### ECO 310: Empirical Industrial Organization Lecture 10: Models of Competition in Prices or Quantities: Conjectural Variations [2]

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#### Outline on today's lecture

- 1. Estimating CV parameters without data on MCs
- 2. Application: Genesove & Mullin
- 3. Conjectural variations with product differentiation

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# 1. Estimating CV parameters without data on MCs

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#### Estimating CV parameters without data on MCs

- So far, we have considered the estimation of CV parameters when the researcher knows both demand and firms' marginal costs.
- We now consider the case where the **researcher knows the demand**, **but it does not know firms' marginal costs**.
- Identification of CVs requires also de identification of MCs.
- Under some conditions, we can **jointly identify CVs and MCs** using the marginal conditions of optimality and the demand.

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#### Data

• Researcher observes data:

$$\mathsf{Data} = \left\{ \mathsf{P}_{t}, \; \mathsf{q}_{it}, \; \mathsf{X}_{t}^{\mathsf{D}}, \; \mathsf{X}_{t}^{\mathsf{MC}}: \; i = 1, ..., \mathsf{N}_{t}; \; t = 1, ..., \; \mathsf{T} 
ight\}$$

- $X_t^D$  are variables affecting consumer demand, e.g., average income, population.
- $X_t^{MC}$  are variables affecting marginal costs, e.g., some input prices.

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### Model: Demand and MCs

• Consider the linear (inverse) demand equation:

$$P_t = \alpha_0 + \alpha_1 X_t^D - \alpha_2 Q_t + \varepsilon_t^D$$

with  $\alpha_2 \geq 0$ , and  $\varepsilon_t^D$  is unobservable to the researcher.

• Consider the marginal cost function:

$$MC_{it} = \beta_0 + \beta_1 X_t^{MC} + \beta_2 q_{it} + \varepsilon_{it}^{MC}$$

with  $\beta_2 \geq 0$ , and  $\varepsilon_{it}^{MC}$  is unobservable to the researcher.

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#### Model: Profit maximization

• Profit maximization implies  $MR_{it} = MC_{it}$ , or equivalently:

$$P_t + \frac{dP_t}{dQ_t} \left[ 1 + CV_{it} \right] \ q_{it} = MC_{it}$$

• In the model above,  $\frac{dP_t}{dQ_t} = -\alpha_2$ . Therefore,

$$P_t - \alpha_2 \left[1 + CV_{it}\right] \ q_{it} = \beta_0 + \beta_1 \ X_t^{MC} + \beta_2 \ q_{it} + \varepsilon_{it}^{MC}$$

Or equivalently,

$$P_t = \beta_0 + \beta_1 X_t^{MC} + [\beta_2 + \alpha_2(1 + CV_{it})] q_{it} + \varepsilon_{it}^{MC}$$

 This equation describes the marginal condition for profit maximization. We assume now that CV<sub>it</sub> = CV for every observation *i*, *t* in the data.

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#### Complete structural model

• The structural equations of the model are:

Demand:  $P_t = \alpha_0 + \alpha_1 X_t^D - \alpha_2 Q_t + \varepsilon_t^D$ 

F.O.C.:  $P_t = \beta_0 + \beta_1 X_t^{MC} + [\beta_2 + \alpha_2(1 + CV)] q_{it} + \varepsilon_{it}^{MC}$ 

- Using this model and data, can we identify (estimate consistently, without asymptotic bias) the CV parameter?
- First, we will see that NO. In this model we cannot separately identify CV and MC.
- Second, we will see that a simple modification of this model implies separate identification of CV and MC.

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### Identification of demand parameters

$$\mathsf{Demand}: \ \ \mathsf{P}_t = \ \ lpha_0 + lpha_1 \ X^D_t - lpha_2 \ \ \mathsf{Q}_t + arepsilon_t^D$$

• Endogeneity problem: in equilibrium,  $cov(Q_t, \varepsilon_t^D) \neq 0$ .

