

Econ 421  
Fall, 2023  
Li, Hao  
UBC

## LECTURE 7. STATIC GAMES OF INCOMPLETE INFORMATION

### 1. Bayesian game

Static games of incomplete information, where players

- simultaneously choose actions and the game ends;
- have preferences over actions chosen by all of the players;
- have incomplete, and often asymmetric, information about game.

- **Example** (Cournot duopoly with private cost). Consider two firms, 1 and 2, choosing quantities  $q_1$  and  $q_2$  simultaneously. The inverse demand function is given by  $p = a - (q_1 + q_2)$ . Firm 1's marginal cost is known to be  $c$ , with no fixed cost, while Firm 2 has private knowledge about its marginal cost. In particular, Firm 1 only knows that Firm 2's cost is either  $c_H$  or  $c_L < c_H$ , with probabilities  $p_H$  and  $p_L$  respectively ( $p_H + p_L = 1$ ).

- **Example** (First price auction with independent private values).  
Consider a sealed-bid first price auction. Two potential buyers bid for an indivisible good. Each has his own valuation  $v_i$ ; Each has private knowledge about his own valuation for the good, but knows only that the other bidder's valuation is either  $v_H$  or  $v_L < v_H$ , with probabilities  $p_H$  and  $p_L$  respectively ( $p_H + p_L = 1$ ). The bidder with the higher bid wins the object and pays his bid; if two bids are same, each gets object with probability  $\frac{1}{2}$  and pays the bid. The payoff for the winning bidder is his valuation less his bid, and the losing bidder pays nothing and has payoff 0.

- Model of static games of incomplete information: *Bayesian* game
- As in static games of complete information, we need to specify
  - A set of players.
  - A set of available actions for each player.
  - A payoff function for each player.
- New element is how to model incomplete information of players have about relevant aspects of the game.

Formally, a Bayesian game consists of

- A set of players,  $i = 1, \dots, n$ .
- Each player  $i$  can be any *type* from a *type space*  $T_i$ .
  - *Prior probability* of each *type profile*  $t = (t_1, \dots, t_n)$  is  $p(t)$ .
  - Each  $i$  knows his type own  $t_i$  and *conditional probability distribution*  $p(t_{-i}|t_i)$  of opponents' type profile.
- Each player  $i$  chooses  $a_i$  from action space  $A_i$ .
- Player  $i$ 's vNM payoff from action profile  $a = (a_1, \dots, a_n)$  is  $u_i(a; t)$ .

## Remarks on Bayesian game

- Incomplete information is modeled by uncertainty about types of opponents.
  - Type is payoff relevant:  $u_i(a; t)$  depends on type profile as well as on action profile.
  - Type is general concept used to model incomplete information about: which game you are playing; whether your opponents are “rational” or not.

- Conditional probability distribution  $p(t_{-i}|t_i)$  is derived from  $p(t)$  according to Bayes rule

$$p(t_{-i}|t_i) = \frac{p(t_i, t_{-i})}{\sum_{\tilde{t}_{-i} \in T_{-i}} p(t_i, \tilde{t}_{-i})}.$$

- A special case is independent types: no Bayesian updating, with  $p(t_{-i}|t_i)$  does not depend on  $t_i$ , because  $p(t) = \prod_i p_i(t_i)$  for some given profile of probability distributions  $(p_1, \dots, p_n)$ .

- General model of Bayesian games: state space approach.
  - In most economic applications, payoff type is sufficient to capture all relevant incomplete information.
  - This is true both in Cournot duopoly with private cost, and in First price auction with independent private values.
  - When payoff type is insufficient, a more general approach based on state space is needed.



A general model of Bayesian games.

- A set of players,  $i = 1, \dots, n$ .
- A state space  $\Omega$ , with a prior probability  $\pi(\omega)$  of each state  $\omega \in \Omega$ .
  - A type space  $T_i$  for each player  $i$ , and a signal function  $\tau_i(\cdot)$  that maps each state  $\omega \in \Omega$  to a type  $\tau_i(\omega) \in T_i$ .
- Each player  $i$  chooses  $a_i$  from action space  $A_i$ .
- Player  $i$ 's vNM payoff from action profile  $a = (a_1, \dots, a_n)$  is  $u_i(a; \omega)$ .

