

Does it Pay to Get a Reverse Mortgage?

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Motivation

Empirical Evidences:

- The Baby Boomer retirement started in 2001.
- By 2030, one out of five people is projected to be 65 years or older.
- More than 80 percent of older households own their homes, which are worth approximately \$4 trillion.
- In the 1990s, reverse mortgages became available and provided a new way to convert home equity into cash.

Reverse Mortgage

- Reverse mortgages let homeowners
 - a) remain in their homes for as long as possible
 - b) borrow against their home equity at terms that include large up-front costs and high interest rates.
- There are no monthly or other payments to be made during the term of the loan.
- Repayment of principal plus cumulated interests is triggered by moving, is repaid out of house sale proceeds, and is capped by the value of those proceeds (nonrecourse loan).
- To be eligible:
 - a) a borrower must be 62 or older
 - b) own the home outright or have a low loan balance
 - c) have no other liens against the home
 - d) no credit or income requirements

Example of Loan Calculation

Your home is in San Francisco County. Your attained age is 62.

| | |
|-----------------------------|-----------|
| Value of your home | \$100,000 |
| Loan principal limit | \$49,200 |
| Less loan fees to lender | \$2,000 |
| Less Mortgage Insurance | \$2,000 |
| Less other closing costs | \$1,914 |
| Less service fee set-aside | \$4,631 |
| Cash available to you | \$38,655 |
| Current interest rate index | 2.8% |
| Lender's margin | 3.10% |
| Current loan interest rate | 5.18% |
| HUD Mortgage Insurance | 0.50% |
| Current effective loan rate | 5.68% |

Source reversemortgage.com (October 2008)

The Reverse Mortgage Market

- The reverse mortgage market was created in 1987 with the HUD (Department of Housing and Urban Development) program called Home Equity Conversion Mortgage (HECM)
- Reverse mortgages are specifically designed for house-rich but cash-poor homeowners. However, these homeowners have not bought them.
- Potential Market: 13.2 million older households (Stucki 2005)
- Actual Market: 265,234 federally insured reverse mortgages in 2007, about 1% of the 30.8 million households with at least one member age 62 and older in 2006

My Study

- Questions:
 - 1- Does it pay to get a reverse mortgage?
 - 2- Why do house-rich but cash-poor homeowners choose not to cash in the home equity through a reverse mortgage but prefer to maintain low level of consumption?
- This study estimates a dynamic structural life-cycle model of retiree consumption, housing and moving decisions. These decisions are made in light of lifespan and mobility uncertainty.
- Data: Subsample of single retirees from the Health and Retirement Study that could qualify for a reverse mortgage.

My Contribution

- 1** I obtain reasonable estimates for the structural preference parameters.
- 2** I provide a plausible explanation for the relative weakness in reverse mortgage demand.
Reverse mortgages:
 - A bad option for house-rich but cash-poor households
 - PROS: Provide liquidity and a form of longevity insurance
 - CONS: Large up-front costs and the moving risk
- 3** I use a set of tools from numerical analysis to estimate and solve the empirical model.

- 1 Outline
- 2 Reverse Mortgage
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Empirical Evidence about Reverse Mortgagees

- Between 1993 and 2004, the median annual income of reverse mortgage borrowers increased from \$12,289 to \$18,240.
- The median net worth among the general population of older households, excluding home equity, was \$23,369 in 2000.
- Close to half of reverse mortgage borrowers (46 percent) have homes worth \$100,000 to \$200,000, compared with only about one-third of general homeowners (34 percent)
- Davidoff et al. (2007) shows that, empirically, reverse mortgagees have exited homes unusually rapidly.
- About 60% of loan terminations are attributed to death and about 40% to moving out.

Literature Review

- Life-cycle and Precautionary Savings:
Gourinchas and Parker (2002), Cagetti (2003),
French (2005), Hubbard et al. (1994),
Palumbo (1999), Hurd (1989)
- Housing and Portfolio Choice:
Cocco (2005), Yao and Zhang (2005)
- Discrete Choice:
Rust (1987), Hotz and Miller (1993),
Aguirragabiria and Mira (2002), Keane and Wolpin (1997)
- Numerical Analysis:
Judd and Su (2007)

Model

- Preferences

$$U_{it}(C_{it}, H_{it}) = \frac{(C_{it}^{1-\omega} (\psi^{rent} H_{it})^{\omega})^{1-\gamma}}{1-\gamma} + \varepsilon_{it}(d_{it})$$

where $\varepsilon_{it}(d_{it})$ is a vector of unobserved utility components associated to the discrete housing choice and it is Extreme Value Type I distributed