- The model implies a valid instrument to estimate demand.
- In equilibrium,  $Q_t$  depends on  $X_t^{MC}$ . Note that  $X_t^{MC}$  does not enter in demand. If  $X_t^{MC}$  is not correlated with  $\varepsilon_t^D$ , then  $X_t^{MC}$  satisfies all the conditions for being a valid instrument.
- Parameters  $\alpha_0$ ,  $\alpha_1$ , and  $\alpha_2$  are identified using this IV estimator.

#### Identification of CV and MCs

$$F.O.C.: P_t = \beta_0 + \beta_1 X_t^{MC} + [\beta_2 + \alpha_2(1 + CV)] q_{it} + \varepsilon_{it}^{MC}$$

• Endogeneity problem: in equilibrium,  $cov(q_{it}, \varepsilon_{it}^{MC}) \neq 0$ .

- The model implies a valid instrument to estimate demand.
- In equilibrium,  $q_{it}$  depends on  $X_t^D$ . Note that  $X_t^D$  does not enter enter in the F.O.C. If  $X_t^D$  is not correlated with  $\varepsilon_{it}^{MC}$ , then  $X_t^D$  satisfies all the conditions for being a valid instrument.
- Parameters  $\beta_0$ ,  $\beta_1$ , and  $\gamma \equiv \beta_2 + \alpha_2(1 + CV)$  are identified using this IV estimator.

### The identification problem

F.O.C.: 
$$P_t = \beta_0 + \beta_1 X_t^{MC} + \gamma q_{it} + \varepsilon_{it}^{MC}$$

- Note that we can identify the parameter  $\gamma$ , where  $\gamma \equiv \beta_2 + \alpha_2(1 + CV)$ , and the slope of inverse demand function,  $\alpha_2$ .
- However, knowledge of  $\gamma$  and  $\alpha_2$  is not sufficient to identify separately CV and the slope of the MC,  $\beta_2$ .
- Suppose that  $\gamma = 1$  and  $\alpha_2 = 0.4$ , such that we have the constraint:

$$1 = \beta_2 + 0.4 \ (1 + CV)$$

 This equation is satisfied by any of the following: [Perfect competition] CV = -1 and β<sub>2</sub> = 1.0 [Cournot] CV = 0 and β<sub>2</sub> = 0.6 [Cartel, with N = 3] CV = N - 1 = 2 and β<sub>2</sub> = 0.2

# The identification problem [2]

- The IV estimator identifies the MC by using the instrument  $X_t^D$  that shifts the demand.
- When we make an assumption about the form of competition, shifts in the demand curve are able to trace out the marginal cost curve, i.e., to identify the MC parameters.
- However, without specifying the form of competition, shifts in the demand alone are not sufficient to separately identify MC and CV.
- Let  $\hat{q}_{it}(X_t^D)$  be the part of  $q_{it}$  explained  $X_t^D$ . When  $X_t^D$  varies, we see a positive correlation between  $P_t$  and  $\hat{q}_{it}(X_t^D)$ . But the magnitude of this correlation can be explained by the combination of:
  - either zero/negative CV and positive and large  $\beta_2$ ;
  - or positive CV and small or zero  $\beta_2$ .

## The identification problem [3]



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## The identification problem [4]



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## Solving the identification problem

- Solving the identification problem involves generalizing demand so that changes in exogenous variables do **more than just parallel shift** the demand curve and MR.
- In particular, we need to allow for additional exogenous variables that are capable of **rotating** the demand curve as well.
- "Demand Rotators" are exogenous variables affecting the slope of the demand curve:

## Solving the identification problem [2]



- Note that  $E_1$  could be an equilibrium either for a perfectly competitive industry with cost  $MC_c$  or for a monopolist with cost  $MC_m$ .

- There is no observable distinction between the hypotheses of competition and  $\ensuremath{\mathsf{occ}}$ 

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Competition

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## Solving the identification problem [3]



- Now, rotate the demand curve to  $D_2$ , with  $MR_2$ 

- Competitive equilibrium stays at  $E_1$ . But monopoly equilibrium moves to  $E_2$ 

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## Solving the identification problem [4]

• Consider now the following demand equation:

$$P_t = \alpha_0 + \alpha_1 X_t^D - \alpha_2 Q_t - \alpha_3 [R_t Q_t] + \varepsilon_t^D$$

- *R<sub>t</sub>* is an observable variable that affects the slope of the demand, i.e., the price of a substitute or complement product.
- Key condition:  $\alpha_3 \neq 0$ .
- That is, when  $R_t$  varies, there should be rotation (i.e., change in the slope of the demand curve).

