

Minmax and Dominance

CPSC 532A Lecture 6

September 28, 2006

Lecture Overview

Recap

Maxmin and Minmax

Linear Programming

Computing

Fun Game

Domination

What are solution concepts?

- ▶ **Solution concept**: a subset of the outcomes in the game that are somehow interesting.
- ▶ There is an implicit computational problem of finding these outcomes given a particular game.
- ▶ Depending on the concept, existence can be an issue.

Solution concepts we've seen so far:

What are solution concepts?

- ▶ **Solution concept**: a subset of the outcomes in the game that are somehow interesting.
- ▶ There is an implicit computational problem of finding these outcomes given a particular game.
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Solution concepts we've seen so far:

- ▶ Pareto-optimal outcome
- ▶ Pure-strategy Nash equilibrium
- ▶ Mixed-strategy Nash equilibrium
- ▶ Other Nash variants:
 - ▶ **weak** Nash equilibrium
 - ▶ **strict** Nash equilibrium

Mixed Strategies

- ▶ It would be a pretty bad idea to play any deterministic strategy in matching pennies
- ▶ Idea: confuse the opponent by playing **randomly**
- ▶ Define a **strategy** s_i for agent i as any probability distribution over the actions A_i .
 - ▶ **pure strategy**: only one action is played with positive probability
 - ▶ **mixed strategy**: more than one action is played with positive probability
 - ▶ these actions are called the **support** of the mixed strategy
- ▶ Let the set of **all strategies** for i be S_i
- ▶ Let the set of **all strategy profiles** be $S = S_1 \times \dots \times S_n$.

Best Response and Nash Equilibrium

Our definitions of best response and Nash equilibrium generalize from actions to strategies.

▶ **Best response:**

▶ $s_i^* \in BR(s_{-i})$ iff $\forall s_i \in S_i, u_i(s_i^*, s_{-i}) \geq u_i(s_i, s_{-i})$

▶ **Nash equilibrium:**

▶ $s = \langle s_1, \dots, s_n \rangle$ is a Nash equilibrium iff $\forall i, s_i \in BR(s_{-i})$

▶ **Every finite game has a Nash equilibrium!** [Nash, 1950]

▶ e.g., matching pennies: both players play heads/tails 50%/50%

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Max-Min Strategies

- ▶ Player i 's **maxmin strategy** is a strategy that maximizes i 's worst-case payoff, in the situation where all the other players (whom we denote $-i$) happen to play the strategies which cause the greatest harm to i .
- ▶ The **maxmin value** (or **safety level**) of the game for player i is that minimum amount of payoff guaranteed by a maxmin strategy.
- ▶ Why would i want to play a maxmin strategy?

Max-Min Strategies

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- ▶ The **maxmin value** (or **safety level**) of the game for player i is that minimum amount of payoff guaranteed by a maxmin strategy.
- ▶ Why would i want to play a maxmin strategy?
 - ▶ a conservative agent maximizing worst-case payoff
 - ▶ a paranoid agent who believes everyone is out to get him

Definition

The **maxmin strategy** for player i is $\arg \max_{s_i} \min_{s_{-i}} u_i(s_1, s_2)$, and the **maxmin value** for player i is $\max_{s_i} \min_{s_{-i}} u_i(s_1, s_2)$.

Min-Max Strategies

- ▶ Player i 's **minmax strategy** in a 2-player game is a strategy that minimizes the other player $-i$'s best-case payoff.
- ▶ The **minmax value** of the 2-player game for player i is that maximum amount of payoff that $-i$ could achieve under i 's minmax strategy.
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Min-Max Strategies

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- ▶ Why would i want to play a minmax strategy?
 - ▶ to punish the other agent as much as possible

Definition

The **maxmin strategy** for player i is $\arg \max_{s_i} \min_{s_{-i}} u_i(s_1, s_2)$, and the **maxmin value** for player i is $\max_{s_i} \min_{s_{-i}} u_i(s_1, s_2)$.

Definition

In a two-player game, the **minmax strategy** for player i is $\arg \min_{s_i} \max_{s_{-i}} u_{-i}(s_1, s_2)$, and the **minmax value** for player i is $\min_{s_i} \max_{s_{-i}} u_{-i}(s_1, s_2)$.

