## Applications of dynamic game model

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#### Outline

Hospital industry: Gowrisankaran and Town (1997)

Cement industry: Ryan (2012)

Hard disk drive industry: Igami & Uetake (2020)

Summary of estimation methodology of dynamic game model

Reference

## Hospital industry: Gowrisankaran and Town (1997)

- Two types of hospitals (players): for-profit (FP) and non-profit (NP).
- Three types of patients in a market: with Medicare (MD, $\Theta_{MD}$ ), private insurance (PI,  $\Theta_{PI}$ ), and no insurance (UI,  $\Theta_{UI}$ ).
- Hospitals makes static decisions (price) and dynamic decisions (investment, exit and entry)

#### Patient's decision

- In a market, a fixed population M is ill. So each type of patients has total number: Θ<sub>MD</sub>M, Θ<sub>PI</sub>M, Θ<sub>UI</sub>M.
- Hospital j is free to set  $p_j^{NM}$ , the base price that it charges for non-medicare patients.
- Poor patients of all medicare types don't have to necessarily pay the full price.
   The price that each patient must pay is:

$$\begin{aligned} p_{ij}^{MD} &= \max \left\{ \min \left( d, Y_i^{MD} - Y_{MIN} \right), 0 \right\} \\ p_{ij}^{PI} &= \max \left\{ \min \left( c p_j^{NM}, Y_i^{PI} - Y_{MIN} \right), 0 \right\} \\ p_{ij}^{UI} &= \max \left\{ \min \left( p_i^{NM}, Y_i^{UI} - Y_{MIN} \right), 0 \right\} \end{aligned}$$

## Patient's decision

Hence, the utility for patient i of type T choosing hospital j is

$$U_{ii}^T = k_i + \gamma_1 \ln(Y_i^T - \gamma_2 p_{ii}^T) + \epsilon_{ii}^T$$

• The utility for patient i of type T buying outside good is

$$U_{i0}^T = \gamma_1 \ln(Y_i^T - \gamma_2 p_{i0}^T) + \epsilon_{i0}^T$$

• We can obtain the probability (share) of patient i choosing hospital j:

$$s_{ij}^{T}(p_{ij}^{T}) = \frac{\exp\left\{k_{j} + \gamma_{1} \ln\left(\frac{Y_{i}^{T} - \gamma_{2} p_{ij}^{T}}{Y_{i}^{T} - \gamma_{2} p_{i0}^{T}}\right)\right\}}{1 + \sum_{k=1}^{J} \left[k_{k} + \exp\left(\frac{Y_{i}^{T} - \gamma_{2} p_{i0}^{T}}{Y_{i}^{T} - \gamma_{2} p_{i0}^{T}}\right)\right]}$$

## Hospital's static production decision

- Each hospital j has identical fixed costs F and differing marginal costs (on each patient)  $mc_i = \bar{m}c + bk_i$
- Hence, the gross variable profit of j can be constructed as  $\pi_i^{GV}(p^{NM})$
- Hospitals' static decision is to choose  $p_i^{NM}$  to maximize the static profits.

## Hospital's dynamic decision

- State variable vector  $s_t = \{k_{it}\}_{i=1,2..n}$
- Timing: in each period, players take actions in this order:
  - Existing hospitals decides whether to exit. Each firm simultaneously receives an i.i.d. scrap value draw from a uniform distribution  $U(\phi-\sigma_\phi,\phi+\sigma_\phi)$ , and immediately decides whether to accept its scrap value draw and exit forever or stay alive in the current period.
  - Then, the remaining incumbents simultaneously invest. Each hospital j invests amount  $x_{it}$  at a cost of  $c_i^{NP}$  or  $c_i^{FP}$ .
  - Next, potential entrants decide whether to enter. There are two potential entrants in each period, one NP and one FP. Each of them receive a sunk cost from distribution  $U(S \sigma_S, S + \sigma_S)$ . If it enters, it would have an initial quality in the next period.
  - Finally, a static game starts.

## Transition between states

• Hospital's investment converts to quality in a certain way:

$$k_j = g(w_j)$$
 $w_{j,t+1} = w_{jt} + v_{jt} - \bar{v}_t$ 
 $v_j = \left\{ egin{array}{l} 1, ext{with prob. } ax_{jt}/(1+ax_{jt}) \ 0, ext{with prob. } 1/(1+ax_{jt}) \end{array} 
ight.$ 
 $ar{v} = \left\{ egin{array}{l} 1, ext{with prob. } \delta \ 0, ext{with prob. } 1-\delta \end{array} 
ight.$ 

#### Estimation method

#### Nested fixed-point algorithm (NFXP)'s central idea:

- Given a structural parameter vector, solve the Bellman equation, get the value function.
- Based on the solution, compute the conditional choice probabilities (CCPs) of actions, examine how well these CCPs match observations.
- Find a parameter vector that causes the predictions to most closely match the data.

