

Estimating the lock-in effects of switching costs from firm-level data

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December 9, 2009. GVH

Introduction

- The presence of switching costs is likely to
 - ① give a profitable possibility for ex-post price increases by existing firms
 - ② increase barriers to entry and expansion of new firms
- Identifying&quantifying switching cost is important but not easy
 - ① Conceptual problems (dynamic choice problem)
 - ② Data requirements (ideal consumer-level data are rare)
- Our contributions
 - ① A simple intuitive method for estimating the **lock-in effects** of switching costs
 - ② Using **firm-level data** that might be requested by a competition or regulatory authority
 - ③ Application: estimating the lock-in effects of switching costs on the Hungarian personal loan market

The intuition of the method

- Compare the price responsiveness of **new consumers** and **old consumers**
- New consumers represent the behavior of old consumers if there are no switching costs
 - ▶ a counterfactual logic
- The difference in the price responsiveness is a measure of the lock-in effects
- In essence, it measures the effect of switching costs on the residual demand
 - ▶ closely connected to market power
- Since the behavior of new and old consumers is not directly observed from firm-level data, we use proxy variables
 - ▶ we derive the bias due to using proxy variables and correct the estimates for it
- We estimate that on the Hungarian personal loan market old consumers' price responsiveness is 70% lower because of switching costs

Previous works with aggregate data

- Structural form approaches
 - ① Shy (2002): static equilibrium model of switching costs, leading to stable market shares and different prices
Estimates for Finnish bank deposit market: switching costs are between 0 to 11% of average balance
 - ② Kim et al (2003): dynamic equilibrium model of consumer transitions and firms' intertemporal pricing
Estimates for Norwegian loan market: switching costs are around 4% of average loan
- Reduced form approaches
 - ① NERA (2003) idea: if with homogenous goods you estimate small cross-price elasticity, then it can be due to switching costs
It is not the magnitude of switching costs that is estimated, but whether their presence has a visible impact on consumers' decisions
- All of these papers use prices and one firm-level aggregate at most (sales-per-period or overall market share)

Assumptions on consumer choice

- Firm j , time t
- Two consumers: N and O , ex ante identical, 0-1 demand
 - 1 Consumer N is new, probability of choosing j at time t is n_{jt}
 - 2 Consumer O is old (choice in $t - 1$ was j), probability of choosing j at time t (the probability of staying loyal to j) is l_{jt}
- Now suppose a price increase for j
 - 1 n_j decreases: $\partial n_{jt} / \partial p_{jt} < 0$
 - 2 If no switching costs, l_j would decrease the same way
 - 3 If switching costs have lock-in effects, $\partial n_{jt} / \partial p_{jt} < \partial l_{jt} / \partial p_{jt} \leq 0$

Two measures to capture the lock-in effects of switching costs

- *1st measure for switching costs:*

$$\delta = \partial l_{jt} / \partial p_{jt} - \partial n_{jt} / \partial p_{jt} = |\partial n_{jt} / \partial p_{jt}| - |\partial l_{jt} / \partial p_{jt}|$$

How much more likely to turn away from i if new than if old?
Interpretation with heterogenous consumers: what fraction of consumers remain locked-in who would have switched otherwise?

- *2nd measure for switching costs:*

$$\theta = \frac{\partial n_{jt} / \partial p_{jt} - \partial l_{jt} / \partial p_{jt}}{\partial n_{jt} / \partial p_{jt}} = \frac{|\partial n_{jt} / \partial p_{jt}| - |\partial l_{jt} / \partial p_{jt}|}{|\partial n_{jt} / \partial p_{jt}|}$$

How smaller is old consumers' responsiveness to price changes than new ones'? - Better to compare different consumer groups or markets

Measurement

- Let us have a panel of J firms and T time periods
- Evaluation of consumers' stock for firm j in t :

$$S_{jt} = S_{jt-1} + \underbrace{IN_{jt}}_{\text{incoming}} - \underbrace{OUT_{jt}}_{\text{terminating}} - \underbrace{X_{jt}}_{\text{expiring}} =$$

$$S_{jt-1} + \left[\underbrace{N_{jt}}_{\text{new}} + \underbrace{F_{jt}}_{\text{from others}} \right] - \left[\underbrace{Q_{jt}}_{\text{quitters}} + \underbrace{T_{jt}}_{\text{to others}} \right] - X_{jt}$$

- Realized probability of choosing j in t if new:

$$n_{jt} = N_{jt} / \sum_i N_{jt}$$

- Realized probability of staying loyal to j in t if old:

$$l_{jt} = 1 - \frac{T_{jt}}{S_{jt-1} - Q_{jt} - X_{jt}}$$

- Data problem: ideally we want to measure N_{jt} and T_{jt} , but we usually have data only on S_{jt} , IN_{jt} , OUT_{jt} and X_{jt}

Ideal estimation

- Our goal is to estimate $\partial n_{jt}/\partial p_{jt}$ and $\partial l_{jt}/\partial p_{jt}$

$$\Delta n_{jt} = \alpha_n + \beta_n \Delta p_{jt-1} + u_{njt}$$

$$\Delta l_{jt} = \alpha_l + \beta_l \Delta p_{jt-1} + u_{ljt}$$

- The reason we use lagged prices:
 - 1 transactions follow after some time of price changes
 - 2 might take care of endogeneity (can be controlled more by adding Δp_{jt})
- OLS estimators for the lock-in measures of interest

$$\hat{\delta} = \hat{\beta}_l - \hat{\beta}_n \text{ and } \hat{\theta} = \frac{\hat{\beta}_n - \hat{\beta}_l}{\hat{\beta}_n} \text{ if}$$

