

# Algorithmic game theory

Martin Balko

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# Nash equilibria in bimatrix games

# What we learned last time

- Last lecture we introduced a **brute-force algorithm** to find all Nash equilibria in bimatrix games.
- Recall that we have payoff matrices  $M$  and  $N$  with  $M_{i,j} = u_1(i,j)$  and  $N_{i,j} = u_2(i,j)$ .
- The **best response condition**: If  $x$  and  $y$  are mixed strategy vectors of players 1 and 2, respectively, then  $x$  is a **best response to  $y$**  iff

$$\forall i \in A_1 : x_i > 0 \implies M_{i,*}y = \max\{M_{k,*}y : k \in A_1\}.$$

Analogously,  $y$  is a **best response to  $x$**  iff

$$\forall j \in A_2 : y_j > 0 \implies N_{j,*}^\top x = \max\{N_{k,*}^\top x : k \in A_2\}.$$

- Today, we reveal a **geometric structure** behind finding NE in bimatrix games and show one of the **fastest known algorithms** for this task.

# Best response polyhedra

- To reveal the geometric structure, we define labeled polyhedra for a bimatrix game  $G$ .
- The best response polyhedron for player 1 in  $G$  is a polyhedron

$$\bar{P} = \{(x, v) \in \mathbb{R}^m \times \mathbb{R} : x \geq \mathbf{0}, \mathbf{1}^\top x = 1, N^\top x \leq \mathbf{1}v\}.$$

Similarly, we define the best response polyhedron for player 2 in  $G$  as

$$\bar{Q} = \{(y, u) \in \mathbb{R}^n \times \mathbb{R} : y \geq \mathbf{0}, \mathbf{1}^\top y = 1, My \leq \mathbf{1}u\}.$$

- Let  $(x, v) \in \bar{P}$ ,  $(y, u) \in \bar{Q}$  and let  $s = (s_1, s_2)$  be a strategy profile with mixed strategy vectors  $x$  and  $y$ .
- Then,  $My \leq \mathbf{1}u$  says that  $u_1(s) \leq u$ , as the expected payoff is always

$$u_1(s) = x^\top My \leq x^\top \mathbf{1}u = u \sum_{i \in A_1} x_i = u.$$

- Analogously, the condition  $N^\top x \leq \mathbf{1}v$  says that  $u_2(s) \leq v$ .
- Thus, points of  $\bar{P}$  and  $\bar{Q}$  are the mixed strategies with the “upper envelope” of expected payoffs of the other player.