Cement industry: Ryan (2012)

#### Some features of cement market:

- Cement is a fine mineral dust with binding properties, the key ingredient of concrete. Its production generates significant emissions, making it a frequent target of environmental regulations.
- As a concentrated industry, sunk costs of entry and costly investment are important determinants of market structure. 1990 Clean Air Act (CAA) amendment affects the cost structure.
- Cement is a largely homogeneous commodity.
- Cement easily absorbs water from air, making storage expensive. Short shipping distance makes cement markets quasi-independent geographically.

## Model's setup

- Model's key setup:
  - firm's static decision: quantity of output
  - firm's dynamic decision: exit or not, enter or not, invest/divest or not
  - state: firms' capacity in the market. Firms' actions affect the state in the next period.
- In each period, the sequence of events are:
  - incumbent firms decide whether to exit the industry
  - potential entrants and incumbents who decided not to exit simultaneously make entry and investment decisions.
  - incumbent firms compete over quantities in the market.

## Firms' static decision

- Each market is full described by the state vector  $s_t$ .  $s_{it}$  is firm i's capacity in t.
- In each regional market m, firms face a constant elasticity of demand curve

$$\ln Q_m(\alpha) = \alpha_{0m} + \alpha_1 \ln P_m,$$

• the cost of production is:

$$C_i(q_i;\delta) = \delta_0 + \delta_1 q_i + \delta_2 1(q_i > \nu s_i)(q_i - \nu s_i)^2,$$

where  $\delta_0$  is fixed cost,  $\delta_1$  is constant marginal cost. If output level  $q_i$  reaches a threshold  $\nu s_i$ , it would cause extra costs.

• Then, in each period, solve the static Cournot competition, firm i's profit is determined:  $\bar{\pi}_i(s; \alpha, \delta)$ 

## Firms' investment decision

• Recall that firms can adjust their capacity by investment or divestment  $x_i$ . The cost of adjustments is:

$$\Gamma(x_i; \gamma) = 1(x_i > 0)(\gamma_{i1} + \gamma_2 x_i + \gamma_3 x_i^2) + 1(x_i < 0)(\gamma_{i4} + \gamma_5 x_i + \gamma_6 x_i^2)$$

if invest,  $x_i > 0$ , fixed cost  $\gamma_1$  is drawn each period from the common distribution  $F_{\gamma}$   $N(\mu_{\gamma}^+, \sigma_{\gamma}^+)$ . if divest,  $x_i < 0$ , fixed cost  $\gamma_4$  is drawn each period from the common distribution  $G_{\gamma}$   $N(\mu_{\gamma}^-, \sigma_{\gamma}^-)$ .

## Firms' entry and exit decision

Sunk cost of entry and scrap value

$$\Phi_i(a_i; k_i, \phi_i) = \begin{cases} -k_i, & (a_i = enter) \\ \phi_i, & (a_i = exit) \end{cases}$$
 (1)

fixed cost of entry  $k_i$  is private information, drawn from common distribution  $F_k$ .  $\phi_i$  is the payment firms receive when exiting the market (scape value), drawn from common distribution  $F_{\phi}$ .

Collecting all the parameters, firm i's per-period payoff function is

$$\pi_i(\mathbf{s}, \mathbf{a}; \theta) = \pi_i(\mathbf{s}, \mathbf{a}; \alpha, \delta, \gamma_i, k_i, \phi_i) = \bar{\pi}_i(\mathbf{s}; \alpha, \delta) - \Gamma(x_i; \gamma_i) + \Phi_i(\mathbf{a}_i; k_i, \phi_i)$$

## Transitions between states

• actions  $a_i$  (investment, entry and exit) in t take one period to occur in  $s_{t+1}$  , such as  $s_{i,t+1} = s_{it} + x_{it}$ 

## Equilibrium

 In a Markovian setting, firms only condition on the current state and their private shocks when making decisions.

Each firm's strategy  $a_i = \sigma(s, \epsilon_i)$ 

Then, we get value functions of incumbents and potential entrants

$$V_i(\mathbf{s}; \sigma(\mathbf{s}), \theta, \epsilon_i)$$

 MPNE requires each firm's strategy profile is the best response of competitors' strategy profiles.

#### Estimation method

- Use Bajari, Benkard, and Levin (2007)'s two-step estimation method (BBL)
- Intuition of BBL: the econometrician lets the agents in the model solve the dynamic program, and finds parameters of the underlying model such that their behavior is optimal.