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## Solving the identification problem

• Given this demand model, we have that:

$$\frac{dP_t}{dQ_t} = -\alpha_2 - \alpha_3 R_t$$

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• And the F.O.C. for profit maximization

$$P_t + \frac{dP_t}{dQ_t} \left[ 1 + CV \right] \ q_{it} = MC_{it}$$

become:

$$P_t + (-\alpha_2 - \alpha_3 R_t) \quad [1 + CV] \quad q_{it} = MC_{it}$$

or equivalently:

$$P_t = MC_{it} + (\alpha_2 + \alpha_3 R_t) [1 + CV] q_{it}$$

## Solving the identification problem [6]

• Combining this F.O.C. with the MC function,  $MC_{it} = \beta_0 + \beta_1 X_t^{MC} + \beta_2 q_{it} + \varepsilon_{it}^{MC}$ , we have:

$$P_t = \beta_0 + \beta_1 X_t^{MC} + \beta_2 \ q_{it} + (\alpha_2 + \alpha_3 \ R_t) \ [1 + CV] \ q_{it} + \varepsilon_{it}^{MC}$$

• That we can represent using the following regression model:

$$P_t = \beta_0 + \beta_1 X_t^{MC} + \gamma_1 \ q_{it} + \gamma_2 \ (R_t \ q_{it}) + \varepsilon_{it}^{MC}$$
  
with  $\gamma_1 \equiv \beta_2 + \alpha_2 \ [1 + CV]$  and  $\gamma_2 \equiv \alpha_3 \ [1 + CV]$ .

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## Solving the identification problem [7]

• The structural equations of the model are:

Demand: 
$$P_t = lpha_0 + lpha_1 \; X^D_t - lpha_2 \; Q_t - lpha_3 \; [R_t \; Q_t] + arepsilon^D_t$$

$$F.O.C.: \quad P_t = \beta_0 + \beta_1 X_t^{MC} + \gamma_1 q_{it} + \gamma_2 (R_t q_{it}) + \varepsilon_{it}^{MC}$$

• Using this model and data, we can identify separately CV and MC parameters.

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### Identification of demand parameters

Demand: 
$$P_t = \alpha_0 + \alpha_1 X_t^D - \alpha_2 Q_t - \alpha_3 [R_t Q_t] + \varepsilon_t^D$$

• Endogeneity problem: in equilibrium,  $cov(Q_t, \varepsilon_t^D) \neq 0$ .

- The model implies a valid instrument to estimate demand.
- In equilibrium,  $Q_t$  depends on  $X_t^{MC}$ . Note that  $X_t^{MC}$  does not enter in demand. If  $X_t^{MC}$  is not correlated with  $\varepsilon_t^D$ , then  $X_t^{MC}$  satisfies all the conditions for being a valid instrument.
- Parameters  $\alpha_0$ ,  $\alpha_1$ ,  $\alpha_2$ , and  $\alpha_3$  are identified using this IV estimator.

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#### Identification of CV and MCs

$$F.O.C.: P_t = \beta_0 + \beta_1 X_t^{MC} + \gamma_1 q_{it} + \gamma_2 (R_t q_{it}) + \varepsilon_{it}^{MC}$$

• Endogeneity problem: in equilibrium,  $cov(q_{it}, \varepsilon_{it}^{MC}) \neq 0$ .

- The model implies a valid instrument to estimate demand.
- In equilibrium,  $q_{it}$  depends on  $X_t^D$ . Note that  $X_t^D$  does not enter enter in the F.O.C. If  $X_t^D$  is not correlated with  $\varepsilon_{it}^{MC}$ , then  $X_t^D$  satisfies all the conditions for being a valid instrument.
- Parameters  $\beta_0$ ,  $\beta_1$ ,  $\gamma_1$ , and  $\gamma_2$  are identified.