# Labelings of the Best response polyhedra

- Now, we define the labels of points from the best response polyhedra

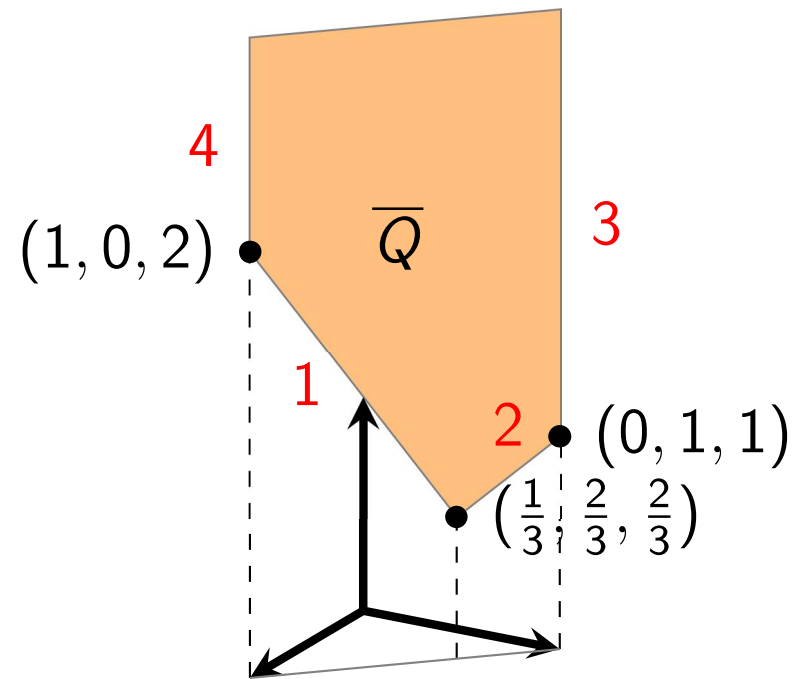
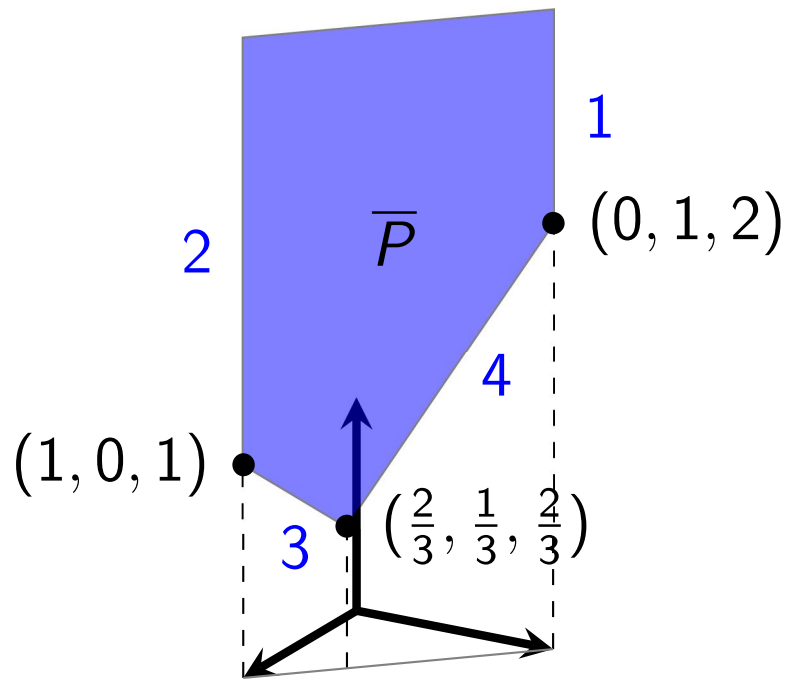
$$\bar{P} = \{(x, v) \in \mathbb{R}^m \times \mathbb{R} : x \geq \mathbf{0}, \mathbf{1}^\top x = 1, N^\top x \leq \mathbf{1}v\}.$$

and

$$\bar{Q} = \{(y, u) \in \mathbb{R}^n \times \mathbb{R} : y \geq \mathbf{0}, \mathbf{1}^\top y = 1, My \leq \mathbf{1}u\}.$$

- We say that a point  $(x, v) \in \bar{P}$  has label  $i \in A_1 \cup A_2$  if either  $i \in A_1$  and  $x_i = 0$  or if  $i \in A_2$  and  $N_{i,*}^\top x = v$ .
  - That is, if the inequality in the definition of  $\bar{P}$  is binding ( $i \in A_1$  is not in the support or  $i \in A_2$  is a best response to  $x$ ).
- Similarly, a point  $(y, u) \in \bar{Q}$  has a label  $i \in A_1 \cup A_2$  if either  $i \in A_1$  and  $M_{i,*}y = u$ , or if  $i \in A_2$  and  $y_i = 0$ .
- Each point from  $\bar{P}$  or  $\bar{Q}$  may have more labels.

# Best response polyhedra $\bar{P}$ and $\bar{Q}$ for the Battle of sexes



$$\bar{P} = \{(x_1, x_2, v) \in \mathbb{R}^2 \times \mathbb{R} : x_1, x_2 \geq 0, x_1 + x_2 = 1, x_1 \leq v, 2x_2 \leq v\}$$

$$\bar{Q} = \{(y_3, y_4, u) \in \mathbb{R}^2 \times \mathbb{R} : y_3, y_4 \geq 0, y_3 + y_4 = 1, 2y_3 \leq u, y_4 \leq u\}.$$

# NE and Best response polyhedra

- What are the labels (and Best response polyhedra) for?
- They help us identify NE in  $G$ !

## Proposition 2.27

A strategy profile  $s$  is NE of  $G$  iff the pair  $((x, v), (y, u)) \in \bar{P} \times \bar{Q}$  is **completely labeled**, that is, every label  $i \in A_1 \cup A_2$  appears as a label of either  $(x, v)$  or  $(y, u)$ .

