

Estimating Discrete-Choice Models of Product Differentiation

Steven Berry (1994), *RAND Journal of Economics*

Chenhui Lu

University of Florida

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- ④ Contributions
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Background: General Problem

- In differentiated product markets, reliable estimation of **demand and costs** is challenging.
- Prices are typically **correlated with unobserved product characteristics** (quality, brand perception, style) \Rightarrow **endogeneity**.
- Ignoring endogeneity yields biased estimates (e.g., spurious findings that higher prices increase demand).
- **Societal relevance**: accurate demand is essential for antitrust, merger policy, and welfare analysis.

Specific Research Problem

- How to model consumer choice with **product differentiation** in a discrete-choice framework.
- How to address **price endogeneity** in demand estimation.

Why Is It Important?

- In IO, Public, **demand elasticities** are central for policy and welfare analysis.
- Traditional methods for homogeneous goods fail in differentiated settings.
- This paper introduces a method that became the foundation for the **BLP framework** (Berry, Levinsohn, Pakes 1995).

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Consumer Demand Model

- Consumers choose among N products and an outside good.
- Indirect utility:

$$u_{ij} = x_j\beta - \alpha p_j + \xi_j + \epsilon_{ij}$$

- Components: x_j observed characteristics; p_j price ($\alpha > 0$); ξ_j unobserved quality; ϵ_{ij} error term.
- Each consumer chooses the product with the highest utility.

The Endogeneity Problem

- Unobserved quality ξ_j affects both **demand** and **price**.
- Better products tend to be priced higher $\Rightarrow p_j$ and ξ_j are correlated.
- Direct regressions of market share on price are biased (sometimes suggesting consumers prefer higher prices).
- **Key challenge**: separate the true effect of price from unobserved quality.

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Innovation 1: Mean Utility Inversion

- Market shares s_j map **one-to-one** to mean utilities δ_j .
- Define mean utility:

$$\delta_j = x_j\beta - \alpha p_j + \xi_j$$

- **Inversion** recovers δ_j from observed shares.
- Special cases:
 - **Logit**: $\delta_j = \ln(s_j) - \ln(s_0)$
 - **Nested logit**: includes a within-group term
- Interpretation: shares contain information on product attractiveness; inversion recovers latent utility.

Innovation 2: IV Estimation

- With δ_j in hand, estimate

$$\delta_j = x_j\beta - \alpha p_j + \xi_j$$

- Endogeneity remains: p_j correlated with ξ_j .
- **Instrumental variables (IV):**
 - Cost shifters (input prices, wages, transportation costs)
 - Rival product characteristics (affect equilibrium pricing but not utility directly)

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Special Cases of Inversion

Logit model

$$\ln(s_j) - \ln(s_0) = x_j\beta - \alpha p_j + \xi_j$$

Simplest closed-form inversion.

Vertical differentiation

- Products are ordered by quality; market shares defined by cutoff rules.

Nested logit

- Allows correlation within groups; adds $\sigma \ln(s_{j|g})$.

Random coefficients logit

- Most general: consumer heterogeneity interacts with characteristics.
- Market shares approximated via simulation.

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Supply Side (Firm Pricing)

- Assume **Bertrand Nash** competition.
- First-order condition (FOC):

$$p_j = c_j + \text{markup}_j$$

- Markups depend on demand elasticities:
 - Few substitutes \Rightarrow higher markups
 - Intense competition \Rightarrow lower markups

Joint Estimation Framework

Steps

- ① Invert observed shares to recover δ_j .
- ② Estimate demand parameters (β, α) via IV.
- ③ Use supply-side pricing FOC to identify cost parameters.
- ④ **Joint GMM**: combine demand and supply moments to estimate the full parameter vector.
- ⑤ Outcome: preference and cost estimates that enable **counterfactual policy analysis**.

Advantages of the Approach

Flow

(Observed shares s_j) \rightarrow [Inversion: $\delta_j(s)$] \rightarrow [IV
Regression $\Rightarrow \beta, \alpha$]

Demand moments: $\delta_j = x_j\beta - \alpha p_j + \xi_j$ Supply moments:
 $p_j = c_j(\gamma) + \text{markup}$

\Downarrow Joint GMM Estimation \Rightarrow Demand & Cost Parameters
 \Rightarrow Counterfactuals

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Monte Carlo Experiments

- Simulation of 500 duopoly markets.

TABLE 1 Monte Carlo Parameter Estimates 100
Random Samples of 500 Duopoly Markets
Logit Utility

Parameter	True Value	$(\sigma_{\epsilon d} = 1)$		$(\sigma_{\epsilon d} = 3)$	
		(1) OLS	(2) IV	(3) OLS	(4) IV
β_o	5	3.46 (.158)	4.98 (.226)	0.378 (.415)	4.89 (.738)
β_x	2	1.41 (.058)	1.99 (.091)	.325 (.127)	1.95 (.272)
α	1	.726 (.029)	.995 (.039)	.181 (.076)	.979 (.128)

Notes: The values given in the table are empirical means and (standard errors).

The utility function is $u_{ij} = \beta_o + \beta_x x_{ij} + \sigma_{\epsilon d} \epsilon_{ij} - \alpha p_j + \epsilon_{ij}$.

Marginal cost is $c_j = e^{\gamma_o + \gamma_x x_j + \sigma_{\epsilon c} \epsilon_{cj} + \gamma_w w_j + \sigma_{\epsilon w} \epsilon_{wj}}$.

- OLS**: severely underestimates the price coefficient; may imply consumers prefer higher prices.
 - IV**: recovers true parameters consistently.
- Demonstrates that ignoring unobserved characteristics yields **systematic bias**.

Empirical Applications

- **Automobile industry** (Berry, Levinsohn, Pakes 1993, 1995): plausible demand elasticities and substitution patterns.
- **Computer industry** (Greenstein 1992): vertical differentiation with sensible estimates.
- The framework has become a **standard tool** in empirical IO.

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Contributions

- 1 Introduces **mean-utility inversion** with **IV** estimation.
- 2 Provides a solution to **price endogeneity** in differentiated demand.
- 3 Integrates demand with **supply-side FOC**.

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Discussion Questions

- Why can market shares be inverted to recover mean utilities?
- Why are rival product characteristics valid instruments?
- Why does the random coefficients logit require simulation?

Thank You