## Identification of CV and MCs [2]

F.O.C.: 
$$P_t = \beta_0 + \beta_1 X_t^{MC} + \gamma_1 q_{it} + \gamma_2 (R_t q_{it}) + \varepsilon_{it}^{MC}$$

Note that:

$$\gamma_1 = \beta_2 + \alpha_2 \left[ 1 + CV \right]$$

$$\gamma_2 = \alpha_3 [1 + CV]$$

- It is clear that given  $\gamma_2$  and  $\alpha_3$ , we identify CV.
- And given  $\gamma_1$ ,  $\alpha_2$ , and CV we identify  $\beta_2$ .

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## Identification of CV and MCs [3]

F.O.C.: 
$$P_t = \beta_0 + \beta_1 X_t^{MC} + \gamma_1 q_{it} + \gamma_2 (R_t q_{it}) + \varepsilon_{it}^{MC}$$

with

$$\gamma_2 = \alpha_3 \left[ 1 + CV \right]$$

• The identification of CV is very intuitive:  $1 + CV = \gamma_2/\alpha_3$ . It measures the ratio between the sensitivity of  $P_t$  with respect to  $(R_t \ q_{it})$  in the F.O.C. and the sensitivity of  $P_t$  with respect to  $(R_t \ Q_t)$  in the demand.

• Example: 
$$\alpha_3 = 0.5$$
 and  $N = 3$ .  
[Perfect competition]  $CV = -1$  such that  $\gamma_2/\alpha_3 = 0$   
[Cournot]  $CV = 0$  such that  $\gamma_2/\alpha_3 = 1/0.5 = 2$   
[Cartel, with  $N = 3$ ]  $CV = N - 1 = 2$  such that  $\gamma_2/\alpha_3 = 2/0.5 = 4$ 

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## 2. Empirical application: Genesove & Mullin

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#### An Application: US sugar industry 1890-1914

- Genesove and Mullin (GM) study competition in the US sugar industry during the period 1890-1914.
- Why this period? High quality information on the value of marginal costs because:

(1) the production technology of refined sugar during this period was very simple;

(2) there was an important investigation of the industry by the US anti-trust authority. As a result of that investigation, there are multiple reports from expert witnesses that provide estimates about the structure and magnitude of production costs in this industry.

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#### The industry

- Homogeneous product industry.
- Highly concentrated during 1890-1914. The industry leader, American Sugar Refining Company (ASRC), had more than 65% of the market share during most of these years.
- Refined sugar companies buy "raw sugar" from suppliers in national or international markets, transformed it into refined sugar, and sell it to grocers.

#### Production technology.

- Raw sugar is 96% sucrose and 4% water. Refined sugar is 100% sucrose. Process of transforming raw sugar into refined sugar is called "melting".
- Industry experts reported that the industry is a "fixed coefficient" production technology

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#### Production technology: Costs

• "Fixed coefficient" production technology

$$q^{refined} = \lambda \, q^{raw}$$

where  $q^{refined}$  is refined sugar output,  $q^{raw}$  is the input of raw sugar, and  $\lambda \in (0, 1)$  is a technological parameter.

• Marginal cost function. Given this production technology, the marginal cost function is:

$$\mathit{MC} = \mathit{c}_0 + rac{1}{\lambda} \; \mathit{P^{raw}}$$

- *P<sup>raw</sup>* is the price of the input raw sugar (in dollars per pound).
- c<sub>0</sub> is a component of the marginal cost that depends on labor and energy.

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## Production technology: Costs [2]

$$MC=c_{0}+rac{1}{\lambda} P^{raw}$$

- Industry experts unanimously report that the value of the parameter  $\lambda$  was close to 0.93, and  $c_0$  was around \$0.26 per pound.
- Therefore, the marginal cost at period (quarter) *t*, in dollars per pound of sugar, was:

$$MC_t = 0.26 + 1.075 P_t^{raw}$$

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#### Data

- Quarterly US data for the period 1890-1914.
- The dataset contains 97 quarterly observations on industry output, price, price of raw sugar, imports of raw sugar, and a seasonal dummy.

$$\mathsf{Data} = \{ \ Q_t, \ P_t, \ P_t^{\mathsf{raw}}, \ \mathit{IMP}_t, \ S_t: t = 1, 2, ..., 97 \}$$

- *IMP*<sub>t</sub> represents the imports of raw sugar from Cuba.
- And  $S_t$  is a dummy variable for the Summer season:  $S_t = 1$  is observation t is a Summer quarter, and  $S_t = 0$  otherwise.
- The summer was a high demand season for sugar because most the production of canned fruits was concentrated during that season, and the canned fruit industry accounted for an important fraction of the demand of sugar.

