

9 Games of Incomplete Information

9.1 Cournot Duopoly With Unknown Costs

Assume that two firms produce a homogenous good with inverse demand $P(Q) = \max\{2 - Q, 0\}$, where Q is interpreted as the total quantity of the good on the market. Also, assume that the firms produce the good at constant unit cost $c_1, c_2 \geq 0$.

The payoff function for firm i is thus

$$u_i(q_i, q_j) = P(q_1 + q_2) q_i - c_i q_i = [2 - q_1 - q_2 - c_i] q_i.$$

If costs would be known we would just have a simple (possibly asymmetric) static Cournot duopoly. However, assume that:

1. Firm 1 has cost $c_1 = 1$, which is common knowledge.
2. Firm 2, has cost $c_2 = \frac{1}{2}$ with probability $\frac{1}{2}$ and $c_2 = \frac{3}{2}$ with probability $\frac{1}{2}$.

Assume that q_1 is the quantity chosen by firm 1, then the optimal quantity for firm 2 is

$$q_2^L(q_1) = \arg \max \left[2 - q_1 - q_2 - \frac{1}{2} \right] q_2 = \arg \max \left[\frac{3}{2} - q_1 - q_2 \right] q_2$$

if $c_2 = \frac{1}{2}$ and

$$q_2^H(q_1) = \arg \max \left[2 - q_1 - q_2 - \frac{3}{2} \right] q_2 = \arg \max \left[\frac{1}{2} - q_1 - q_2 \right] q_2$$

if $c_2 = \frac{3}{2}$. From the first order conditions we have that the cost specific best replies are

$$\begin{aligned} q_2^L(q_1) &= \frac{\frac{3}{2} - q_1}{2} \\ q_2^H(q_1) &= \frac{\frac{1}{2} - q_1}{2} \end{aligned}$$

Firm 1 maximizes its expected profit given q_2^L, q_2^H

$$\begin{aligned} q_1(q_2^L, q_2^H) &= \arg \max_{q_1} \frac{1}{2} [2 - q_1 - q_2^L - 1] q_1 + \frac{1}{2} [2 - q_1 - q_2^H - 1] q_1 \\ &= \arg \max_{q_1} \left[1 - q_1 - \left(\frac{1}{2} q_2^L + \frac{1}{2} q_2^H \right) \right] q_1 \Rightarrow \\ q_1(q_2^L, q_2^H) &= \frac{1 - \left(\frac{1}{2} q_2^L + \frac{1}{2} q_2^H \right)}{2} \end{aligned}$$

In a Nash equilibrium we have that q_1 satisfies

$$\begin{aligned} q_1 &= q_1(q_2^L(q_1), q_2^H(q_1)) = \frac{2 - q_2^L - q_2^H}{4} = \frac{2 - \frac{\frac{3}{2} - q_1}{2} - \frac{\frac{1}{2} - q_1}{2}}{4} \\ &= \frac{2 - \frac{2 - 2q_1}{2}}{4} = \frac{2 - 1 + q_1}{4} \Leftrightarrow q_1 \frac{3}{4} = \frac{1}{4} \Leftrightarrow q_1^* = \frac{1}{3}. \end{aligned}$$

This is plausible since the expected cost is 1 for both firms, and $(\frac{1}{3}, \frac{1}{3})$ is the equilibrium if $c_1 = c_2 = 1$. We then have that

$$\begin{aligned} q_2^{*L} &= q_2^L(q_1^*) = \frac{\frac{3}{2} - q_1^*}{2} = \frac{\frac{3}{2} - \frac{1}{3}}{2} = \frac{7}{12} \\ q_2^{*H} &= q_2^H(q_1^*) = \frac{\frac{1}{2} - q_1^*}{2} = \frac{\frac{1}{2} - \frac{1}{3}}{2} = \frac{1}{12} \\ &\Rightarrow \frac{1}{2}q_2^{*L} + \frac{1}{2}q_2^{*H} = \frac{7}{24} + \frac{1}{24} = \frac{8}{24} = \frac{1}{3}, \end{aligned}$$

which confirms the intuition on the relationship with the complete information game.

9.2 Formal Treatment

Definition 1 A game of incomplete information is an object $G = (n, A, \Theta, u, p)$ where $A = \times_{i=1}^n A_i$ is the action space, $\Theta = \times_{i=1}^n \Theta_i$ is the type space, $u = (u_1, \dots, u_n)$, where

$$u_i : A \times \Theta \rightarrow R$$

is the payoff function and $p \in \Delta(\Theta)$ is the prior type distribution.

