

Game Theory - Week 7

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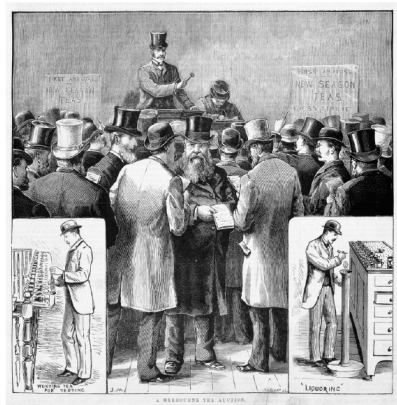
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Overview

- Bayesian Games: Taste
- Bayesian Games: First Definition
- Bayesian Games: Second Definition
- Analyzing Bayesian Games
- Analyzing Bayesian Games: Another Example

Auctions



Tea Auction, Melbourne, Australia, 1885.

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Auctions



A silent auction. Looks suspiciously like a game.

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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?

Introduction

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- Why is this true in imperfect information games?

Now we'll relax this. We'll still assume:

1. All possible games have the same number of agents and the same strategy space for each agent; differing only in payoffs.
2. Agents' beliefs are posteriors, obtained by conditioning a common prior on individual private signals.

Definition I : 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 I: Example

		$I_{2,1}$	$I_{2,2}$																			
		<table style="border-collapse: collapse; margin: auto;"> <tr> <td></td> <td colspan="2">MP</td> <td colspan="2">PD</td> </tr> <tr> <td rowspan="2" style="border: none; padding-right: 10px;">$I_{1,1}$</td> <td style="border: 1px solid black; padding: 5px;">2, 0</td> <td style="border: 1px solid black; padding: 5px;">0, 2</td> <td style="border: 1px solid black; padding: 5px;">2, 2</td> <td style="border: 1px solid black; padding: 5px;">0, 3</td> </tr> <tr> <td style="border: 1px solid black; padding: 5px;">0, 2</td> <td style="border: 1px solid black; padding: 5px;">2, 0</td> <td style="border: 1px solid black; padding: 5px;">3, 0</td> <td style="border: 1px solid black; padding: 5px;">1, 1</td> </tr> <tr> <td style="border: none;"></td> <td colspan="2" style="border: none; padding-top: 5px;">$p = 0.3$</td> <td colspan="2" style="border: none; padding-top: 5px;">$p = 0.1$</td> </tr> </table>		MP		PD		$I_{1,1}$	2, 0	0, 2	2, 2	0, 3	0, 2	2, 0	3, 0	1, 1		$p = 0.3$		$p = 0.1$		
	MP		PD																			
$I_{1,1}$	2, 0	0, 2	2, 2	0, 3																		
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		<table style="border-collapse: collapse; margin: auto;"> <tr> <td></td> <td colspan="2">Coord</td> <td colspan="2">BoS</td> </tr> <tr> <td rowspan="2" style="border: none; padding-right: 10px;">$I_{1,2}$</td> <td style="border: 1px solid black; padding: 5px;">2, 2</td> <td style="border: 1px solid black; padding: 5px;">0, 0</td> <td style="border: 1px solid black; padding: 5px;">2, 1</td> <td style="border: 1px solid black; padding: 5px;">0, 0</td> </tr> <tr> <td style="border: 1px solid black; padding: 5px;">0, 0</td> <td style="border: 1px solid black; padding: 5px;">1, 1</td> <td style="border: 1px solid black; padding: 5px;">0, 0</td> <td style="border: 1px solid black; padding: 5px;">1, 2</td> </tr> <tr> <td style="border: none;"></td> <td colspan="2" style="border: none; padding-top: 5px;">$p = 0.2$</td> <td colspan="2" style="border: none; padding-top: 5px;">$p = 0.4$</td> </tr> </table>		Coord		BoS		$I_{1,2}$	2, 2	0, 0	2, 1	0, 0	0, 0	1, 1	0, 0	1, 2		$p = 0.2$		$p = 0.4$		
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Definition 2: 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 2: Example

		$I_{2,1}$		$I_{2,2}$	
		MP		PD	
$I_{1,1}$		2, 0	0, 2	2, 2	0, 3
		0, 2	2, 0	3, 0	1, 1
		$p = 0.3$		$p = 0.1$	
		Coord		BoS	
$I_{1,2}$		2, 2	0, 0	2, 1	0, 0
		0, 0	1, 1	0, 0	1, 2
		$p = 0.2$		$p = 0.4$	

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

Bayesian (Nash) Equilibrium

- A plan of action for each player as a function of types that maximize each type's expected utility:

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- $s_i : (a_i \mid \theta_i)$
 - the probability under mixed strategy s_i that agent i plays action a_i , given that i 's type is θ_i .

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

Interim expected utility

- Given a Bayesian game (N, A, Θ, p, u) with finite sets of players, actions, and types, i 's **interim expected utility** with respect to type θ_i and a mixed strategy profile s is

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

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- i 's **ex ante expected utility** with respect to a mixed strategy profile s is

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

Bayesian Equilibrium or Bayes-Nash equilibrium

A **Bayesian equilibrium** is a mixed strategy profile s that satisfies

$$s_i \in \arg \max_{s'_i} EU_i(s'_i, s_{-i} | \theta_i)$$

for each i and $\theta_i \in \Theta_i$.

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The above is defined based on interim maximization. It is equivalent to an ex ante formulation:

If $p(\theta_i) > 0$ for all $\theta_i \in \Theta_i$, then this is equivalent to requiring that

$$s_i \in \arg \max_{s'_i} EU_i(s'_i, s_{-i}) = \arg \max_{s'_i} \sum_{\theta_i} p(\theta_i) EU_i(s'_i, s_{-i} | \theta_i)$$

for each i .

Bayesian (Nash) Equilibrium

- Explicitly models behavior in an uncertain environment
- Players choose strategies to maximize their payoffs in response to others accounting for:
 - strategic uncertainty about how others will play and
 - payoff uncertainty about the value to their actions.

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A Sheriff's Dilemma

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- the suspect is either a criminal with probability p or not with probability $1 - p$.
- the sheriff would rather shoot if the suspect shoots, but not if the suspect does not.
- the criminal would rather shoot even if the sheriff does not, as the criminal would be caught if does not shoot.
- the innocent suspect would rather not shoot even if the sheriff shoots.

A Sheriff's Dilemma

Sheriff

<i>Good</i>	<i>Shoot</i>	<i>Not</i>
<i>Shoot</i>	-3, -1	-1, -2
<i>Not</i>	-2, -1	0, 0

<i>Bad</i>	<i>Shoot</i>	<i>Not</i>
<i>Shoot</i>	0, 0	2, -2
<i>Not</i>	-2, -1	-1, 1

Summary: Bayesian (Nash) Equilibrium

- Explicitly models behavior in an uncertain environment
- Players choose strategies to maximize their payoffs in response to others accounting for:
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