- these are consistent if
 - 1 new and old consumers would have the same reaction if all were new (e.g. their characteristics would be the same on average)
 - 2 price changes are exogenous to demand
- Use of cross-section and time fixed effect can control for firm-specific trends and common shocks

Applied estimation

- Estimate the previous system with proxies

$$\Delta m_{jt} = \alpha_m + \beta_m \Delta p_{jt-1} + u_{mjt}, \text{ where } m_{jt} = \frac{IN_{jt}}{\sum_i IN_{jt}}$$

$$\Delta k_{jt} = \alpha_k + \beta_k \Delta p_{jt-1} + u_{kjt}, \text{ where } k_{jt} = 1 - \frac{OUT_{jt}}{S_{jt-1} - X_{jt}}$$

- Additional sufficient condition for $\hat{\beta}_m$ and $\hat{\beta}_k$ to be consistent if

$$Cov(\Delta m_{jt} - \Delta n_{jt}, \Delta p_{jt-1}) = 0 \text{ and}$$

$$Cov(\Delta k_{jt} - \Delta l_{jt}, \Delta p_{jt-1}) = 0$$

Applied estimation, cont.

- $Cov(\Delta k_{jt} - \Delta l_{jt}, \Delta p_{jt-1}) = 0$ is satisfied approximately:

$$l_{jt} = 1 - \frac{T_{jt}}{S_{jt-1} - Q_{jt} - X_{jt}}$$
$$k_{jt} = 1 - \frac{OUT_{jt}}{S_{jt-1} - X_{jt}} = 1 - \frac{T_{jt} + Q_{jt}}{S_{jt-1} - Q_{jt} - X_{jt} + Q_{jt}}$$

- so that $\beta_k \approx \beta_l$

Applied estimation, cont.

- $Cov(\Delta m_{jt} - \Delta n_{jt}, \Delta p_{jt-1}) = 0$ is not satisfied

$$n_{jt} = \frac{N_{jt}}{\sum_j N_{jt}}$$
$$m_{jt} = \frac{IN_{jt}}{\sum_j IN_{jt}} = \frac{N_{jt} + F_{jt}}{\sum_j (N_{jt} + F_{jt})}$$

- ▶ IN_{jt} may include switchers (F_{jt})
 - ▶ an increase in p_j might discourage switchers to j so that $Cov(p_j, F_j) < 0$
 - ▶ as a result, β_m can show a stronger (more negative) reaction than the true β_n
- However, we can derive an upper bound for this bias $a\beta_l \approx a\beta_k$
 - So the lower bound for bias-corrected estimations are

$$\hat{\delta}_{corr} = \hat{\beta}_k - \hat{\beta}_m + a\hat{\beta}_k \text{ and } \hat{\theta}_{corr} = \frac{\hat{\beta}_m - a\hat{\beta}_k - \hat{\beta}_k}{\hat{\beta}_m - a\hat{\beta}_k}$$

Our application: market of personal loans

- Market shares (stocks over all consumers)

Loan type	2002	2003	2004	2005	2006
Home currency, unsecured	44	56	53	39	28
Foreign currency, unsecured	0	0	4	10	10
Home currency, secured	56	44	17	6	4
Foreign currency, secured	0	0	26	44	58

- Concentrate on home unsecured segment: smaller changes, most "mature" segment
- Our database
 - 1 10 banks having at least 1% market share each
 - 2 quarterly data for 5 years (monthly data are very noisy)
 - 3 S_{jt} , IN_{jt} , OUT_{jt} , X_{jt} for both number and value of contracts
 - 4 prices on the modal product: APR already including entry costs

Estimation results

	in consumer number	in loan value
Response of new consumers $\hat{\beta}_m$ (confidence interval)	-0.61 (-0.93, -0.14)	-0.74 (-0.99, -0.22)
Response of old consumers $\hat{\beta}_k$ (confidence interval)	-0.13 (-0.18, -0.01)	-0.18 (-0.24, -0.00)
Switching costs: $\hat{\delta}$ upper bound (confidence interval)	0.48 (0.13, 0.87)	0.56 (0.22, 0.81)
Switching costs: $\hat{\theta}$ upper bound (confidence interval)	0.79 (0.66, 1.00)	0.76 (0.68, 1.00)
Switching costs: $\hat{\delta}_{corr}$ lower bound (confidence interval)	0.33 (0.03, 0.80)	0.31 (0.10, 0.61)
Switching costs: $\hat{\theta}_{corr}$ lower bound (confidence interval)	0.70 (0.35, 1.00)	0.63 (0.41, 1.00)

- Estimated value of the proportional correction factor is $a = 1.4$
- Block-bootstrap confidence intervals (5th & 95th percentile) with 2000 runs

Conclusion

- Developed a simple method to identify the lock-in effects of switching costs
 - ▶ using prices and two firm-level aggregates
 - ▶ correcting for bias in not measuring exactly what we want
- Estimated the model on personal loans in Hungary
 - ▶ old consumers' responsiveness is 70% lower because of switching costs
 - ▶ implying significant lock-in effects

THANK YOU FOR YOUR ATTENTION