- **Proof:** By the **Best response condition**, for every player  $i \in P$ , a mixed strategy  $s_i$  is a best response to  $s_{-i}$  if and only if all pure strategies in the support of  $s_i$  are best responses to  $s_{-i}$ .
- A **missing label**  $i \in A_1$  means that a pure strategy  $i \in \text{Supp}(s_1)$  ( $x_i > 0$ ) was not a best response ( $M_{i,*}y < u$ ). Analogously for  $i \in A_2$ . Then,  $s$  is not NE.
- If **all labels appear**, then  $s_1$  and  $s_2$  are mutually best responses, as each pure strategy is a best response or is not in the support. Then,  $s$  is NE. □

# Best response polytopes

- That is nice. But we will make it even nicer!
- The best response polyhedra have some unnecessary complications (they are unbounded and have  $u$  and  $v$  in their coordinates). We get rid of these under certain mild assumptions.
- We assume that  $M$  and  $N^\top$  are non-negative and have no zero column. (simply add a large constant to the payoffs)
- Then, we can divide each inequality  $N_{i,*}^\top x \leq v$  with  $v$ , treating  $x_i/v$  as a new variable, and do the same for  $\bar{Q}$ . This normalizes the payoffs to 1 and we get the following polytopes.
- The (normalized) best response polytope for player 1 in  $G$  is a polytope

$$P = \{x \in \mathbb{R}^m : x \geq \mathbf{0}, N^\top x \leq \mathbf{1}\}.$$

Similarly, the best response polytope for player 2 in  $G$  is a polytope

$$Q = \{y \in \mathbb{R}^n : y \geq \mathbf{0}, My \leq \mathbf{1}\}.$$



# What did we get?

- The inequalities have the **same meaning**: if  $x_i \geq 0$  is binding, then  $i \in A_1$  is not in the support and if  $N_{j,*}^\top x \leq 1$  is binding, then  $j \in A_2$  is a best response to  $s_1$ . Analogously for  $Q$ .
- From the assumption on  $M$  and  $N$ , the polyhedra  $P$  and  $Q$  are bounded (are **polytopes**) and have **dimensions  $m$  and  $n$** .
- **Disadvantage**: coordinates of  $x$  and  $y$  do not sum up to 1. We can rescale to  $x/(\mathbf{1}^\top x)$  and  $y/(\mathbf{1}^\top y)$ .
- The polytope  $\bar{P}$  is in a one-to-one correspondence with  $P \setminus \{\mathbf{0}\}$  under the **projective transformation**  $(x, v) \mapsto x/v$ . Similarly  $\bar{Q}$  and  $Q \setminus \{\mathbf{0}\}$ .
  - Projective transformations preserve incidences, so the **labels stay the same**.

## Corollary 2.30

A strategy profile  $(s_1, s_2)$  with mixed strategy vectors  $x$  and  $y$  is NE of  $G$  if and only if the point  $(x/u_2(s), y/u_1(s)) \in P \times Q \setminus \{(\mathbf{0}, \mathbf{0})\}$  is completely labeled. □

# NE in nondegenerate games

- Recall that a bimatrix game is **nondegenerate** if there are at most  $k$  pure best responses to every mixed strategy with support of size  $k$ .
- In these games NE correspond to pairs of completely labeled **vertices**.
  - Since  $G$  is nondegenerate, **each point of  $P$  has at most  $m$  labels**. This is because if  $x$  has support of size  $k$ , then  $x$  has at most  $m - k$  labels in  $A_1$  and so if  $x$  had more than  $m$  labels, then  $x$  would have more than  $k$  best responses in  $A_2$ . Analogously, **each point of  $Q$  has at most  $n$  labels**.
  - Thus,  $P$  and  $Q$  are both **simple polytopes** (each point of  $P$  or  $Q$  contained in more than  $m$  or  $n$  facets has more than  $m$  or  $n$  labels).
  - Since  $\dim(P) = m$  and  $\dim(Q) = n$ , only vertices of  $P$  and  $Q$  can have  $m$  and  $n$  labels.
  - By **Corollary 2.30**, only vertices of  $P$  and  $Q$  can be NE.
- $\Rightarrow$  **Algorithm for finding NE**: check all pairs of vertices and their labels!

# Algorithm for finding NE with vertex enumeration

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**Algorithm 0.4:** VERTEX ENUMERATION( $G$ )

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*Input* : A nondegenerate bimatrix game  $G$ .

*Output* : All Nash equilibria of  $G$ .

**for** each pair  $(x, y)$  of vertices from  $(P \setminus \{0\}) \times (Q \setminus \{0\})$   
  { if  $(x, y)$  is completely labeled,  
  { then return  $(x/(\mathbf{1}^\top x), y/(\mathbf{1}^\top y))$  as a Nash equilibrium

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- All vertices of a simple polytope in  $\mathbb{R}^d$  with  $v$  vertices and  $N$  defining inequalities can be found in time  $O(dNv)$  (Avis and Fukuda).
- However, if  $m = n$ , the best response polytopes can have  $c^n$  vertices for some constant  $c$  with  $1 < c < 2.9$ .