- Budget Constraint

$$A_{it+1} = (1+r)A_{it} + y - C_{it} - \psi_{it} - M_{it}$$

- Bequest Function:

$$b(TW_{it}) = \theta_B \frac{TW_{it}^{1-\gamma}}{1-\gamma}$$

Choice Set

- Discrete Choice: Housing

$$d_{it}^1 = \begin{cases} D_{it}^M = 1 & \text{if household } i \text{ moves out of her house in period } t \\ D_{it}^M = 0 & \text{otherwise} \end{cases}$$

$$d_{it}^2 | d_{it}^1 = \begin{cases} D_{it}^O = 1 & \text{if household } i \text{ owns her house in period } t \\ D_{it}^O = 0 & \text{if household } i \text{ rents her house in period } t \end{cases}$$

$$d_{it}^3 | d_{it}^1, d_{it}^2 = H_{it}$$

The discrete choice set is

$$d_{it} = \{d_{it}^1, d_{it}^2, d_{it}^3\}$$

- Continuous Choice: Consumption C_{it}

Housing Expenses

■ Per-Period Cost

$$\psi_{it} = [D_i^O \psi^{own} + (1 - D_i^O) \psi^{rent}] H_{it}^*$$

where

$$H_{it}^* = D_{it}^M H_{it} + (1 - D_{it}^M) H_{it-1}$$

■ Moving Cost

$$M_{it} = D_{it}^M D_{it-1}^O [D_{it}^O H_{it} - H_{it-1} + H_{it} \phi(D_{it}^O)] + D_{it}^M (1 - D_{it-1}^O) (1 - D_{it}^O) H_{it} \phi^{rent}$$

where the transaction costs are:

$$\phi(D_{it}^O) = [D_{it}^O \phi^{own} + (1 - D_{it}^O) \phi^{rent}]$$

Value Function

$$V_t(X_{it}, \varepsilon_{it}) = \max_{d_{it}, C_{it}} U(d_{it}, C_{it}) + \varepsilon_{it}(d_{it}) \\ + \beta \eta_{it+1} E[V_{it+1}(W_{it+1}, H_{it}^*, D_{it}^O, \varepsilon_{it+1} | X_{it}, C_{it})] + b(TW_{it+1})$$

subject to

$$W_{it+1} = (1 + r)W_{it} + y - C_{it} - \psi_{it} - M_{it}$$

$$H_{it}^* = D_{it}^M H_{it} + (1 - D_{it}^M) H_{it-1}$$

$$C_{it} \geq C_{MIN}$$

State Space: $X_{it} = \{W_{it}, H_{it-1}, D_{it-1}^O, Age_{it}\}$

Preference parameters to estimate: $\theta = \{\gamma, \omega, \sigma, \theta_B\}$



- Bellman Equation:

$$\begin{aligned}
 V_t(X_{it}, \varepsilon_{it}) &= \max_{d_{it}, C_{it}} [U(d_{it}, C_{it}) + \varepsilon_{it}(d_{it}) + \beta\eta_{it+1}EV(X_{it+1})] \\
 &= \max_{d_{it}} \left\{ \left[\max_{C_{it}} \{ [U(d_{it}, C_{it}) + \beta\eta_{it+1}EV(X_{it+1})] | d_{it} \} \right] + \varepsilon_{it}(d_{it}) \right\}
 \end{aligned}$$

- Inner Maximization (consumption conditional on housing)

$$r(X_{it}, d_{it}) = \max_{C_{it}} [U(d_{it}, C_{it}) + \beta\eta_{it+1}EV_{t+1}(X_{it+1}) | d_{it}]$$

- Outer Maximization (housing)
Conditional Choice Probabilities

$$P(j|X_{it}, \theta) = \frac{\exp\{r(X_{it}, j)\}}{\sum_{k \in d_{it}(X_{it})} \exp\{r(X_{it}, k)\}}$$

where

$$EV_{t+1}(X_{it+1}) = \ln \left[\sum_{k \in d_t(X_t)} \exp\{r(X_{it}, k)\} \right]$$

Loglikelihood

- Measurement Error in Consumption

$$\Pr(c_{n,t} | d_{n,tp}^H, A_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(c_{n,tp}^{data} - c_{n,tp}^{pred})^2}{2\sigma^2}}$$

