

Stochastic Games and Bayesian Games

CPSC 532L Lecture 10

Lecture Overview

- 1 Stochastic Games
- 2 Bayesian Games
- 3 Analyzing Bayesian games

Introduction

- What if we didn't always repeat back to the same stage game?
- A stochastic game is a generalization of **repeated games**
 - agents repeatedly play games from a set of normal-form games
 - the game played at any iteration depends on the previous game played and on the actions taken by all agents in that game
- A stochastic game is a generalized **Markov decision process**
 - there are multiple players
 - one reward function for each agent
 - the state transition function and reward functions depend on the action choices of **both** players

Formal Definition

Definition

A **stochastic game** is a tuple (Q, N, A, P, R) , where

- Q is a finite set of states,
- N is a finite set of n players,
- $A = A_1 \times \cdots \times A_n$, where A_i is a finite set of actions available to player i ,
- $P : Q \times A \times Q \mapsto [0, 1]$ is the transition probability function; $P(q, a, \hat{q})$ is the probability of transitioning from state q to state \hat{q} after joint action a , and
- $R = r_1, \dots, r_n$, where $r_i : Q \times A \mapsto \mathbb{R}$ is a real-valued payoff function for player i .

Remarks

- This assumes strategy space is the same in all games
 - otherwise just more notation
- Again we can have average or discounted payoffs.
- Interesting special cases:
 - zero-sum stochastic game
 - single-controller stochastic game
 - transitions (but not payoffs) depend on only one agent

Strategies

- What is a pure strategy?

Strategies

- What is a pure strategy?
 - pick an action conditional on every possible history
 - of course, mixtures over these pure strategies are possible too!
- Some interesting restricted classes of strategies:
 - **behavioral strategy**: $s_i(h_t, a_{i_j})$ returns the probability of playing action a_{i_j} for history h_t .
 - the substantive assumption here is that mixing takes place at each history independently, not once at the beginning of the game
 - **Markov strategy**: s_i is a behavioral strategy in which $s_i(h_t, a_{i_j}) = s_i(h'_t, a_{i_j})$ if $q_t = q'_t$, where q_t and q'_t are the final states of h_t and h'_t , respectively.
 - for a given time t , the distribution over actions only depends on the current state
 - **stationary strategy**: s_i is a Markov strategy in which $s_i(h_{t_1}, a_{i_j}) = s_i(h'_{t_2}, a_{i_j})$ if $q_{t_1} = q'_{t_2}$, where q_{t_1} and q'_{t_2} are the final states of h_{t_1} and h'_{t_2} , respectively.
 - no dependence even on t

Equilibrium (discounted rewards)

- **Markov perfect equilibrium:**
 - a strategy profile consisting of only Markov strategies that is a Nash equilibrium regardless of the starting state
 - analogous to subgame-perfect equilibrium

Theorem

Every n -player, general sum, discounted reward stochastic game has a Markov perfect equilibrium.

Equilibrium (average rewards)

- **Irreducible stochastic game:**
 - every strategy profile gives rise to an irreducible Markov chain over the set of games
 - irreducible Markov chain: possible to get from every state to every other state
 - during the (infinite) execution of the stochastic game, each stage game is guaranteed to be played infinitely often—for any strategy profile
 - without this condition, limit of the mean payoffs may not be defined

Theorem

For every 2-player, general sum, average reward, irreducible stochastic game has a Nash equilibrium.

A folk theorem

Theorem

For every 2-player, general sum, irreducible stochastic game, and every feasible outcome with a payoff vector r that provides to each player at least his minmax value, there exists a Nash equilibrium with a payoff vector r . This is true for games with average rewards, as well as games with large enough discount factors (i.e. with players that are sufficiently patient).

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Fun Game

- Choose a phone number none of your neighbours knows; consider it to be ABC-DEFG

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- Questions:
 - what is the role of uncertainty here?
 - can we model this uncertainty using an imperfect information extensive form game?

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- Questions:
 - what is the role of uncertainty here?
 - can we model this uncertainty using an imperfect information extensive form game?
 - imperfect info means not knowing what node you're in in the info set
 - here we're not sure what game is being played (though if we allow a move by nature, we can do it)

Introduction

- So far, we've assumed that all players know what game is being played. Everyone knows:
 - the number of players
 - the actions available to each player
 - the payoff associated with each action vector
- Why is this true in imperfect information games?
- We'll assume:
 - 1 All possible games have the same number of agents and the same strategy space for each agent; they differ only in their payoffs.
 - 2 The beliefs of the different agents are posteriors, obtained by conditioning a common prior on individual private signals.