# Polytopes can be weird and complex!

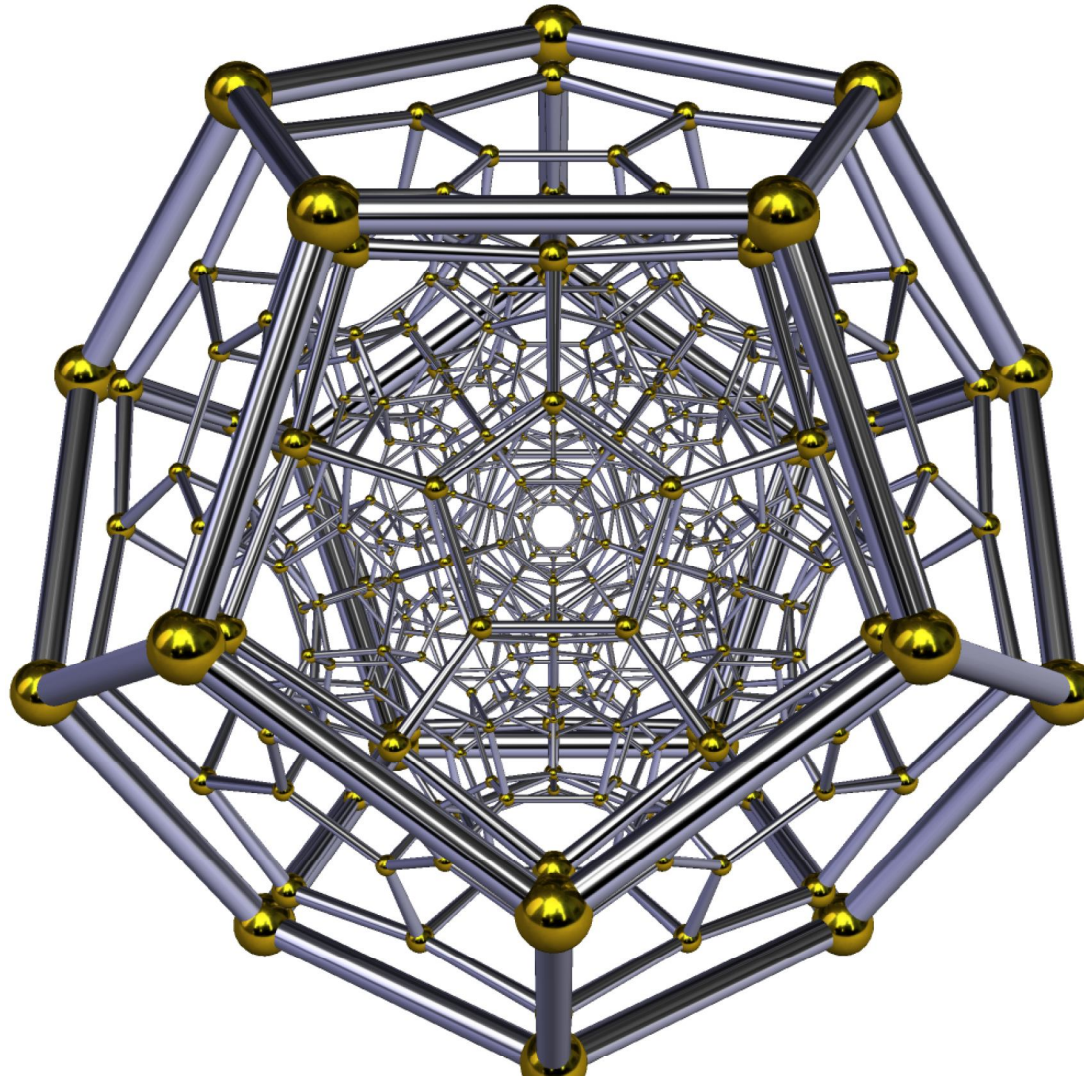


Figure: Schlegel diagram for the 120-cell.

# Algorithm for finding NE with vertex enumeration

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## Algorithm 0.6: VERTEX ENUMERATION( $G$ )

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*Input* : A nondegenerate bimatrix game  $G$ .

*Output* : All Nash equilibria of  $G$ .