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#### Estimates of demand parameters

• GM estimate four different models of demand. The main results are consistent for the four models. Here I concentrate on the linear demand.

$$Q_t = \beta_t \ (\alpha_t - P_t)$$

• GM consider the following specification for  $\alpha_t$  and  $\beta_t$ :

$$\alpha_t = \alpha_L (1 - S_t) + \alpha_H S_t + e_t^D$$

$$\beta_t = \beta_L (1 - S_t) + \beta_H S_t$$

- $\alpha_L$  and  $\beta_L$  are the intercept and the slope of the demand during the "Low Season" (when  $S_t = 0$ ).
- And  $\alpha_H$  and  $\beta_H$  are the intercept and the slope of the demand during the "High Season" (when  $S_t = 1$ ).

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#### Estimates of demand parameters

Demand Estimates				
Parameter	Estimate	Standard Error		
αL	5.81	(1.90)		
α <sub>H</sub>	7.90	(1.57)		
$\beta_{I}$	2.30	(0.48)		
$\beta_{H}$	1.36	(0.36)		

[2]

- According to these estimates, in the high season the demand shifts upwards but it also becomes more inelastic.
- The estimated price elasticities of demand in the low and the high season are  $\varepsilon_L = 2.24$  and  $\varepsilon_H = 1.04$ , respectively.
- According to this, any model of oligopoly competition where firms have some market power predicts that the price cost margin should victor Aguiregabilita ()

#### Estimates of demand parameters

- Importantly, the seasonality in the demand of sugar introduces a "rotator" in the demand curve.
- The slope of the demand curve is steeper in the high season than in the low season.

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#### Estimates of CVs

• Given this estimated demand and the MCs, they obtain the following estimate of the CV parameter  $\theta \equiv \frac{CV}{N}$ .

Estimate of CV Parameter Estimate (s.e.)

$$\theta \equiv \frac{CV}{N} \qquad 0.10 \quad (0.02)$$

- Therefore, we can reject the null hypothesis of Cournot competition in favor of some form of collusion.
- Next week, we will see how it is possible to identify CVs even when the researcher does not have data on marginal costs, and needs to estimate marginal costs together with CVs.

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#### Genesove and Mullin (1998) - Costs and Conduct

 GM specify a constant-cost Marginal Cost function for US sugar producers

$$MC_t = \beta_0 + \beta_1 P_t^{RAW} + \beta_2 q_t + \varepsilon_t^{MC}$$

• The MR = MC condition yields:

$$P_t = eta_0 + eta_1 \ P_t^{RAW} + \gamma_1 \ q_t + \gamma_2 \ (S_t \ q_t) + arepsilon_{it}^{MC}$$

where

$$\gamma_{1} = \beta_{2} + \frac{1}{\beta_{L}} [1 + CV]$$
  
$$\gamma_{2} = \left(\frac{1}{\beta_{H}} - \frac{1}{\beta_{L}}\right) [1 + CV]$$

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#### Genesove and Mullin (1998) - Costs and Conduct

• GMs estimates of the Supply parameters.

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	Estimate	Direct Measure
CV/N	0.038 (0.024)	0.100
$\beta_0$	0.466 (0.285)	0.260
$\beta_1$	1.052 (0.085)	1.075

- Estimated cost parameters not too far from their "direct measures" which seems to validate CV approach.
- Based on the estimates of *CV* / *N*, the *predicted values* for the Lerner index in the low and in the high season are:

$$II = 1.7\%_{Competition}$$
  $II = 3.6\%$  November 25, 2019 37 / 39

# 3. Conjectural variations with product differentiation

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CVs with product differentiation

CVs with product differentiation

• See Problem set #2

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