Minmax Theorem

Theorem (Minmax theorem (von Neumann, 1928))

In any finite, two-player, zero-sum game it is the case that:

- 1. The maxmin value for one player is equal to the minmax value for the other player. By convention, the maxmin value for player 1 is called the **value of the game**.*
- 2. For both players, the set of maxmin strategies coincides with the set of minmax strategies.*
- 3. Any maxmin strategy profile (or, equivalently, minmax strategy profile) is a Nash equilibrium. Furthermore, these are all the Nash equilibria. Consequently, all Nash equilibria have the same payoff vector (namely, those in which player 1 gets the value of the game).*

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Linear Programming

A **linear program** is defined by:

- ▶ a set of real-valued variables
- ▶ a linear objective function
 - ▶ a weighted sum of the variables
- ▶ a set of linear constraints
 - ▶ the requirement that a weighted sum of the variables must be greater than or equal to some constant

Linear Programming

$$\begin{aligned} & \text{maximize} && \sum_i w_i x_i \\ & \text{subject to} && \sum_i w_i^c x_i \geq b^c && \forall c \in C \\ & && x_i \geq 0 && \forall x_i \in X \end{aligned}$$

- ▶ These problems can be solved in **polynomial time** using interior point methods.
 - ▶ Interestingly, the (worst-case exponential) **simplex method** is often faster in practice.

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Computing equilibria of zero-sum games

$$\begin{array}{ll}
 \text{minimize} & U_1^* \\
 \text{subject to} & \sum_{a_2 \in A_2} u_1(a_1, a_2) \cdot s_2^{a_2} \leq U_1^* \quad \forall a_1 \in A_1 \\
 & \sum_{a_2 \in A_2} s_2^{a_2} = 1 \\
 & s_2^{a_2} \geq 0 \quad \forall a_2 \in A_2
 \end{array}$$

- ▶ variables:
 - ▶ U_1^* is the expected utility for player 1
 - ▶ $s_2^{a_2}$ is player 2's probability of playing action a_2 under his mixed strategy
- ▶ each $u_1(a_1, a_2)$ is a constant.

Computing equilibria of zero-sum games

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 & \sum_{a_2 \in A_2} s_2^{a_2} = 1 \\
 & s_2^{a_2} \geq 0 \quad \forall a_2 \in A_2
 \end{array}$$

- ▶ s_2 is a valid probability distribution.

Computing equilibria of zero-sum games

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 & s_2^{a_2} \geq 0 \quad \forall a_2 \in A_2
 \end{array}$$

- ▶ U_1^* is as small as possible.

Computing equilibria of zero-sum games

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 & s_2^{a_2} \geq 0 \quad \forall a_2 \in A_2
 \end{array}$$

- ▶ Player 1's expected utility for playing each of his actions under player 2's mixed strategy is no more than U_1^* .
 - ▶ Because U_1^* is minimized, this constraint will be tight for some actions: the support of player 1's mixed strategy.

Computing equilibria of zero-sum games

$$\begin{array}{ll}
 \text{minimize} & U_1^* \\
 \text{subject to} & \sum_{a_2 \in A_2} u_1(a_1, a_2) \cdot s_2^{a_2} \leq U_1^* \quad \forall a_1 \in A_1 \\
 & \sum_{a_2 \in A_2} s_2^{a_2} = 1 \\
 & s_2^{a_2} \geq 0 \quad \forall a_2 \in A_2
 \end{array}$$

- ▶ This formulation gives us the minmax strategy for player 2.
- ▶ To get the minmax strategy for player 1, we need to solve a second (analogous) LP.

Computing Maxmin Strategies in General-Sum Games

Let's say we want to compute a maxmin strategy for player 1 in an arbitrary 2-player game G .

Computing Maxmin Strategies in General-Sum Games

Let's say we want to compute a maxmin strategy for player 1 in an arbitrary 2-player game G .