**for** each pair  $(\mathbf{x}, \mathbf{y})$  of vertices from  $(P \setminus \{\mathbf{0}\}) \times (Q \setminus \{\mathbf{0}\})$   
    { if  $(\mathbf{x}, \mathbf{y})$  is completely labeled,  
    { then return  $(\mathbf{x}/(\mathbf{1}^\top \mathbf{x}), \mathbf{y}/(\mathbf{1}^\top \mathbf{y}))$  as a Nash equilibrium

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- All vertices of a simple polytope in  $\mathbb{R}^d$  with  $v$  vertices and  $N$  defining inequalities can be found in time  $O(dNv)$  (Avis and Fukuda).
- However, if  $m = n$ , the best response polytopes can have  $c^n$  vertices for some constant  $c$  with  $1 < c < 2.9$ .
- We can speed up the search by performing a walk on  $(P \setminus \{\mathbf{0}\}) \times (Q \setminus \{\mathbf{0}\})$  guided by the labels.

# The Lemke–Howson algorithm

- One of the best algorithms for finding NE in bimatrix games.
- Discovered by [Lemke](#) and [Howson](#) in 1964.

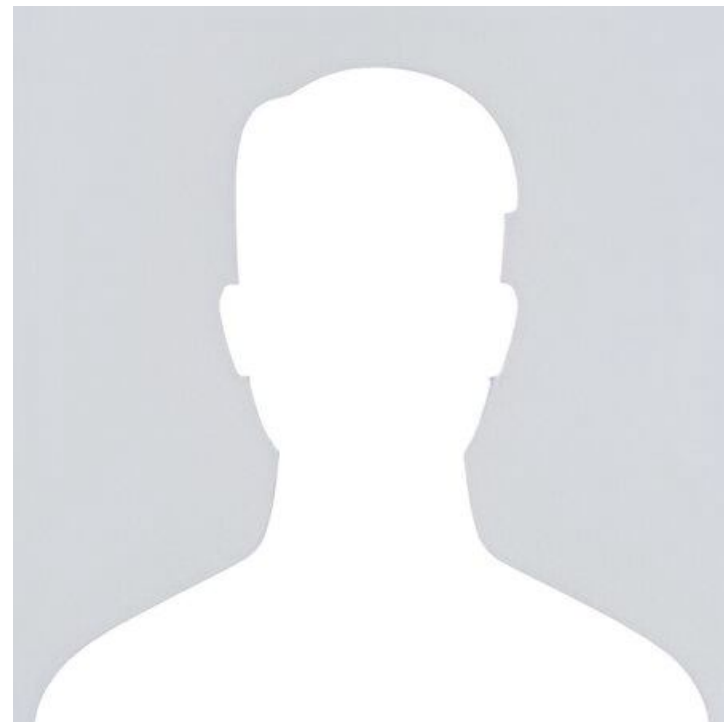


Figure: [Carlton E. Lemke](#) (1920–2004) and [J. T. Howson](#) (1937–2022).

Source: <https://oldurls.inf.ethz.ch>

# The Lemke–Howson algorithm explained

- Since  $P$  is simple and  $\dim(P) = m$ , its vertices are incident to exactly  $m$  facets and  $m$  edges. So each **vertex of  $P$  has  $m$  labels** and each **edge of  $P$  has  $m - 1$  labels**. Similarly for  $Q$  and  $n$ .
- **Dropping a label  $l \in A_1 \cup A_2$**  in a vertex  $x$  of  $P$  means traversing the unique edge of  $P$  that is incident to  $x$  and has all  $m$  labels that  $x$  has besides  $l$ . The other endpoint of this edge has the same labels as  $x$ , only  $l$  is replaced with a new label, which is said to be **picked up**. Dropping and picking up a label in vertices of  $Q$  is defined analogously.
- The algorithm starts at  $(\mathbf{0}, \mathbf{0})$  and **alternately follows edges of  $P$  and  $Q$** .
- At the **first step**, it chooses a label  $k \in A_1 \cup A_2$  and drops it. Then, a new label  $l$  is picked up. This label  $l$  has a **duplicate** in the other polytope. We drop the duplicate of  $l$  in the other polytope in the next step, which leads to picking up a new label  $l'$ . We iterate and stop when  $l' = k$ .
- Duplicate label is either a new best response, which gets a positive probability, or a pure strategy whose probability became 0 and we move away from its best response facet.



# The Lemke–Howson algorithm: pseudocode

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**Algorithm 0.8:** LEMKE–HOWSON( $G$ )

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*Input* : A nondegenerate bimatrix game  $G$ .

*Output* : One Nash equilibrium of  $G$ .