Remark 1 Definition builds in common prior assumption.

A pure strategy is a map $s_i : \Theta_i \rightarrow A_i$ and a mixed strategy would be a map $\sigma_i : \Theta \rightarrow \Delta(A_i)$. We also note that the beliefs over Θ_{-i} are given by

$$\Pr[\theta_{-i} | \theta_i] = \frac{\Pr[\theta_{-i}, \theta_i]}{\sum_{\theta'_{-i} \in \Theta_{-i}} \Pr[\theta'_{-i}, \theta_i]} = \frac{\Pr[\theta_{-i}, \theta_i]}{\Pr[\theta_i]}$$

Definition 2 A pure strategy Bayesian Nash equilibrium is a strategy profile $s^* = (s_1^*, \dots, s_n^*)$ such that

$$\begin{aligned} & \sum_{\theta_i \in \Theta_i} \left[\sum_{\theta_{-i} \in \Theta_{-i}} u_i(s_i^*(\theta_i), s_{-i}^*(\theta_{-i}), \theta) \Pr[\theta_{-i} | \theta_i] \right] \Pr[\theta_i] \\ & \geq \sum_{\theta_i \in \Theta_i} \left[\sum_{\theta_{-i} \in \Theta_{-i}} u_i(s_i(\theta_i), s_{-i}^*(\theta_{-i}), \theta) \Pr[\theta_{-i} | \theta_i] \right] \Pr[\theta_i] \end{aligned}$$

for every $i \in I$ and every $s_i : \Theta_i \rightarrow A_i$.

An (almost) equivalent formulation is

Definition 3 A pure strategy Bayesian Nash equilibrium is a strategy profile $s^* = (s_1^*, \dots, s_n^*)$ such that

$$\sum_{\theta_{-i} \in \Theta_{-i}} u_i(s_i^*(\theta_i), s_{-i}^*(\theta_{-i}), \theta) \Pr[\theta_{-i} | \theta_i] \Pr[\theta_i] \geq \sum_{\theta_{-i} \in \Theta_{-i}} u_i(s_i(\theta_i), s_{-i}^*(\theta_{-i}), \theta) \Pr[\theta_{-i} | \theta_i] \Pr[\theta_i]$$

for every $i \in I, \theta_i \in \Theta_i$ and $a_i \in A_i$.

9.3 Example: Double Auction

Assume that there is a seller and a buyer, where:

- Seller has valuation $\theta_S \sim U[0, 1]$
- Buyer has valuation $\theta_B \sim U[0, 1]$
- Buyer announces bid price $b \in R_+$
- Seller announces ask price $a \in R_+$
- If $b \geq a$, then the object is transacted at price

$$\frac{b + a}{2}.$$

- If $b < a$, the seller keeps object.

Note that in terms of generic notation we have that $\Theta_B = \Theta_S = [0, 1]$ and $A_B = A_S = R_+$

The payoff for the buyer is

$$u_B(b, a, \theta_B) = \begin{cases} \theta_B - \frac{b+a}{2} & \text{if } b \geq a \\ 0 & \text{if } b < a \end{cases}$$

and the payoff for the seller is

$$u_S(b, a, \theta_S) = \begin{cases} \frac{b+a}{2} - \theta_S & \text{if } b \geq a \\ 0 & \text{if } b < a \end{cases}$$

Fix θ_B and assume that $a : \Theta_S \rightarrow R_+$ is strictly increasing. Then a^{-1} exists and the fact that we get trade if and only if $b \geq a(\theta_S)$ is equivalent with $a^{-1}(b) \geq \theta_S$. Hence, the expected payoff for the buyer is

$$E[u_B(b, a(\theta_S), \theta_B)] = \int_0^{a^{-1}(b)} \left[\theta_B - \frac{b + a(\theta_S)}{2} \right] f(\theta_S) d\theta_S$$

Symmetrically

$$E[u_S(b(\theta_S), a, \theta_B)] = \int_{b^{-1}(a)}^1 \left[\frac{b(\theta_B) + a}{2} - \theta_S \right] f(\theta_B) d\theta_B.$$

