

# Computing Solution Concepts - Intro



# Solution Concepts

- In single agent settings, there is the notion of *optimal strategy*
- In multiagent setting, situation is more complex. Best strategy depends on the strategies of other agents
- *Solution concepts* – certain subsets of outcomes that are interesting
- Pareto optimality, Nash equilibrium



# Computational Concerns

- How to compute a Nash equilibrium?
- Examples we've seen had 2 players, each with 2 actions
- Complexity depends on class of games considered
- 2 player zero-sum games
- 2 player general-sum games
- $n$  players,  $n > 2$
- Other solution concepts



# Computing Nash Equilibrium in 2- player, zero-sum games



# Setup

- Consider 2-player, zero-sum game:

$$G = (\{1, 2\}, A_1 \times A_2, (u_1, u_2))$$

- Let  $U_i^*$  be the equilibrium value of player  $I$
- Recall in a N.E., player 1's value is equal to his maxmin value

**Definition 3.4.1 (Maxmin)** *The maxmin strategy for player  $i$  is  $\arg \max_{s_i} \min_{s_{-i}} u_i(s_i, s_{-i})$ , and the maxmin value for player  $i$  is  $\max_{s_i} \min_{s_{-i}} u_i(s_i, s_{-i})$ .*

- Use this fact to write an LP



# Linear Program

$$\text{minimize } U_1^* \quad (4.1)$$

$$\text{subject to } \sum_{k \in A_2} u_1(a_1^j, a_2^k) \cdot s_2^k \leq U_1^* \quad \forall j \in A_1 \quad (4.2)$$

$$\sum_{k \in A_2} s_2^k = 1 \quad (4.3)$$

$$s_2^k \geq 0 \quad \forall k \in A_2 \quad (4.4)$$

- Variables:  $U_1^*$ ,  $s_2^k$



# Dual Program

$$\text{maximize } U_1^* \quad (4.5)$$

$$\text{subject to } \sum_{j \in A_1} u_1(a_1^j, a_2^k) \cdot s_1^j \geq U_1^* \quad \forall k \in A_2 \quad (4.6)$$

$$\sum_{j \in A_1} s_1^j = 1 \quad (4.7)$$

$$s_1^j \geq 0 \quad \forall j \in A_1 \quad (4.8)$$



# Reformulation with Slack Variables

$$\text{minimize } U_1^* \quad (4.9)$$

$$\text{subject to } \sum_{k \in A_2} u_1(a_1^j, a_2^k) \cdot s_2^k + r_1^j = U_1^* \quad \forall j \in A_1 \quad (4.10)$$

$$\sum_{k \in A_2} s_2^k = 1 \quad (4.11)$$

$$s_2^k \geq 0 \quad \forall k \in A_2 \quad (4.12)$$

$$r_1^j \geq 0 \quad \forall j \in A_1 \quad (4.13)$$



# Complexity of Computing a Nash Equilibrium



# Computing Nash Equilibria

- 2-player, zero-sum in poly
- 2-player, general sum?
- Cannot be formulated as LP, players not diametrically opposed
- No known reduction from NP-complete problem
- Stumbling block with NP: decision problems. But we always know a NE exists



# The PPAD class

- So current knowledge about NE computation is in relation to PPAD class
- PPAD – “Polynomial Parity Argument, Directed Version”
- Family of directed graphs  $G(n)$
- Computational task is finding a source or sink node



# The family $G(n)$

- Defined on set  $N$  of  $2^n$  nodes, but described in polynomial space
- Just encode set of edges
- *Parent, Child* functions from  $N$  to  $N$ : encoded as arithmetic circuits with sizes poly. in  $n$
- An edge exists from node  $j$  to  $k$  iff.  $Parent(k) = j$  and  $Child(j) = k$
- There must exist one distinguished node  $0$  with exactly zero parents
- Find sink or source other than  $0$  in a given graph



# Complexity

**Theorem 4.2.1** *The problem of finding a sample Nash equilibrium of a general-sum finite game with two or more players is PPAD-complete.*