$(x, y) \leftarrow (\mathbf{0}, \mathbf{0}) \in \mathbb{R}^m \times \mathbb{R}^n$ ,

$k \leftarrow$  arbitrary label from  $A_1 \cup A_2$ ,  $l \leftarrow k$ ,

**while** (*true*)

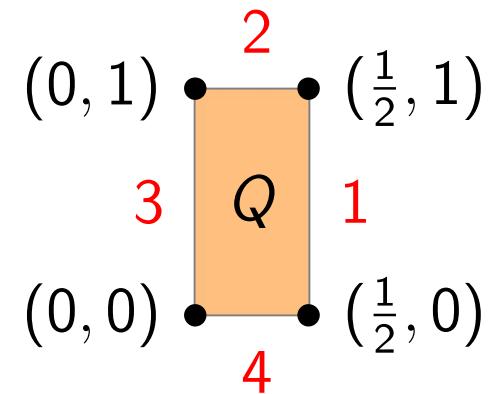
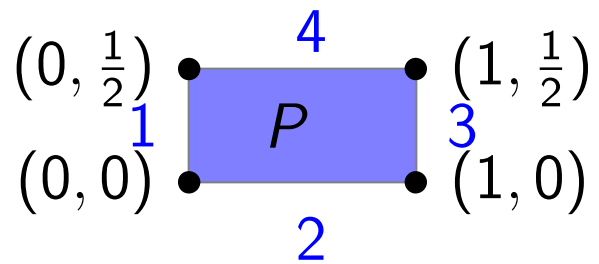
**do** {  
    In  $P$ , drop  $l$  from  $x$  and redefine  $x$  as the new vertex,  
    redefine  $l$  as the newly picked up label. Switch to  $Q$ .  
    If  $l = k$ , stop looping.  
    In  $Q$ , drop  $l$  from  $y$  and redefine  $y$  as the new vertex,  
    redefine  $l$  as the newly picked up label. Switch to  $P$ .  
    If  $l = k$ , stop looping.

Output  $(x/(\mathbf{1}^\top x), y/(\mathbf{1}^\top y))$ .

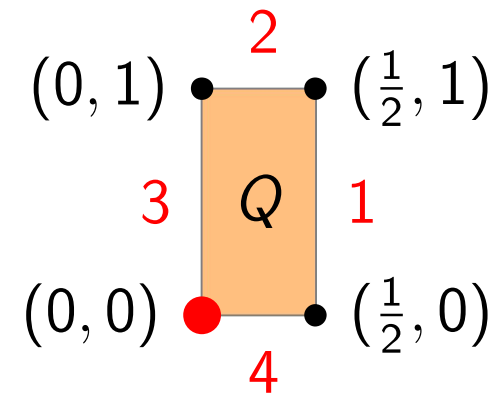
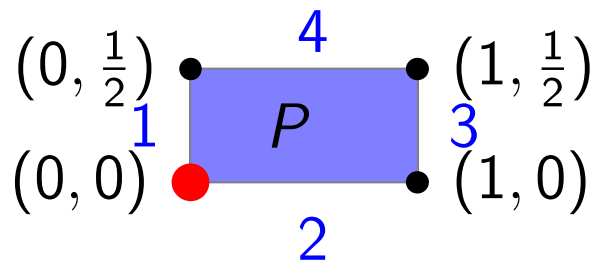
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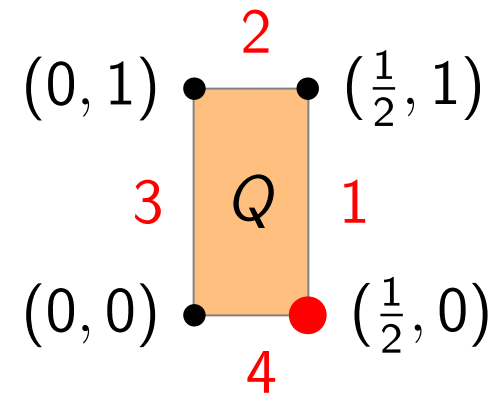
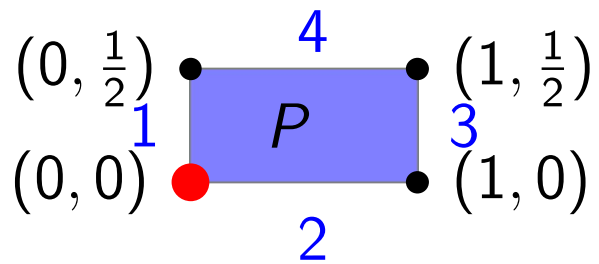
# Lemke–Howson on the Battle of sexes ( $k = 3$ )