Remarks on the state-space model.

- State space  $\Omega$  is a complete description of all states of the world that are relevant to the game, not just in terms of payoffs.
- Incomplete information of each player is modeled by the signal function, which partitions the state space into type space.
- Type-space model is a special case of state-space model: take  $\Omega$  to be set of all type profiles, with  $\pi(\omega) = p(t)$ , and for each  $i$ , take the signal function  $\tau_i(\cdot)$  be such that  $\tau_i(\omega) = \tau_i(t_1, \dots, t_n) = t_i$ .

An example of state-space model of Bayesian games.

- Two players, R and C, and one of two games  $\Gamma_1$  and  $\Gamma_2$  is to be played, where each player has the same two possible actions,  $r_1$  and  $r_2$  for R and  $c_1$  and  $c_2$  for C, with different payoffs.
- Consider following different scenarios: neither R nor C knows which game is being played; only R knows which game is being played; R knows which game is being played but does not know if C knows which game is being played; R knows which game is being played and C does not know if R knows that C knows which game is being played.

## 2. Bayesian Nash equilibrium

- Each player  $i$  knows his type  $t_i$ , has belief  $p(t_{-i}|t_i)$  about the opponents' types, and must choose an action from  $A_i$ .
- A *pure strategy* of player  $i$  is a mapping  $s_i : T_i \rightarrow A_i$ ; denote the set of all strategies as  $S_i$ .
  - A strategy  $s_i$  of player  $i$  is a complete plan: it specifies an action  $s_i(t_i)$  for each of his possible type  $t_i$ .
- Why is it necessary to define a strategy as a complete plan, even though each player playing the Bayesian game knows his own type?

- *Bayesian Nash equilibrium* is simply NE of Bayesian game.
- Formally, a Bayesian Nash equilibrium is a strategy profile  $(s_1^*, \dots, s_n^*)$ , such that for each  $i$ , the strategy  $s_i^*$  solves

$$\max_{s_i \in S_i} \sum_t u_i(s_i(t_i), s_{-i}^*(t_{-i}); t) p(t),$$

where  $s_{-i}^*(t_{-i})$  stands for  $(s_1^*(t_1), \dots, s_{i-1}^*(t_{i-1}), s_{i+1}^*(t_{i+1}), \dots, s_n^*(t_n))$ .

- Bayesian Nash equilibrium is a profile of plans, made before each  $i$  is informed of his type  $t_i$ , such that no player has a profitable unilateral deviation from his plan.

**Proposition** (Alternative definition of BNE). A strategy profile  $(s_1^*, \dots, s_n^*)$  is a Bayesian Nash equilibrium if and only if for each  $i$  and each  $t_i \in T_i$ , the action  $s_i^*(t_i)$  solves  $\max_{a_i \in A_i} \sum_{t_{-i}} u_i(a_i, s_{-i}^*(t_{-i}); t_i, t_{-i}) p(t_{-i} | t_i)$ .

- *Proof.* Writing  $p(t_i) = \sum_{t_{-i}} p(t_i, t_{-i})$ , for each  $i$  and each  $s_i \in S_i$ ,

$$\sum_t u_i(s_i(t_i), s_{-i}^*(t_{-i}); t_i, t_{-i}) p(t) = \sum_{t_i} p(t_i) \left[ \sum_{t_{-i}} u_i(s_i(t_i), s_{-i}^*(t_{-i}); t) p(t_{-i} | t_i) \right].$$

- Interpretation: each type of a player is a separate player.
- The alternative definition is simpler to apply in practice.

### 3. Cournot duopoly game with incomplete information

- The Bayesian game is given by:
  - Players: Firm 1 and Firm 2.
  - Firm 1's type space is a singleton set; and Firm 2's type space is  $\{c_H, c_L\}$  with probabilities  $p_H$  and  $p_L$ .
  - Action space for both firms is  $A_1 = A_2 = [0, \infty)$ .
  - Payoff function for Firm 1 is  $\pi_1(q_1, q_2) = (a - c - q_1 - q_2)q_1$ , and for Firm 2 is  $\pi_2(q_1, q_2; c_2) = (a - c_2 - q_1 - q_2)q_2$  for each type  $c_2 = c_H, c_L$ .