- CNE is in PPAD and any other problem in PPAD can be reduced to it
- CNE is in PPAD reduction proceeds quite directly from the proof in the textbook that every game has a NE that uses Sperner's lemma
- Harder part is showing CNE is PPAD-hard. Result was proven in 2005, a culmination of intermediate results achieved over a decade



# Complexity

**Theorem 4.2.1** *The problem of finding a sample Nash equilibrium of a general-sum finite game with two or more players is PPAD-complete.*

- Not known if  $P=PPAD$ . Generally believed not
- It is known that finding an NE in 2 player games is no easier than finding an NE in  $n$  player games
- Finding a NE is no easier than finding an arbitrary Brouwer fixed point



# LCP Formulation of 2-player NE



# Computing Nash Equilibria

- 2-player, zero-sum in poly
- 2-player, general sum?
- Cannot be formulated as LP, players not diametrically opposed
- The LCP formulation



# The LCP Formulation

$$\sum_{k \in A_2} u_1(a_1^j, a_2^k) \cdot s_2^k + r_1^j = U_1^* \quad \forall j \in A_1 \quad (4.14)$$

$$\sum_{j \in A_1} u_2(a_1^j, a_2^k) \cdot s_1^j + r_2^k = U_2^* \quad \forall k \in A_2 \quad (4.15)$$

$$\sum_{j \in A_1} s_1^j = 1, \quad \sum_{k \in A_2} s_2^k = 1 \quad (4.16)$$

$$s_1^j \geq 0, \quad s_2^k \geq 0 \quad \forall j \in A_1, \forall k \in A_2 \quad (4.17)$$

$$r_1^j \geq 0, \quad r_2^k \geq 0 \quad \forall j \in A_1, \forall k \in A_2 \quad (4.18)$$

$$r_1^j \cdot s_1^j = 0, \quad r_2^k \cdot s_2^k = 0 \quad \forall j \in A_1, \forall k \in A_2 \quad (4.19)$$



# The Lemke-Howson Algorithm



# Lemke-Howson Algorithm

- 2-player, general sum games
- Algorithm is for solving linear complementarity programs
- Searches vertices of strategy simplices (like the simplex algorithm for solving LPs)



# Lemke-Howson – a graphical exposition

|      |      |
|------|------|
| 0, 1 | 6, 0 |
| 2, 0 | 5, 2 |
| 3, 4 | 3, 3 |

Figure 4.1: A game for the exposition of the Lemke–Howson algorithm.