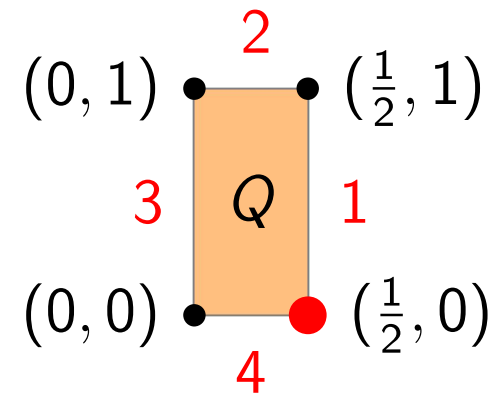
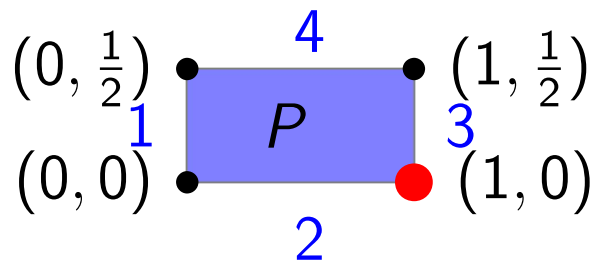
# Lemke–Howson on the Battle of sexes ( $k = 3$ )



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# Correctness of the Lemke–Howson algorithm I

## Proposition 2.31

The Lemke–Howson algorithm stops after a finite number of steps and outputs mixed strategy vectors of NE in  $G$ .

- **Proof:** Let  $k$  be the label chosen in the first step.
- We define a **configuration graph**  $\mathcal{G}$  with the vertices formed by pairs  $(x, y)$  of vertices from  $P \times Q$  that are  **$k$ -almost completely labeled** (every label from  $A_1 \cup A_2 \setminus \{k\}$  is a label of  $x$  or  $y$ ). A pair  $\{(x, y), (x', y')\}$  is an edge of  $\mathcal{G}$  if  **$(x = x' \ \& \ yy' \in E(Q))$**  or  **$(xx' \in E(P) \ \& \ y = y')$** . Clearly,  $\mathcal{G}$  is finite.
- $\mathcal{G}$  has degrees only 1 or 2 ( $\mathcal{G}$  is a disjoint union of paths and cycles).
  - If  **$(x, y)$  has all labels** from  $A_1 \cup A_2$ , then  $(x, y)$  is connected to exactly one other vertex, as exactly one of  $x$  and  $y$  has the label  $k$  and we can drop  $k$  only from this one vertex.
  - Otherwise,  $(x, y)$  has all labels from  $A_1 \cup A_2 \setminus \{k\}$  and there is a **unique label shared by  $x$  and  $y$** . Then  $(x, y)$  is adjacent to two vertices, as we can drop the duplicate label from  $x$  in  $P$  or  $y$  in  $Q$ .

# Correctness of the Lemke–Howson algorithm II

- The Lemke–Howson algorithm **starts at the leaf**  $(\mathbf{0}, \mathbf{0})$  of a path in  $\mathcal{G}$ .
- Then it walks along this path and does not visit any vertex of the configuration graph twice.
  - The next vertex pair on the path is always unique (we move to the other endpoint of the unique edge that contains the current vertex and corresponds to the dropped label).
  - Visiting the same vertex twice would give a vertex of degree larger than 2 in  $\mathcal{G}$  since we started at a leaf.
- Thus, the **algorithm terminates** after a finite number of steps in the other leaf  $(x^*, y^*)$  of the path. Since  $(x^*, y^*)$  is a leaf in the configuration graph, it is **completely labeled**.
- This endpoint is not of the form  $(x, \mathbf{0})$  or  $(\mathbf{0}, y)$  (**Exercise**).
- By **Corollary 2.30**,  $(x^*, y^*)$  corresponds to **NE** after rescaling. □

# The Lemke–Howson algorithm: remarks

- By the **Degree sum formula**, the number of vertices of degree 1 is even.

## Corollary 2.32

A nondegenerate bimatrix game has an odd number of NE.

