

# Stochastic Games and Bayesian Games

## CPSC 532I Lecture 10

# Lecture Overview

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

# Finitely Repeated Games

- Everything is straightforward if we repeat a game a finite number of times
- we can write the whole thing as an extensive-form game with imperfect information
  - at each round players don't know what the others have done; afterwards they do
  - overall payoff function is additive: sum of payoffs in stage games

# Infinitely Repeated Games

- Consider an infinitely repeated game in extensive form:
  - an infinite tree!
- Thus, payoffs cannot be attached to terminal nodes, nor can they be defined as the sum of the payoffs in the stage games (which in general will be infinite).

## Definition

Given an infinite sequence of payoffs  $r_1, r_2, \dots$  for player  $i$ , the **average reward** of  $i$  is

$$\lim_{k \rightarrow \infty} \sum_{j=1}^k \frac{r_j}{k}.$$

# Nash Equilibria

- With an infinite number of equilibria, what can we say about Nash equilibria?
  - we **won't** be able to construct an induced normal form and then appeal to Nash's theorem to say that an equilibrium exists
  - Nash's theorem only applies to finite games
- Furthermore, with an infinite number of strategies, there could be an infinite number of pure-strategy equilibria!
- It turns out we can characterize a set of **payoffs** that are achievable under equilibrium, without having to enumerate the equilibria.

# Definitions

- Consider any  $n$ -player game  $G = (N, A, u)$  and any payoff vector  $r = (r_1, r_2, \dots, r_n)$ .
- Let  $v_i = \min_{s_{-i} \in S_{-i}} \max_{s_i \in S_i} u_i(s_{-i}, s_i)$ .
  - $i$ 's **minmax value**: the amount of utility  $i$  can get when  $-i$  play a minmax strategy against him

## Definition

A payoff profile  $r$  is **enforceable** if  $r_i \geq v_i$ .

## Definition

A payoff profile  $r$  is **feasible** if there exist rational, non-negative values  $\alpha_a$  such that for all  $i$ , we can express  $r_i$  as  $\sum_{a \in A} \alpha_a u_i(a)$ , with  $\sum_{a \in A} \alpha_a = 1$ .

- a payoff profile is feasible if it is a convex, rational combination of the outcomes in  $G$ .

# Folk Theorem

## Theorem (Folk Theorem)

Consider any  $n$ -player game  $G$  and any payoff vector  $(r_1, r_2, \dots, r_n)$ .

- 1 If  $r$  is the payoff in any Nash equilibrium of the infinitely repeated  $G$  with average rewards, then for each player  $i$ ,  $r_i$  is enforceable.
- 2 If  $r$  is both feasible and enforceable, then  $r$  is the payoff in some Nash equilibrium of the infinitely repeated  $G$  with average rewards.

# Folk Theorem (Part 1)

Payoff in Nash  $\rightarrow$  enforceable

**Part 1:** Suppose  $r$  is not enforceable, i.e.  $r_i < v_i$  for some  $i$ . Then consider a deviation of this player  $i$  to  $b_i(s_{-i}(h))$  for any history  $h$  of the repeated game, where  $b_i$  is any best-response action in the stage game and  $s_{-i}(h)$  is the equilibrium strategy of other players given the current history  $h$ . By definition of a minmax strategy, player  $i$  will receive a payoff of at least  $v_i$  in every stage game if he adopts this strategy, and so  $i$ 's average reward is also at least  $v_i$ . Thus  $i$  cannot receive the payoff  $r_i < v_i$  in any Nash equilibrium.



# Folk Theorem (Part 2)

## Feasible and enforceable $\rightarrow$ Nash

**Part 2:** Since  $r$  is a feasible payoff profile, we can write it as  $r_i = \sum_{a \in A} \left( \frac{\beta_a}{\gamma} \right) u_i(a)$ , where  $\beta_a$  and  $\gamma$  are non-negative integers.<sup>1</sup> Since the combination was convex, we have  $\gamma = \sum_{a \in A} \beta_a$ . We're going to construct a strategy profile that will cycle through all outcomes  $a \in A$  of  $G$  with cycles of length  $\gamma$ , each cycle repeating action  $a$  exactly  $\beta_a$  times. Let  $(a^t)$  be such a sequence of outcomes. Let's define a strategy  $s_i$  of player  $i$  to be a trigger version of playing  $(a^t)$ : if nobody deviates, then  $s_i$  plays  $a_i^t$  in period  $t$ . However, if there was a period  $t'$  in which some player  $j \neq i$  deviated, then  $s_i$  will play  $(p_{-j})_i$ , where  $(p_{-j})$  is a solution to the minimization problem in the definition of  $v_j$ .

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<sup>1</sup>Recall that  $\alpha_a$  were required to be rational. So we can take  $\gamma$  to be their common denominator.