Definition 1: Information Sets

- **Bayesian game**: a set of games that differ only in their payoffs, a common prior defined over them, and a partition structure over the games for each agent.

Definition (Bayesian Game: Information Sets)

A **Bayesian game** is a tuple (N, G, P, I) where

- N is a set of agents,
- G is a set of games with N agents each such that if $g, g' \in G$ then for each agent $i \in N$ the strategy space in g is identical to the strategy space in g' ,
- $P \in \Pi(G)$ is a common prior over games, where $\Pi(G)$ is the set of all probability distributions over G , and
- $I = (I_1, \dots, I_N)$ is a set of partitions of G , one for each agent.

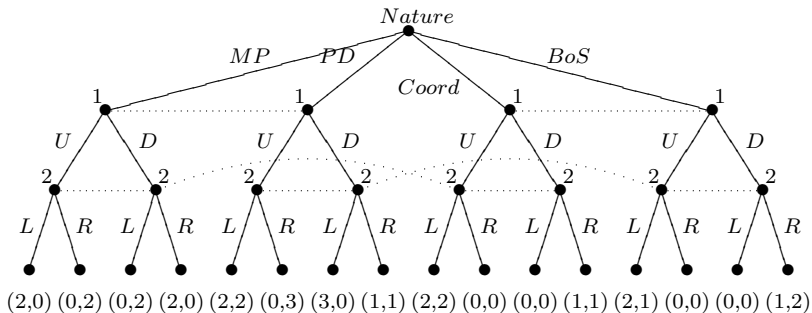
Definition 1: Example

	$I_{2,1}$	$I_{2,2}$																
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Definition 2: Extensive Form with Chance Moves

- Add an agent, “Nature,” who follows a commonly known mixed strategy.
- Thus, reduce Bayesian games to extensive form games of imperfect information.
- This definition is cumbersome for the same reason that IIEF is a cumbersome way of representing matrix games like Prisoner’s dilemma
 - however, it makes sense when the agents really do move sequentially, and at least occasionally observe each other’s actions.

Definition 2: Example



Definition 3: Epistemic Types

- Directly represent uncertainty over utility function using the notion of **epistemic type**.

Definition

A **Bayesian game** is a tuple (N, A, Θ, p, u) where

- N is a set of agents,
- $A = (A_1, \dots, A_n)$, where A_i is the set of actions available to player i ,
- $\Theta = (\Theta_1, \dots, \Theta_n)$, where Θ_i is the type space of player i ,
- $p : \Theta \rightarrow [0, 1]$ is the common prior over types,
- $u = (u_1, \dots, u_n)$, where $u_i : A \times \Theta \rightarrow \mathbb{R}$ is the utility function for player i .

Definition 3: Example

	$I_{2,1}$	$I_{2,2}$								
$I_{1,1}$	<p>MP</p> <table border="1"> <tr><td>2,0</td><td>0,2</td></tr> <tr><td>0,2</td><td>2,0</td></tr> </table> <p>$p = 0.3$</p>	2,0	0,2	0,2	2,0	<p>PD</p> <table border="1"> <tr><td>2,2</td><td>0,3</td></tr> <tr><td>3,0</td><td>1,1</td></tr> </table> <p>$p = 0.1$</p>	2,2	0,3	3,0	1,1
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2,2	0,0									
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2,1	0,0									
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a_1	a_2	θ_1	θ_2	u_1	u_2
U	L	$\theta_{1,1}$	$\theta_{2,1}$	2	0
U	L	$\theta_{1,1}$	$\theta_{2,2}$	2	2
U	L	$\theta_{1,2}$	$\theta_{2,1}$	2	2
U	L	$\theta_{1,2}$	$\theta_{2,2}$	2	1
U	R	$\theta_{1,1}$	$\theta_{2,1}$	0	2
U	R	$\theta_{1,1}$	$\theta_{2,2}$	0	3
U	R	$\theta_{1,2}$	$\theta_{2,1}$	0	0
U	R	$\theta_{1,2}$	$\theta_{2,2}$	0	0

a_1	a_2	θ_1	θ_2	u_1	u_2
D	L	$\theta_{1,1}$	$\theta_{2,1}$	0	2
D	L	$\theta_{1,1}$	$\theta_{2,2}$	3	0
D	L	$\theta_{1,2}$	$\theta_{2,1}$	0	0
D	L	$\theta_{1,2}$	$\theta_{2,2}$	0	0
D	R	$\theta_{1,1}$	$\theta_{2,1}$	2	0
D	R	$\theta_{1,1}$	$\theta_{2,2}$	1	1
D	R	$\theta_{1,2}$	$\theta_{2,1}$	1	1
D	R	$\theta_{1,2}$	$\theta_{2,2}$	1	2

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Strategies

- **Pure strategy:** $s_i : \Theta_i \rightarrow A_i$
 - a mapping from every type agent i could have to the action he would play if he had that type.
- **Mixed strategy:** $s_i : \Theta_i \rightarrow \Pi(A_i)$
 - a mapping from i 's type to a probability distribution over his action choices.
- $s_j(a_j|\theta_j)$
 - the probability under mixed strategy s_j that agent j plays action a_j , given that j 's type is θ_j .