Remark 2 *This kind of game has uncountable number of equilibria. We will only solve for one particular kind of equilibrium.*

9.3.1 Solving for a Linear Equilibrium

We guess that $a : \Theta_S \rightarrow R_+$ is given on form

$$a(\theta_S) = a_S + c_S \theta_S.$$

Remark 3 *We will NOT impose that strategies are linear. We will guess linearity and show that the best response to linear strategies are linear.*

Now

$$a^{-1}(x) = \frac{x - a_S}{c_S}.$$

Hence, the problem that defines the best response for a type θ_B buyer is

$$\max_{b \geq 0} \int_0^{\frac{b-a_S}{c_S}} \left[\theta_B - \frac{b + a(\theta_S)}{2} \right] d\theta_S = \max_{b \geq 0} \int_0^{\frac{b-a_S}{c_S}} \left[\theta_B - \frac{b + a_S + c_S \theta_S}{2} \right] d\theta_S$$

Cranking it out Use Leibnitz rule to get FOC

$$\begin{aligned}
0 &= \left[\theta_B - \frac{b + a_S + c_S \frac{b-a_S}{c_S}}{2} \right] \frac{d}{db} \left[\frac{b-a_S}{c_S} \right] + \int_0^{\frac{b-a_S}{c_S}} \left[\frac{d}{db} \left[\theta_B - \frac{b+a}{2} \right] \right] d\theta_S \\
&= \frac{\theta_B - b}{c_S} - \int_0^{\frac{b-a_S}{c_S}} \frac{1}{2} d\theta_S = \frac{\theta_B - b}{c_S} - \frac{1}{2} \frac{b-a_S}{c_S} \Leftrightarrow 0 = \theta_B - b - \frac{1}{2}b + \frac{1}{2}a_S \Leftrightarrow \\
\frac{3}{2}b &= \theta_B + \frac{1}{2}a_S \Leftrightarrow b = \frac{1}{3}a_S + \frac{2}{3}\theta_B,
\end{aligned}$$

WHICH IS A STRATEGY THAT IS LINEAR IN TYPE.

- $b(\theta_B) = a_B + c_B\theta_B$
- $\Rightarrow b^{-1}(x) = \frac{x-a_B}{c_B}$

Given a linear buyer strategy, we have that the seller wants to solve the problem

$$\max_{a \geq 0} \int_{b^{-1}(a)}^1 \left[\frac{b(\theta_B) + a}{2} - \theta_S \right] d\theta_B = \max_{a \geq 0} \int_{\frac{a-a_B}{c_B}}^1 \left[\frac{a_B + c_B\theta_B + a}{2} - \theta_S \right] d\theta_B$$

Again using Leibnitz rule, the FOC is

$$\begin{aligned}
0 &= \int_{\frac{a-a_B}{c_B}}^1 \frac{1}{2} d\theta_B - \left[\frac{a_B + c_B \frac{a-a_B}{c_B} + a}{2} - \theta_S \right] \frac{1}{c_B} \\
&= \left[1 - \frac{a-a_B}{c_B} \right] \frac{1}{2} - [a - \theta_S] \frac{1}{c_B} = \frac{[c_B + a_B - a]}{2} \frac{1}{c_B} - [a - \theta_S] \frac{1}{c_B} \\
\Leftrightarrow 0 &= \frac{[c_B + a_B - a]}{2} - [a - \theta_S] \\
\frac{3}{2}a &= \frac{c_B + a_B}{2} + \theta_S \Leftrightarrow \\
a &= \frac{c_B + a_B}{3} + \frac{2}{3}\theta_S
\end{aligned}$$

Hence, we started with hypothesized strategies

$$a(\theta_S) = a_S + c_S\theta_S$$

$$b(\theta_B) = a_B + c_B\theta_B$$

and derived best replies

$$a(\theta_S) = \frac{c_B + a_B}{3} + \frac{2}{3}\theta_S$$

$$b(\theta_B) = \frac{1}{3}a_S + \frac{2}{3}\theta_B$$

Hence

$$c_S = c_B = \frac{2}{3}$$

Moreover a_B, a_S solves the system

$$\begin{aligned} a_B &= \frac{1}{3}a_S \\ a_S &= \frac{c_B + a_B}{3} = \frac{\frac{2}{3} + a_B}{3} \end{aligned}$$

