artificial equilibrium).

For generic games (simple polytopes), the number of completely labeled vertices is **even**, and hence the number of NE is odd.

**Evenness = Parity Argument, complexity class PPAD.**

Degeneracy resolution

Integer pivoting

# Degeneracy

In pivoting, **degeneracy** means at least one **zero** basic variable in a basic feasible solution

⇒ additional **labels** as binding inequalities (not just the nonbasic variables)

occurs when **leaving variable not unique**

**Example:**  $z_2$  enters:

$$s_1 = 1 - 3z_2$$

$$z_1 = \frac{1}{3} + \frac{2}{3}s_2 - z_2$$

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Apply to general system  $Ax = b$ ,  $x \geq 0$  written as

$A_B x_B + A_N x_N = b$  with basic columns  $B$ , nonbasic columns  $N$

# Lexicographic degeneracy resolution

$$\mathbf{Ax} = \mathbf{b}$$



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$$Ax = b$$

$$A_B x_B + A_N x_N = b$$

$$A_B x_B = b - A_N x_N$$

dictionary

$x_B$	=	$A_B^{-1} b$	-	$A_B^{-1} A_N x_N$
-------	---	--------------	---	--------------------

## Lexicographic degeneracy resolution

perturb  $\mathbf{b}$  to  $\mathbf{b} + \vec{\epsilon}_*$  with small  $\epsilon > 0$ ,  $\vec{\epsilon} = (1, \epsilon, \epsilon^2, \dots, \epsilon^m)^\top$

$$\mathbf{A}\mathbf{x} = \mathbf{b}$$

$$\mathbf{A}_B \mathbf{x}_B + \mathbf{A}_N \mathbf{x}_N = \mathbf{b}$$

$$\mathbf{A}_B \mathbf{x}_B = \mathbf{b} - \mathbf{A}_N \mathbf{x}_N$$

dictionary  $\mathbf{x}_B = \mathbf{A}_B^{-1} \mathbf{b} - \mathbf{A}_B^{-1} \mathbf{A}_N \mathbf{x}_N$

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perturb  $\mathbf{b}$  to  $\mathbf{b} + \vec{\epsilon}_*$  with small  $\epsilon > 0$ ,  $\vec{\epsilon} = (1, \epsilon, \epsilon^2, \dots, \epsilon^m)^\top$

$$\mathbf{A}\mathbf{x} = [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon}$$

$$\mathbf{A}_B \mathbf{x}_B + \mathbf{A}_N \mathbf{x}_N = [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon}$$

$$\mathbf{A}_B \mathbf{x}_B = [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} - \mathbf{A}_N \mathbf{x}_N$$

dictionary  $\mathbf{x}_B = \mathbf{A}_B^{-1} [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} - \mathbf{A}_B^{-1} \mathbf{A}_N \mathbf{x}_N$

## Lexicographic degeneracy resolution

perturb  $\mathbf{b}$  to  $\mathbf{b} + \vec{\epsilon}_*$  with small  $\epsilon > 0$ ,  $\vec{\epsilon} = (1, \epsilon, \epsilon^2, \dots, \epsilon^m)^\top$

$$\begin{aligned} \mathbf{A}\mathbf{x} &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} \\ \mathbf{A}_B \mathbf{x}_B + \mathbf{A}_N \mathbf{x}_N &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} \\ \mathbf{A}_B \mathbf{x}_B &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} - \mathbf{A}_N \mathbf{x}_N \\ \mathbf{x}_B &= \mathbf{A}_B^{-1} [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} - \mathbf{A}_B^{-1} \mathbf{A}_N \mathbf{x}_N \\ \mathbf{x}_B &= [\mathbf{A}_B^{-1} \mathbf{b} \mid \mathbf{A}_B^{-1}] \vec{\epsilon} - \mathbf{A}_B^{-1} \mathbf{A}_N \mathbf{x}_N \end{aligned}$$

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perturb  $\mathbf{b}$  to  $\mathbf{b} + \vec{\epsilon}_*$  with small  $\epsilon > 0$ ,  $\vec{\epsilon} = (1, \epsilon, \epsilon^2, \dots, \epsilon^m)^\top$

$$\begin{aligned}
 \mathbf{Ax} &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} \\
 \mathbf{A}_B \mathbf{x}_B + \mathbf{A}_N \mathbf{x}_N &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} \\
 \mathbf{A}_B \mathbf{x}_B &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} - \mathbf{A}_N \mathbf{x}_N \\
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 \end{aligned}$$

nondegeneracy  $\Leftrightarrow \mathbf{x}_B > \mathbf{0}$  for small  $\epsilon > 0 \Leftrightarrow [\mathbf{A}_B^{-1} \mathbf{b} \mid \mathbf{A}_B^{-1}]$   
**lexico-positive** (first nonzero element in each row is  $> 0$ ).

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perturb  $\mathbf{b}$  to  $\mathbf{b} + \vec{\epsilon}_*$  with small  $\epsilon > 0$ ,  $\vec{\epsilon} = (1, \epsilon, \epsilon^2, \dots, \epsilon^m)^\top$

$$\begin{aligned}
 \mathbf{A}\mathbf{x} &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} \\
 \mathbf{A}_B \mathbf{x}_B + \mathbf{A}_N \mathbf{x}_N &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} \\
 \mathbf{A}_B \mathbf{x}_B &= [\mathbf{b} \mid \mathbf{I}] \vec{\epsilon} - \mathbf{A}_N \mathbf{x}_N \\
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**lexico-positive** (first nonzero element in each row is  $> 0$ ).

**Example:**

$$\begin{bmatrix} 1 & -9 & 4 & 0 \\ 0 & 3 & -100 & 2 \\ 0 & 0 & 0 & 5 \end{bmatrix} \vec{\epsilon} = \begin{bmatrix} 1 & -9\epsilon + 4\epsilon^2 \\ 3\epsilon - 100\epsilon^2 + 2\epsilon^3 \\ 5\epsilon^3 \end{bmatrix}$$



## Integer pivoting for $Cz + s = 1$

(basic columns in red)

	$z_1$	$z_2$	$s_1$	$s_2$	RHS
	4	3	1	0	1
$z_1$ enters, $s_2$ leaves	7	2	0	1	1

## Integer pivoting for $Cz + s = 1$

(basic columns in red)	$z_1$	$z_2$	$s_1$	$s_2$	RHS	
	4	3	1	0	1	$\times 7$
$z_1$ enters, $s_2$ leaves	7	2	0	1	1	
	28	21	7	0	7	
	7	2	0	1	1	

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	28	21	7	0	7	
	7	2	0	1	1	
$z_2$ enters, $s_1$ leaves	0	13	7	-4	3	
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$z_2$ enters, $s_1$ leaves	0	13	7	-4	3	
	7	2	0	1	1	$\times 13$
	0	13	7	-4	3	
	91	26	0	13	13	

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(basic columns in red)	$z_1$	$z_2$	$s_1$	$s_2$	RHS	
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	7	2	0	1	1	$\times 13$
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	91	26	0	13	13	
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(numbers grow)	91	0	-14	21	7	

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(numbers grow)	91	0	-14	21	7	$/ 7$
	0	13	7	-4	3	
	13	0	-2	3	1	