- Discrete Choice Probability

$$\Pr(d_{n,tp}^H | A_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) = \frac{e^{V_{d,n,tp}}}{\sum_m e^{V_{m,n,tp}}}$$

- Joint Probability of Housing and Consumption Choice

$$\Pr(d_{n,tp}^H, c_{n,tp} | A_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) = \Pr(d_{n,tp}^H | A_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) \cdot \Pr(c_{n,t} | d_{n,tp}^H, A_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data})$$

- Log-Likelihood

$$\mathcal{L}(\theta) = \sum_{n=1}^N \sum_{tp=1}^{TP} \log \Pr(d_{n,tp}^H, c_{n,tp} | A_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}, \theta)$$

Four Tools from Applied Mathematics

Basic Idea:

Use methods and software developed by computational scientists and mathematicians to reduce the computational burdens of structural estimation and to significantly increase the precision in the economic results.

- Mathematical Programming with Equilibrium Constraints
- Flexible Polynomial Approximation
- Shape Preservation
- Envelope Theorem

Simple Life-Cycle Model: One continuous state variable

■ Backward Solution for the True Value Function

The last period value function is known and equal to $V_T(W)$

In periods $t = 1 \dots (T - 1)$ the Bellman equation is:

$$V_t(W) = \max_c (u(c) + \beta EV_{t+1}(RW - c))$$

Given V_{t+1} , the Bellman equation implies, for each wealth level W , three equations that determine optimal consumption, c^* , $V_t(W)$, and $V'_t(W)$:

- Bellman equation: $V_t(A) = u(c^*) + \beta V_{t+1}(RW - c^*)$
- Euler equation: $u'(c^*) - \beta V'_{t+1}(RW - c^*) = 0$
- Envelope condition: $V'(W) = \beta R V'(RW - c^*)$

Backward Solution for the Approximate Value Function

- Choose a functional form and a finite grid of wealth levels

Time t value function is approximated by

$$V_t(W) = \Phi(W; a_t) = \sum_{k=0}^7 a_{k+1,t} (W - \bar{W}_t)^k$$

- We would like to find coefficients a_t such that each time t Bellman equation, along with the Euler equation and envelope condition, holds with the Φ approximation

$$\Phi(W; a_t) = \max_c (u(c) + \beta \Phi_{t+1}(RW - c; a_{t+1}))$$

- Define three set of errors, $\lambda_t^b \geq 0$, $\lambda_{i,t}^e \geq 0$, $\lambda_t^{env} \geq 0$, that satisfy the following inequalities

- Bellman error

$$-\lambda_t^b \leq \Phi(W_{i,t}; a_t) - [u(c_{i,t}^*) + \beta\Phi_{t+1}(RW_{i,t} - c_{i,t}^*; a_{t+1})] \leq \lambda_t^b$$

- Euler error

$$-\lambda_{i,t}^e \leq u'(c_{i,t}^*) - \beta\Phi'(RW_{i,t} - c_{i,t}^*; a_{t+1}) \leq \lambda_{i,t}^e$$

- Envelope error

$$-\lambda_t^{env} \leq \Phi'(W_{i,t}; a_t) - R\beta\Phi'_{t+1}(RW_{i,t} - c_{i,t}^*; a_{t+1}) \leq \lambda_t^{env}$$

Dynamic Programming with Approximation of the Value Function

- Minimize the sum of the errors:

$$\min_{a,c,\lambda} \sum_t \sum_i \lambda_{i,t}^e + \sum_t \lambda_t^b + \sum_t \lambda_t^{env}$$

subject to:

- Bellman error
- Euler error
- Envelope error
- Transversality condition

where the Transversality condition:

$$\Phi(W_{i,t}; a_{i,t}) \geq \Phi(W_{i-1,t}; a_t) + \frac{(\Phi(W_{i+1,t}; a_t) - \Phi(W_{i-1,t}; a_t))}{(W_{i+1,t} - W_{i-1,t})} (W_{i,t} - W_{i-1,t})$$

Empirical Part

- We have continuous data on assets and consumption.
- We assume that the measurement error in consumption is normally distributed with mean 0 and unknown variance σ^2 .
- We can use the Euler equation to recover the predicted value of consumption.
- The probability that household n chooses consumption $c_{n,tp}$ in period tp is:

$$\Pr(c_{n,tp} | W_{n,tp}^{data}) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(c_{n,tp}^{data} - c_{n,tp}^{pred})^2}{2\sigma^2}}$$

