

Estimation of Demand for Mortgage Loans Using Loan-level Data

Perm Winter School 2013

Ozhegov, Evgeniy M.

National Research University Higher School of Economics Center for Empirical Analysis of Industrial Organization

Perm Winter School, Perm, 05.02.2013



Outline

- Demand for loans
- Background
- **D** Econometric model
- Discussion





Demand for Loans

- Probability of credit contract agreement;
- □ Credit contract terms when agreed:
 - □ Loan amount (or Loan-to-Value);
 - □ Interest rate;
 - □ Maturity;
 - Downpayment.
- All are functions of:
 - □ Socio-demographic characteristics of borrower;
 - □ Terms of credit program (contract terms limits);
 - Desired property;
 - □ Macrovariables;
 - **Expected** loan performance:
 - □ Probability of default and loss given default for borrower;
 - Probability of prepayment and loss given prepayment for borrower.



Background

Mortgage borrowing as a simultaneous decision-making process of borrower and credit organization generates **endogeneity**.

Follain (1990):

- 1. How much to borrow;
- 2. If and when to refinance;
- 3. Choice of mortgage instrument.

Rachlis, Yezer (1993):

- 1. Borrower's application;
- 2. Borrowers's selection of mortgage terms;
- 3. Lender's endorsement;
- 4. Borrower's default.



Background

Sequential process generates sample selection.

Plillips, Yezer (1996), Ross (2000):

- 1. Borrower's application and selection of mortgage terms;
- 2. Lender's endorsement;
- 3. Borrower's default.



1. Instumental variables for endogenous demographics:

$$D^{en} = Z_D \beta_D + e_D$$

 D^{en} – vector of endogenous socio-demographic characteristics, Z_D – Instrumental variables for demographics.

2. Modeling the probability of application:

$$y_{1} = \begin{cases} 1, if \ D\beta_{D}^{1} + M\beta_{M}^{1} + e_{1} \ge \alpha_{1} \\ 0, if \ D\beta_{D}^{1} + M\beta_{M}^{1} + e_{1} < \alpha_{1} \end{cases}$$

 $y_1 = 1$ is an application decision,

 $D = (D^{ex}, \widehat{D}^{en})$ – vector of exogenous demographics and fitted endogenous demographics,

M – macrovariables.



3. Modeling the probability of endorsement for all applied:

$$y_{2} = \begin{cases} 1, if \ D\beta_{D}^{2} + M\beta_{M}^{2} + E[e_{2}|y_{1} = 1] \geq \alpha_{2} \\ 0, if \ D\beta_{D}^{2} + M\beta_{M}^{2} + E[e_{2}|y_{1} = 1] < \alpha_{2} \end{cases}$$

 $y_1 = 1$ is an endorsement decision.

4. Choice of loan amount limit for all endorsed:

$$\bar{L} = D\beta_D^{\bar{L}} + M\beta_M^{\bar{L}} + E[e_{\bar{L}}|y_2 = 1]$$

 \overline{L} is a decision on loan limit.



5. Modeling the probability of contract agreement:

$$y_{3} = \begin{cases} 1, if \ D\beta_{D}^{3} + M\beta_{M}^{3} + \hat{L}\beta_{\bar{L}}^{3} + E[e_{3}|y_{2} = 1] \ge \alpha_{3} \\ 0, if \ D\beta_{D}^{3} + M\beta_{M}^{3} + \hat{L}\beta_{\bar{L}}^{3} + E[e_{3}|y_{2} = 1] < \alpha_{3} \end{cases}$$

 $y_3 = 1$ is an agreement decision; \hat{L} is a fitted value of loan amount limit.

6. Choice of credit terms and property:

$$\begin{cases} C_{i} = D\beta_{D}^{C_{i}} + M\beta_{M}^{C_{i}} + C_{-i}\beta_{C_{-i}}^{C_{i}} + V\beta_{V}^{C_{i}} + F\beta_{F}^{C_{i}} + E[e_{C_{i}}|y_{3} = 1, C \in \bar{C}] \\ V = D\beta_{D}^{V} + M\beta_{M}^{V} + C\beta_{C}^{V} + F\beta_{F}^{V} + E[e_{V}|y_{3} = 1, C \in \bar{C}] \end{cases}$$

 $C = (C_i, C_{-i})$ is a vector of contract terms (Loan amount, Maturity, Downpayment, Interest rate),

 \overline{C} is a feasible set of contract terms determined by credit program,

V is a property value,

F is a property characteristics.



7. Modeling the probability of contract events:

$$y_4 = k, if \ D\beta_D^4 + M\beta_M^4 + \hat{C}\beta_C^4 + \hat{V}\beta_V^4 + A\beta_A^4 + E[e_4|y_3 = 1] \in \mathcal{L}_k$$

 $y_4 = k$ is a fact of *k*-th credit event, \hat{C} and \hat{V} are fitted values of credit terms and property value, *A* is a loss given *k*-th event.



Approaches to Sample Selection Correction

Sequential decisions

$$y_{1} = \begin{cases} 1, if \ X\beta_{X}^{1} + e_{1} \ge \alpha_{1} \\ 0, if \ X\beta_{X}^{1} + e_{1} < \alpha_{1} \end{cases}$$

$$y_2 = Z\beta_Z^2 + E[e_2|y_1 = 1]$$

Since
$$y_1 = g(X\beta_X^1, F(e_1))$$
 then
 $E[e_2|y_1 = 1] = \psi(X\beta_X^1, F(e_1), F(e_2))$

Since e_1 and e_2 are independent then

$$y_2 = Z\beta_Z^2 + \psi\left(X\beta_X^1, F(e_1)\right) + e_2$$



Approaches to Sample Selection Correction

Depends on $F(e_1)$:

If $e_1 \sim N(0, \sigma_1)$ then standart Heckman:

 $\Box \ y_2 = Z\beta_Z^2 + \rho * IMR\left(\widehat{X\beta_X^1}\right) + e_2$

If $F(e_1)$ is undefined then use semi-parametrics:

 $\Box \psi$ is a function on $\widehat{X\beta}_X^1$ (polynomial or etc.);

□ difference-in-difference.





Discussion

- 1. Full rationality of agents causes simultaneity in all decisions.
- 2. Credit history is unobservable.



Thank you for attention

E-mail: tos600@gmail.com