# Lemke-Howson – a graphical exposition

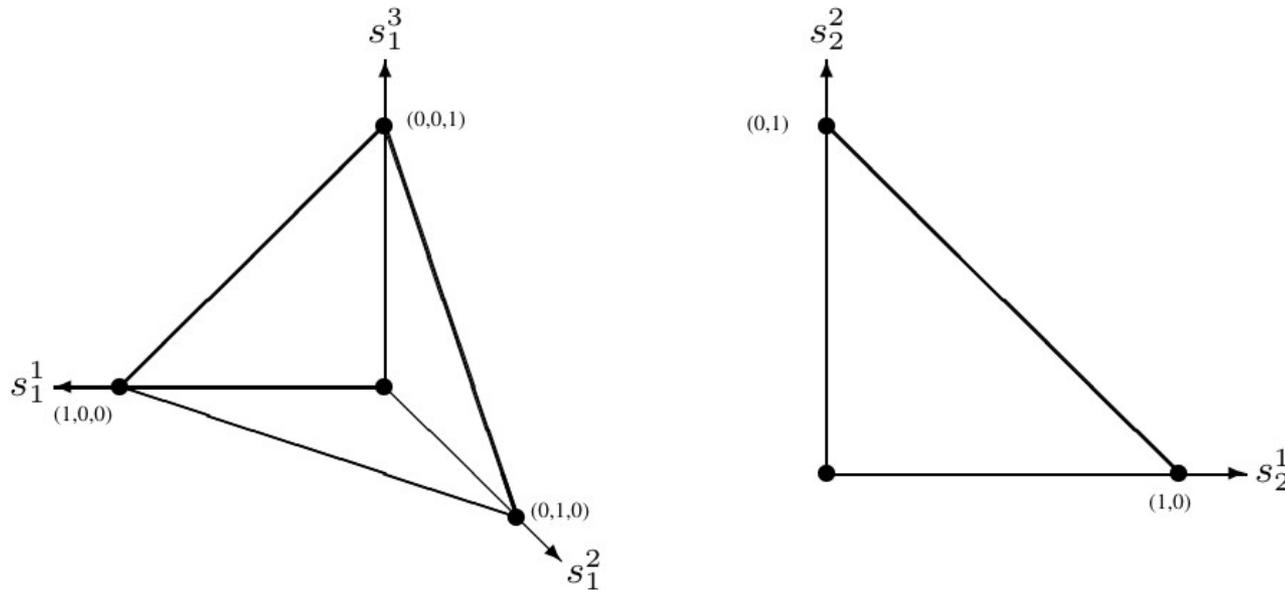


Figure 4.2: Strategy spaces for player 1 (left) and player 2 (right) in the game from Figure 4.1.

# Lemke-Howson – a graphical exposition

Our next step in defining the Lemke–Howson algorithm is to define a labeling on the strategies. Every possible mixed strategy  $s_i$  is given a set of labels  $L(s_i^j) \subseteq A_1 \cup A_2$  drawn from the set of available actions for both players. Denoting a given player as  $i$  and the other player as  $-i$ , mixed strategy  $s_i$  for player  $i$  is labeled as follows:

- with each of player  $i$ 's actions  $a_i^j$  that is *not* in the support of  $s_i$ ; and
- with each of player  $-i$ 's actions  $a_{-i}^j$  that *is* a best response by player  $-i$  to  $s_i$ .



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- 
- A strategy profile is a Nash equilibrium iff. it is completely labeled