## **Applications**

- $\theta = \{\alpha, \delta, \gamma, k, \phi\}$
- 1<sup>st</sup> step: recover the policy functions governing entry, exit and investment, along with the production relevant parameters  $\Rightarrow$  get  $\sigma(s)$ ,  $\alpha$ ,  $\delta$
- $2^{nd}$  step: take policy functions and restrictions of MPNE to recover the dynamic parameters  $\Rightarrow$  get  $\mu_{\gamma}^+, \sigma_{\gamma}^+, \mu_{\gamma}^-, \sigma_{\gamma}^-, \mu_{\phi}, \sigma_{\phi}, \mu_k, \sigma_k$

#### Results

- Finding: the CAA increased the mean entry cost by 22 percent, with no significant change on variance. Moreover, the increase in entry costs greatly reduces the chances that marginal firms enter a market, and this has significant effects on product market competition.
- Counterfactual analysis: As a result of entry rates, the overall welfare in the US cement market decreased by at least \$810M.

## Hard disk drive industry: Igami & Uetake (2020)

- Antitrust towards mergers is the most important area in which IO economists shape the debate on policy.
- Conventional merger analysis takes a proposed merger as given and focuses on its immediate effects on competition. Such a static analysis could be appropriate only if mergers were completely random events. However, mergers can be endogenous!

## Model's setup

- Time is discrete with a finite horizon, t = 0, 1, 2..., T. The final period T is the time at which the demand for HDDs becomes zero (the industry comes to an end).
- Finite number of incumbent firms,  $i=1,2...,n_t$ . Potential entrant i=0 Each firm has productivity on a discretized grid with unit interval  $\omega_{it} \in \{\omega^1, \omega^2, ...\}$ , representing its tacit knowledge.
- State variable is  $\omega_t$
- A potential entrant exists in every period a<sup>0</sup> ∈ A<sup>0</sup> = {enter, out}.
  a<sup>0</sup> has sunk cost k<sup>a<sup>0</sup></sup> + ε(a<sub>it</sub><sup>0</sup>).
  An incumbent take actions in every period a ∈ A = {exit, innovate, (propose merger to j)<sub>ji</sub>, (innovate and propose to j)<sub>ji</sub>, idle}.
  a has sunk cost k<sup>a</sup> + ε(a<sub>it</sub>).

#### Transitions between state

- If incumbent i  $a_{it} = exit$ , then  $\omega_{i,t+1} = \omega^{00}$  ("dead")
- If  $a_{it} = innovate$ , then  $\omega_{i,t+1} = \omega_{it} + 1$
- If  $a_{it} = \text{propose merger to j, and firm } j$  takes the deal, then  $\omega_{i,t+1} = \max \{\omega_{it}, \omega_{jt}\} + \triangle_{i,t+1}$ .  $\triangle_{i,t+1}$  reflects "synergies".

## **Timing**

# Instead of assuming players simultaneous move in each period, here the authors consider an alternating-move game, in which the time interval is relatively short and only one firm can make a dynamic discrete choice within a period.

- The timeline of the stochastically alternating moves:
  - At the beginning of each period, nature equally chooses one firm to make a move.
  - If it's i's turn to make a move, it observes the current industry state  $\omega_t$ , draws sunk costs of actions.
  - Based on these information, i makes the discrete choice a<sub>it</sub> ∈ A<sub>it</sub>. Corresponding sunk costs immediately occurs.
     If i negotiate a merger agreement with j, a acquisition price p<sub>ij</sub> would be paid if the deal is made. It's a take-it-or-leave-it offer.
  - All active firms participate static competition, earn period profits and pay costs of operation.

## Dynamic optimization and equilibrium

• To reach an incumbent firm *i*'s dynamic optimization, the corresponding Bellman equation is:

$$V_{it}(\omega_t, \epsilon_{it}) = \pi_{it}(\omega_t) - \phi_t(\omega_{it}) + \max\left\{\bar{V}_{it}^x, \bar{V}_{it}^c, \bar{V}_{it}^i, \bar{V}_{ijt}^m, \bar{V}_{ijt}^{i\&m}\right\}$$

 $\bar{V}^a_{it}$  represents conditional values of exiting, idling, innovating, proposing merger to rival j, and both merge and innovate. Such as:

$$\bar{V}_{it}^{c}(\omega_{t}, \epsilon_{it}^{c}) = \epsilon_{it}^{c} + \beta E\left[V_{i,t+1}(\omega_{t+1})|\omega_{t}, a_{it} = idle\right],$$

#### Estimation method

#### Conditional choice probability (CCP) based method

- 1<sup>st</sup> step: estimate reduced-form CCP function by using data actions and states.
- 2<sup>nd</sup> step: Using these CCP functions, calculate value functions, find optimal structural parameters.

## Summary of estimation methodology

- Full solution methods
  - Nested fixed point algorithm (Gowrisankaran and Town, 1997)
  - · High precision but computationally expensive
- BBL (Ryan, 2012)
  - Computationally efficient
- CCP (Igami and Uetake, 2020)
  - · Computationally efficient

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