- Firm 1's strategy is just a quantity  $q_1$ ; Firm 2's strategy is a plan  $(q_{2H}, q_{2L})$ .
- Best responses.
  - Type  $c_H$  Firm 2's best response to  $q_1$  is

$$b_H(q_1) = \begin{cases} \frac{1}{2}(a - c_H - q_1) & \text{if } q_1 < a - c_H, \\ 0 & \text{if } q_1 \geq a - c_H. \end{cases}$$

- Type  $c_L$  Firm 2's best response function  $b_L(q_1)$  is similar, with  $c_L$  replacing  $c_H$  in above.



– Firm 1's best response to Firm 2's strategy  $(q_{2H}, q_{2L})$  solves

$$\max_{q_1} p_H(a - c - q_{2H} - q_1)q_1 + p_L(a - c - q_{2L} - q_1)q_1.$$

– Solution is

$$b_1(q_{2H}, q_{2L}) = \begin{cases} \frac{1}{2}(a - c - p_H q_{2H} - p_L q_{2L}) & \text{if positive,} \\ 0 & \text{otherwise.} \end{cases}$$

- BNE is the intersection of best response functions:

$$\left\{ \begin{array}{l} q_{2H}^* = \frac{1}{2}(a - c_H - q_1^*), \\ q_{2L}^* = \frac{1}{2}(a - c_L - q_1^*), \\ q_1^* = \frac{1}{2}(a - c - p_H q_{2H}^* - p_L q_{2L}^*). \end{array} \right.$$

- Bayesian Nash equilibrium if solutions are positive:

$$\left\{ \begin{array}{l} q_{2H}^* = \frac{1}{3}(a - 2c_H + c) + \frac{1}{6}(c_H - c_L)p_L, \\ q_{2L}^* = \frac{1}{3}(a - 2c_L + c) - \frac{1}{6}(c_H - c_L)p_H, \\ q_1^* = \frac{1}{3}(a - 2c + p_H c_H + p_L c_L). \end{array} \right.$$

- Comparison with Nash equilibrium when  $c_2 = c_H$  and known:

$$q_1 = \frac{1}{3}(a - 2c + c_H), q_2 = \frac{1}{3}(a - 2c_H + c).$$

- Firm 1 produces less under incomplete information, because it is unsure about  $c_2$  and  $c_2$  could be low; knowing this, Firm 2 produces more than it would than when  $c_2 = c_H$  is known.
- Incomplete information hurts Firm 1 and benefits Firm 2.

- Opposite conclusions hold in comparison with NE when  $c_2 = c_L$

and known:  $q_1 = \frac{1}{3}(a - 2c + c_L), q_2 = \frac{1}{3}(a - 2c_L + c).$

#### 4. Bayesian Nash equilibrium in state-space model

- A strategy  $s_i$  of  $i$  is still a complete plan: it specifies an action  $s_i(t_i) \in A_i$  for each type  $t_i \in T_i$ .
- Expected payoff for  $i$  from strategy  $s_i$  against opponents strategy profile  $s_{-i}$  is

$$\sum_{\omega \in \Omega} u_i(s_i(\tau_i(\omega)), s_{-i}(\tau_{-i}(\omega)); \omega) \pi(\omega),$$

where  $s_{-i}(\tau_{-i}(\omega))$  is action profile where each player  $j \neq i$  plays action  $s_j(\tau_j(\omega))$  in state  $\omega$ .

- Strategy profile  $s^*$  is Bayesian Nash equilibrium if each strategy  $s_i^*$  maximizes  $i$ 's expected payoff given  $s_{-i}^*$ .
- Alternative definition:  $s^*$  is Bayesian Nash equilibrium if for each player  $i$  and each type  $t_i$ , action  $s_i^*(t_i)$  maximizes  $t_i$ 's expected payoff  $\sum_{\omega \in \Omega} u_i(a_i, s_{-i}(\tau_{-i}(\omega)); \omega) \pi(\omega|t_i)$  over action set  $A_i$ , where  $\pi(\omega|t_i)$  is  $t_i$ 's updated belief of state  $\omega$ , given by Bayes' rule

$$\pi(\omega|t_i) = \frac{\pi(\omega)}{\sum_{\tilde{\omega} \in \Omega: \tau_i(\tilde{\omega})=t_i} \pi(\tilde{\omega})}.$$