# Lemke-Howson – a graphical exposition

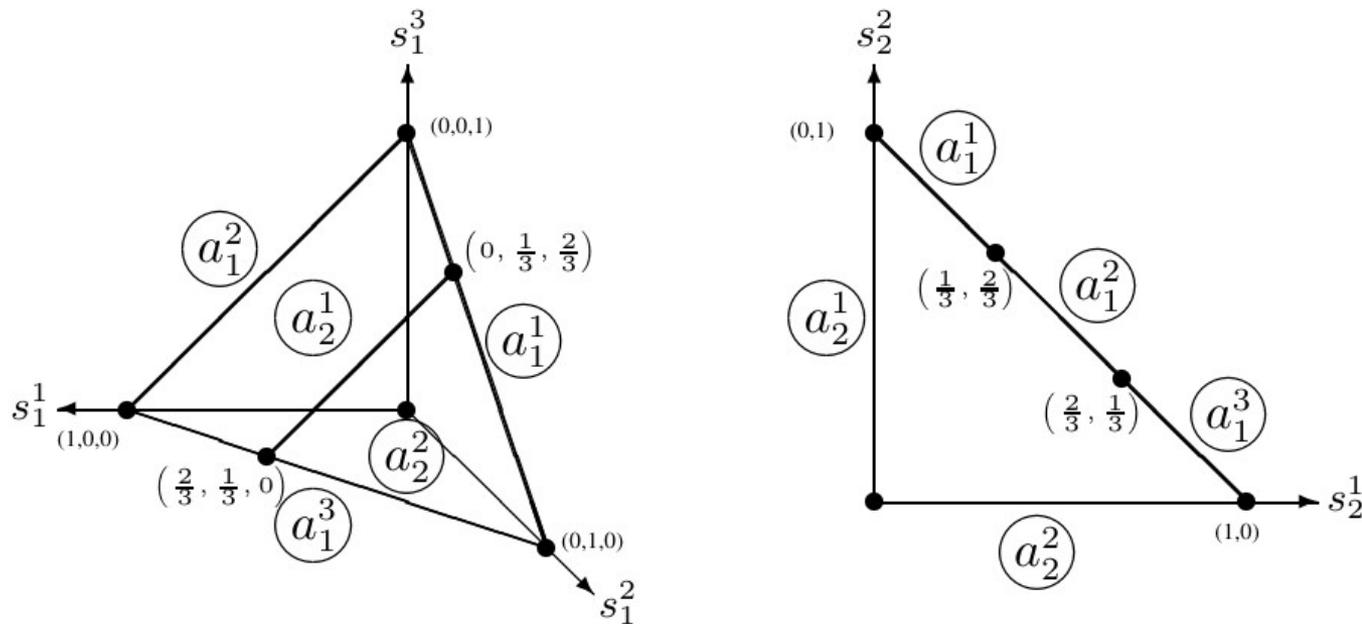
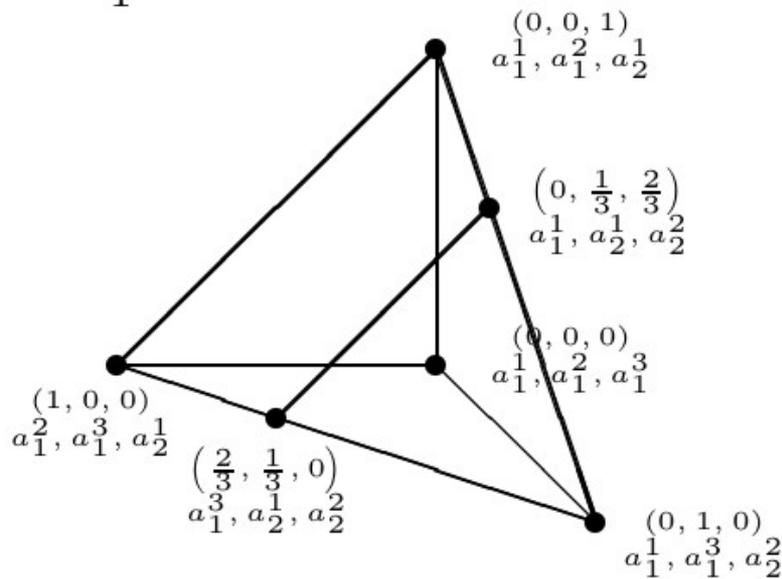


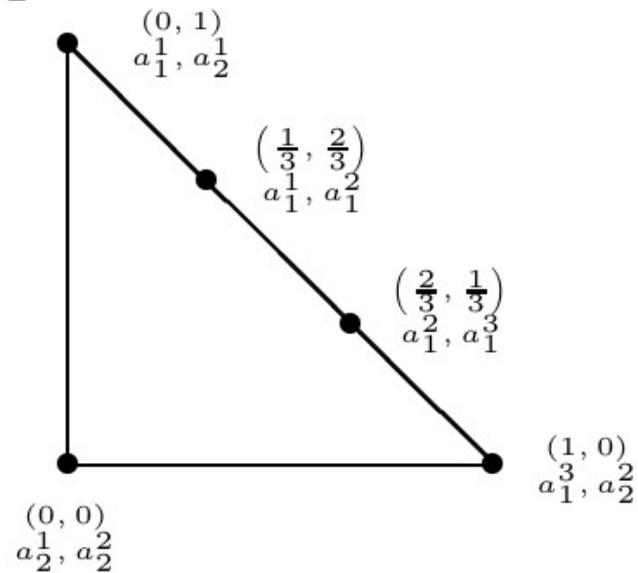
Figure 4.3: Labeled strategy spaces for player 1 (left) and player 2 (right) in the game from Figure 4.1.

# Lemke-Howson – a graphical exposition

$G_1$ :



$G_2$ :



# Lemke-Howson – Properties

- Guaranteed to find a NE
- Alternative proof of the existence of NE
- Path after initial move is unique. Only nondeterminism is in first move
- All paths from the starting point to a NE can be exponential (algorithm is provably exponential)
- No way to assess how close we are to a NE