- ▶ Create a new game G' where player 2's payoffs are just the negatives of player 1's payoffs.
- ▶ The maxmin strategy for player 1 in G does not depend on player 2's payoffs
 - ▶ Thus, the maxmin strategy for player 1 in G is the same as the maxmin strategy for player 1 in G'
- ▶ By the minmax theorem, equilibrium strategies for player 1 in G' are equivalent to a maxmin strategies
- ▶ Thus, to find a maxmin strategy for G , find an equilibrium strategy for G' .

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Traveler's Dilemma

Two travelers purchase identical African masks while on a tropical vacation. Their luggage is lost on the return trip, and the airline asks them to make independent claims for compensation. In anticipation of excessive claims, the airline representative announces: "We know that the bags have identical contents, and we will entertain any claim between \$180 and \$300, but you will each be reimbursed at an amount that equals the minimum of the two claims submitted. If the two claims differ, we will also pay a reward R to the person making the smaller claim and we will deduct a penalty R from the reimbursement to the person making the larger claim."

Traveler's Dilemma

- ▶ Action: choose an integer between 180 and 300
- ▶ If both players pick the same number, they both get that amount as payoff
- ▶ If players pick a different number:
 - ▶ the low player gets his number (L) plus some constant R
 - ▶ the high player gets $L - R$.
- ▶ Play this game *once* with a partner; play with as many different partners as you like.
 - ▶ $R = 5$.

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- ▶ Play this game *once* with a partner; play with as many different partners as you like.
 - ▶ $R = 5$.
 - ▶ $R = 180$.

Traveler's Dilemma

- ▶ What is the equilibrium?

Traveler's Dilemma

- ▶ What is the equilibrium?
 - ▶ $(180, 180)$ is the only equilibrium, for all $R \geq 2$.

Traveler's Dilemma

- ▶ What is the equilibrium?
 - ▶ $(180, 180)$ is the only equilibrium, for all $R \geq 2$.
- ▶ What happens?

Traveler's Dilemma

- ▶ What is the equilibrium?
 - ▶ $(180, 180)$ is the only equilibrium, for all $R \geq 2$.
- ▶ What happens?
 - ▶ with $R = 5$ most people choose 295–300
 - ▶ with $R = 180$ most people choose 180

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Domination

- ▶ Let s_i and s'_i be two strategies for player i , and let S_{-i} be the set of all possible strategy profiles for the other players

Definition

s_i **strictly dominates** s'_i if $\forall s_{-i} \in S_{-i}, u_i(s_i, s_{-i}) > u_i(s'_i, s_{-i})$

Definition

s_i **weakly dominates** s'_i if $\forall s_{-i} \in S_{-i}, u_i(s_i, s_{-i}) \geq u_i(s'_i, s_{-i})$ and $\exists s_{-i} \in S_{-i}, u_i(s_i, s_{-i}) > u_i(s'_i, s_{-i})$

Definition

s_i **very weakly dominates** s'_i if $\forall s_{-i} \in S_{-i}, u_i(s_i, s_{-i}) \geq u_i(s'_i, s_{-i})$

Equilibria and dominance

- ▶ If one strategy dominates all others, we say it is **dominant**.
- ▶ A strategy profile consisting of dominant strategies for every player must be a Nash equilibrium.
 - ▶ An equilibrium in strictly dominant strategies must be unique.

Equilibria and dominance

- ▶ If one strategy dominates all others, we say it is **dominant**.
- ▶ A strategy profile consisting of dominant strategies for every player must be a Nash equilibrium.
 - ▶ An equilibrium in strictly dominant strategies must be unique.
- ▶ Consider Prisoner's Dilemma again
 - ▶ not only is the only equilibrium the only non-Pareto-optimal outcome, but it's also an equilibrium in strictly dominant strategies!

Dominated strategies

- ▶ No equilibrium can involve a strictly dominated strategy (why?)

Dominated strategies

- ▶ No equilibrium can involve a strictly dominated strategy (why?)
 - ▶ Thus we can remove it, and end up with a strategically equivalent game
 - ▶ This might allow us to remove another strategy that wasn't dominated before
 - ▶ Running this process to termination is called **iterated removal of dominated strategies**.