Therefore the log-likelihood is given by:

$$\mathcal{L}(\theta) = \sum_{n=1}^N \sum_{tp=1}^{TP} \log \Pr(c_{n,tp} | A_{n,tp}^{data}, \theta)$$

Structural Estimation with Dynamic Programming

■ Conventional Approach

- 1 Take a guess of the structural parameters
- 2 Solve the dynamic programming problem
- 3 Calculate the log-likelihood
- 4 Repeat 1,2,3 until the log-likelihood is maximized

■ Constrained Optimization (MPEC) Approach

$$\max_{a,c,\lambda,\theta} \mathcal{L}(\theta) - \text{Penalty} \cdot \Lambda$$

- Bellman error
- Euler error
- Envelope error
- Transversality condition

where $\Lambda = \sum_t \sum_i \lambda_{i,t}^e + \sum_t \lambda_t^b + \sum_t \lambda_t^{env}$

▶ Go to the Appendix



Dynamic Programming

■ Minimize the Sum of Errors

$$\Lambda = \sum_t \sum_i \sum_j \sum_z \lambda_{i,j,z,t}^e + \sum_t \sum_j \sum_z \lambda_{j,z,t}^b + \sum_t \sum_j \sum_z \lambda_{j,z,t}^{env} + \sum_t \sum_i \sum_j \sum_z \sum_d \lambda_{i,j,d,z,t}^{cons}$$

subject to:

$$\begin{aligned} -\lambda_{i,j,z,t}^e &\leq EulerEquation_{i,j,z,t} \leq \lambda_{i,j,z,t}^e \\ -\lambda_{j,z,t}^b &\leq BellmanEquation_{j,z,t} \leq \lambda_{j,z,t}^b \\ -\lambda_{j,z,t}^{env} &\leq EnvelopeCondition_{j,z,t} \leq \lambda_{j,z,t}^{env} \\ -\lambda_{i,j,d,z,t}^{cons} &\leq PolicyFunction_{i,j,d,z,t} \leq \lambda_{i,j,d,z,t}^{cons} \end{aligned}$$

Solving DP and Estimation with the MPEC

$$\underset{\theta, a, c}{\text{Max}} \mathcal{L}(\theta) - \text{Penalty} \cdot \Lambda$$

subject to:

Euler error

Bellman error

Envelope error

Policy function error

Transversality condition

where

$$\mathcal{L}(\theta) = \sum_{n=1}^N \sum_{tp=1}^{TP} \log \Pr(d_{n,tp}^H, c_{n,tp} | A_{n,tp}^{data}, H_{n,tp}^{data}, \theta)$$

Data

The Health and Retirement Study (HRS) and The Consumption and Activities Mail Survey (CAMS). US data (2000-2005).

- We select a group of 165 households that qualify for a reverse mortgage.
- Characteristics: 62 years old or older, single, retiree, homeowner, complete information about consumption and financial situation.

| | Percentiles | | | Min | Max |
|------------|-------------|----------|----------|----------|-----------|
| | 25% | 50% | 75% | | |
| <i>H</i> | \$40,000 | \$70,000 | \$90,000 | \$ 3,000 | \$170,000 |
| <i>W</i> | \$6,000 | \$25,000 | \$69,500 | \$0 | \$276,548 |
| <i>C</i> | \$6,347 | \$9,774 | \$15,409 | \$650 | \$84,380 |
| <i>ss</i> | \$7,200 | \$9,600 | \$11,748 | \$0 | \$ 18,907 |
| <i>Age</i> | 69 | 75 | 80 | 66 | 86 |

Financial Portfolio Composition

| | Percentiles | | | Min | Max |
|---------------|-------------|---------|----------|------|-----------|
| | 25% | 50% | 75% | | |
| <i>Stocks</i> | \$0 | \$0 | \$0 | \$ 0 | \$125,000 |
| <i>Chck</i> | \$750 | \$3,600 | \$10,000 | \$ 0 | \$100,000 |
| <i>Cds</i> | \$0 | \$0 | \$5,300 | \$ 0 | \$273,548 |
| <i>Tran</i> | \$1,000 | \$4,000 | \$8,000 | \$ 0 | \$30,000 |
| <i>Bonds</i> | \$0 | \$0 | \$0 | \$ 0 | \$80,000 |
| <i>IRA</i> | \$0 | \$0 | \$2,5000 | \$ 0 | \$137,000 |
| <i>Debt</i> | \$0 | \$0 | \$0 | \$ 0 | \$12,000 |

- For almost all the retirees in the sample, the financial portfolio does not contain risky assets.