Expected Utility

Three meaningful notions of expected utility:

- *ex-ante*
 - the agent knows nothing about anyone's actual type;
- *ex-interim*
 - an agent knows his own type but not the types of the other agents;
- *ex-post*
 - the agent knows all agents' types.

Ex-interim expected utility

Definition (*Ex-interim* expected utility)

Agent i 's *ex-interim expected utility* in a Bayesian game (N, A, Θ, p, u) , where i 's type is θ_i and where the agents' strategies are given by the mixed strategy profile s , is defined as

$$EU_i(s|\theta_i) = \sum_{\theta_{-i} \in \Theta_{-i}} p(\theta_{-i}|\theta_i) \sum_{a \in A} \left(\prod_{j \in N} s_j(a_j|\theta_j) \right) u_i(a, \theta_{-i}, \theta_i).$$

- i must consider every θ_{-i} and every a in order to evaluate $u_i(a, \theta_i, \theta_{-i})$.
- i must weight this utility value by:
 - the probability that a would be realized given all players' mixed strategies and types;
 - the probability that the other players' types would be θ_{-i} given that his own type is θ_i .

Ex-ante expected utility

Definition (*Ex-ante* expected utility)

Agent i 's **ex-ante expected utility** in a Bayesian game (N, A, Θ, p, u) , where the agents' strategies are given by the mixed strategy profile s , is defined as

$$EU_i(s) = \sum_{\theta_i \in \Theta_i} p(\theta_i) EU_i(s|\theta_i)$$

or equivalently as

$$EU_i(s) = \sum_{\theta \in \Theta} p(\theta) \sum_{a \in A} \left(\prod_{j \in N} s_j(a_j|\theta_j) \right) u_i(a, \theta).$$

Ex-post expected utility

Definition (*Ex-post* expected utility)

Agent i 's **ex-post expected utility** in a Bayesian game (N, A, Θ, p, u) , where the agents' strategies are given by s and the agent' types are given by θ , is defined as

$$EU_i(s, \theta) = \sum_{a \in A} \left(\prod_{j \in N} s_j(a_j | \theta_j) \right) u_i(a, \theta).$$

- The only uncertainty here concerns the other agents' mixed strategies, since i knows everyone's type.

Best response

Definition (Best response in a Bayesian game)

The set of agent i 's **best responses** to mixed strategy profile s_{-i} are given by

$$BR_i(s_{-i}) = \arg \max_{s'_i \in S_i} EU_i(s'_i, s_{-i}).$$

- it may seem odd that BR is calculated based on i 's *ex-ante* expected utility.
- However, write $EU_i(s)$ as $\sum_{\theta_i \in \Theta_i} p(\theta_i) EU_i(s|\theta_i)$ and observe that $EU_i(s'_i, s_{-i}|\theta_i)$ does not depend on strategies that i would play if his type were not θ_i .
- Thus, we are in fact performing independent maximization of i 's *ex-interim* expected utility conditioned on each type that he could have.

Nash equilibrium

Definition (Bayes-Nash equilibrium)

A **Bayes-Nash equilibrium** is a mixed strategy profile s that satisfies $\forall i \ s_i \in BR_i(s_{-i})$.

- we can also construct an induced normal form for Bayesian games
- the numbers in the cells will correspond to *ex-ante* expected utilities
 - however as argued above, as long as the strategy space is unchanged, best responses don't change between the *ex-ante* and *ex-interim* cases.

ex-post Equilibrium

Definition (*ex-post* equilibrium)

A ***ex-post* equilibrium** is a mixed strategy profile s that satisfies $\forall \theta, \forall i, s_i \in \arg \max_{s'_i \in S_i} EU_i(s'_i, s_{-i}, \theta)$.

- somewhat similar to **dominant strategy**, but not quite
 - EP: agents do not need to have accurate beliefs about the type distribution
 - DS: agents do not need to have accurate beliefs about others' strategies