Hence

$$\begin{aligned} a_B &= \frac{1}{3}a_S \Leftrightarrow 3a_B = a_S = \frac{\frac{2}{3} + a_B}{3} \\ &\Rightarrow \\ 9a_B &= \frac{2}{3} + a_B \Leftrightarrow 8a_B = \frac{2}{3} \Leftrightarrow \\ a_B &= \frac{1}{12} \Rightarrow \\ a_S &= 3a_B = \frac{3}{12} = \frac{1}{4} \end{aligned}$$

So, we've verified that

$$\begin{aligned} a(\theta_S) &= \frac{1}{4} + \frac{2}{3}\theta_S \\ b(\theta_B) &= \frac{1}{12} + \frac{2}{3}\theta_B \end{aligned}$$

is an equilibrium in the double auction. DRAW!

9.4 Purification

9.4.1 Example

Consider “perturbed” game

	o	f
O	$2 + \varepsilon, 1$	$0, 0$
F	$0, 0$	$1, 2 + \delta$

where,

- $\varepsilon \sim U[0, x]$
- $\eta \sim U[0, x]$
- Eg $\Theta_1 = \Theta_2 = [0, x]$ with uniform i.i.d. priors.

Remark 4 Both pure strategy equilibria (O, o) and (F, f) are equilibria in extended game.

Remark 5 Will look for a pure strategy equilibrium in which each player plays both strategies with strictly positive probability.

Let;

- $q = \Pr[2 \text{ plays } f]$
- $p = \Pr[1 \text{ plays } O]$

We note that O is the best response for player 1 if and only if

$$\begin{aligned} (2 + \varepsilon)(1 - q) &\geq q \Leftrightarrow \\ \varepsilon &\geq \frac{q - 2(1 - q)}{1 - q} = \frac{3q - 2}{1 - q} \end{aligned}$$

By the same token f is a best response iff

$$\eta \geq \frac{3p - 2}{1 - p}$$

We conclude that the equilibrium strategies must have a cutoff property. Hence we'll look for a (Bayesian) Nash equilibrium where

$$\begin{aligned} s_1^*(\varepsilon) &= \begin{cases} O & \text{if } \varepsilon \geq a \\ F & \text{if } \varepsilon < a \end{cases} \\ s_2^*(\eta) &= \begin{cases} o & \text{if } \eta < b \\ f & \text{if } \eta \geq b \end{cases} \end{aligned}$$

It follows that

$$q = \Pr[\eta \geq b] = 1 - \int_0^b \frac{1}{x} d\eta = 1 - \frac{b}{x} = \frac{x-b}{x}$$

$$p = \frac{x-a}{x}$$

Simplifying we have that

$$a = \frac{x-3b}{b}$$

$$b = \frac{x-3a}{b}$$

$$\Rightarrow a = b$$

We conclude that

$$a = \frac{x-3a}{a} \Leftrightarrow a^2 = x-3a \Leftrightarrow$$

$$\left(a + \frac{3}{2}\right)^2 - \frac{9}{4} - x = 0$$

$$a = -\frac{3}{2} \pm \sqrt{\frac{9}{4} + x}$$

Since only a positive root can solve the problem we have that

$$a = b = -\frac{3}{2} + \sqrt{\frac{9}{4} + x},$$

and

$$p(x) = q(x) = \frac{x-b}{x} = \frac{x + \frac{3}{2} - \sqrt{\frac{9}{4} + x}}{x}$$

Using l'Hopitals rule we have that

$$\lim_{x \rightarrow 0} q(x) = \frac{1 - \frac{1}{2} \left(\frac{9}{4} + x\right)^{-\frac{1}{2}}}{1} = \lim_{x \rightarrow 0} \left[1 - \frac{\frac{1}{2}}{\sqrt{\left(\frac{9}{4} + x\right)}} \right] = 1 - \frac{\frac{1}{2}}{\frac{3}{2}} = \frac{2}{3}$$

That is, $(q(x), p(x)) \rightarrow \left(\frac{2}{3}, \frac{2}{3}\right)$, which is the mixed strategy equilibrium with complete information.

The point of the example is that it illustrates (somewhat sloppily stated):

Theorem 1 *For almost all payoffs there exists a sequence of incomplete information games so that for every equilibrium in the unperturbed game there is a sequence of pure strategy equilibria s.t the distribution over actions converge towards the distribution of the unperturbed game.*