# Folk Theorem (Part 2)

Feasible and enforceable  $\rightarrow$  Nash

First observe that if everybody plays according to  $s_i$ , then, by construction, player  $i$  receives average payoff of  $r_i$  (look at averages over periods of length  $\gamma$ ). Second, this strategy profile is a Nash equilibrium. Suppose everybody plays according to  $s_i$ , and player  $j$  deviates at some point. Then, forever after, player  $j$  will receive his min max payoff  $v_j \leq r_j$ , rendering the deviation unprofitable.

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# 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  $s$  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

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

# Fun 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

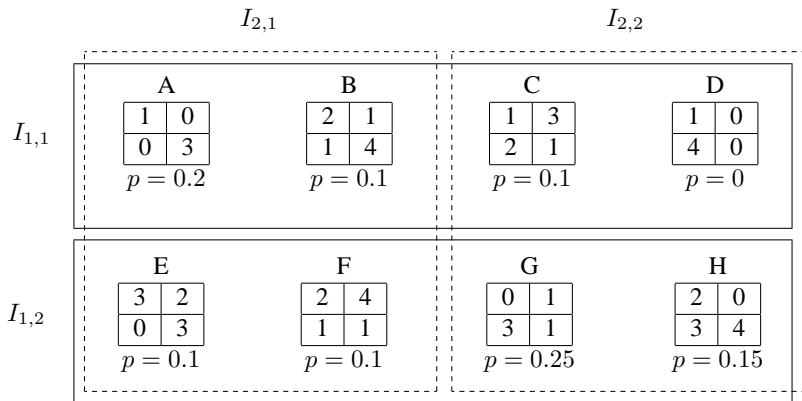
- **Bayesian game**: a set of games that differ only in their payoffs, a common prior defined over them, and a partition structure over the games for each agent.

### 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.

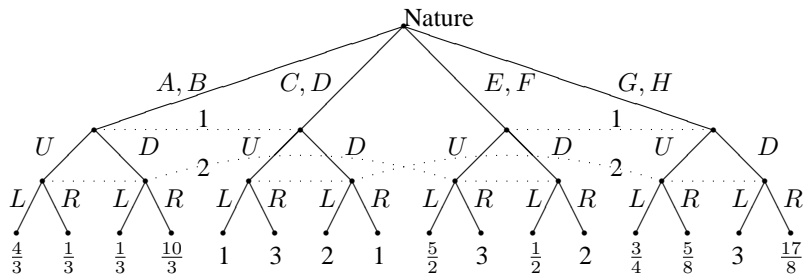
# Definition 1: Example



## 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**.

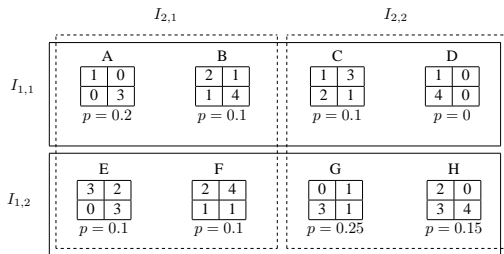
### Definition

A **Bayesian game** is a tuple  $(N, A, \Theta, p, u)$  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  $\Theta_i$  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



$a_1$	$a_2$	$\theta_1$	$\theta_2$	$u_1$
U	L	$\theta_{1,1}$	$\theta_{2,1}$	4/3
U	L	$\theta_{1,1}$	$\theta_{2,2}$	1
U	L	$\theta_{1,2}$	$\theta_{2,1}$	5/2
U	L	$\theta_{1,2}$	$\theta_{2,2}$	3/4
U	R	$\theta_{1,1}$	$\theta_{2,1}$	1/3
U	R	$\theta_{1,1}$	$\theta_{2,2}$	3
U	R	$\theta_{1,2}$	$\theta_{2,1}$	3
U	R	$\theta_{1,2}$	$\theta_{2,2}$	5/8

$a_1$	$a_2$	$\theta_1$	$\theta_2$	$u_1$
D	L	$\theta_{1,1}$	$\theta_{2,1}$	1/3
D	L	$\theta_{1,1}$	$\theta_{2,2}$	2
D	L	$\theta_{1,2}$	$\theta_{2,1}$	1/2
D	L	$\theta_{1,2}$	$\theta_{2,2}$	3
D	R	$\theta_{1,1}$	$\theta_{2,1}$	10/3
D	R	$\theta_{1,1}$	$\theta_{2,2}$	1
D	R	$\theta_{1,2}$	$\theta_{2,1}$	2
D	R	$\theta_{1,2}$	$\theta_{2,2}$	17/8

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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  $\theta_j$ .

# 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  $(N, A, \Theta, p, u)$ , where  $i$ 's type is  $\theta_i$  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  $\theta_{-i}$  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  $\theta_{-i}$  given that his own type is  $\theta_i$ .

# Ex-ante expected utility

## Definition (*Ex-ante* expected utility)

Agent  $i$ 's **ex-ante expected utility** in a Bayesian game  $(N, A, \Theta, p, u)$ , 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  $(N, A, \Theta, p, u)$ , where the agents' strategies are given by  $s$  and the agent' types are given by  $\theta$ , 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  $\theta_i$ .
- 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* Bayes-Nash 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