- Degenerate games can have infinite number of NE.
- The Lemke–Howson algorithm finds only a single NE.
- The algorithm can be implemented using so-called **complementary pivoting** (we will see it at the **tutorials**).
- The running time can still be **exponential** ( $O(2^n)$  steps for  $n = m$ )! It performs well in practise (polynomial on uniformly random games).
- **Is there an efficient algorithm to find NE?**

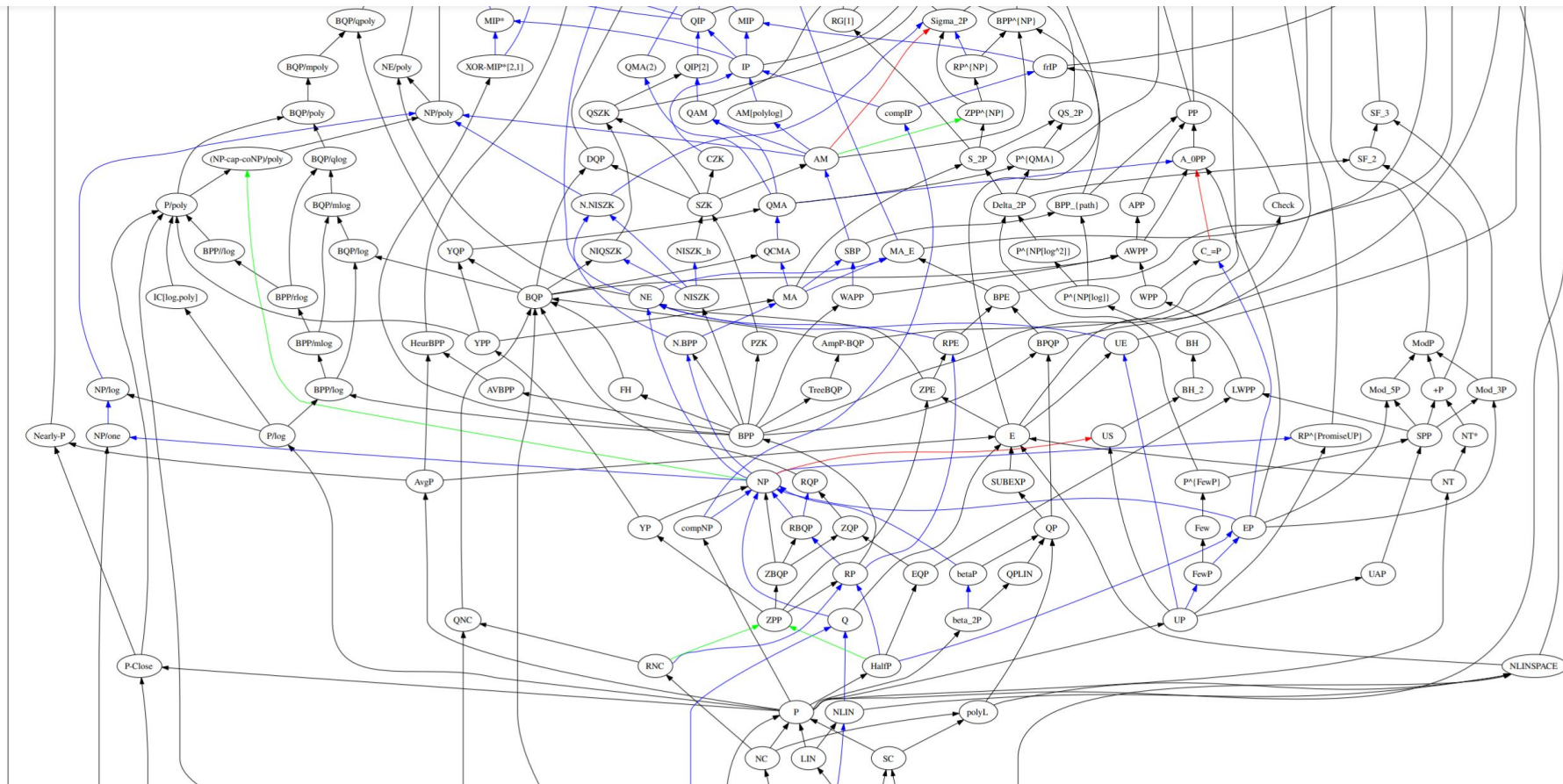


Figure: A view on the complexity classes classification.

Source: [https://complexityzoo.uwaterloo.ca/Complexity\\_Zoo](https://complexityzoo.uwaterloo.ca/Complexity_Zoo)

- “**P=NP**” is one of the most important problems in computer science. The website <https://www.win.tue.nl/~gwoegi/P-versus-NP.htm> contains a collection of over 100 attempts to solve it.

Thank you for your attention.