Moving Decision

- In each period, about 8% of the households in our sample moves out of their homes.
- Among those who moved, about 20% decide to rent a new house, while about 80% buy a new house.
- The moving decision does not appear to be strictly related with age.
- About 50% of the retirees move near or with children or other relatives or friends. About 25% move for financial reasons, and the remaining 25% move because of health problems, weather or climate reasons, to have a better location, or for other reasons.

Results

| Parameter | Variable | Estimate |
|------------|------------------------------------------|----------------|
| γ | Coefficient of relative risk aversion | 1.4196 (0.013) |
| ω | Preference parameter over housing | 0.5325 (0.032) |
| σ | s.d. of measurement error in consumption | 1.206 (0.640) |
| θ_B | Degree of altruism | 0.000 (0.001) |

- The standard errors are computed using a bootstrap procedure.

Simulation of Welfare Gain from Reverse Mortgage

The welfare gain from a reverse mortgage is calculated as a percentage increase in the initial non-housing financial wealth that makes the household without reverse mortgage as well off in expected utility terms as with the reverse mortgage.

Simulation of Welfare Gain from Reverse Mortgage

| | HOUSE | |
|-------------------------|-------------------|-------------------|
| | House-Poor | House-Rich |
| FINANCIAL WEALTH | | |
| Cash-Poor | \$8,500 | \$ 10,600 |
| Cash-Rich | \$107,800 | \$ 90,200 |

where:

Financial Wealth

House Value

Cash-Poor: < \$40,000

House-Poor: <\$60,000

Cash-Rich: > \$40,000

House-Rich: >\$60,000

Simulation of Welfare Gain from Reverse Mortgage

Median Welfare Gain, Baseline Case

| | House-Poor | HOUSE House-Rich |
|-------------------------|----------------|---------------------|
| FINANCIAL WEALTH | | |
| Cash-Poor | -9%(- \$ 767) | -14%(- \$ 1,525) |
| Cash-Rich | 50%(\$ 53,302) | 41% (\$ 36,863) |

- PROS of RM: liquidity and a form of longevity insurance.
- CONS of RM: high up-front cost and moving risk.

No Moving Risk

Median Welfare Gain

| | House-Poor | HOUSE House-Rich |
|-------------------------|-----------------|---------------------|
| FINANCIAL WEALTH | | |
| Cash-Poor | 207%(\$ 17,600) | 430%(\$ 45,622) |
| Cash-Rich | 18%(\$ 19,054) | 54%(\$ 48,883) |

No Up-Front Costs

Median Welfare Gain

| | HOUSE | |
|-------------------------|----------------|----------------|
| | House-Poor | House-Rich |
| FINANCIAL WEALTH | | |
| Cash-Poor | 2%(\$ 146) | -12%(\$ 1,374) |
| Cash-Rich | 56%(\$ 60,021) | 49%(\$ 43,784) |

10% Cut in Current Income

Median Welfare Gain

| | House-Poor | HOUSE House-Rich |
|-------------------------|-----------------|---------------------|
| FINANCIAL WEALTH | | |
| Cash-Poor | -20% (\$ 1,657) | -16% (\$ 1,654) |
| Cash-Rich | 49% (\$ 53,298) | 46% (\$ 41,728) |

Conclusions

- Innovative structural dynamic life-cycle model of consumption, housing and mobility choice to calculate the welfare benefits of allowing retirees to cash in their home equity through a reverse mortgage.
- First application of a set of four mathematical tools to estimate and solve an empirical model.
- Reverse mortgages provide liquidity and a form of longevity insurance, but introduce a new risk, the moving risk. These financial instruments are risky especially for house-rich but cash-poor homeowners.
- Gambling can make someone who is initially poor relatively rich. However, luck plays an important role in this gamble.

Thank you

Continuous and Discrete State Variables

Let W be a continuous state variable and J be a discrete state variable.

Time t value function is approximated by

$$V_t(W, J) = \Phi(W, J; a_t) = \sum_{k=0}^7 a_{k+1,t} (W - \bar{W}_t)^k$$

The constrained optimization approach to a life-cycle model with continuous and discrete state variables is:

$$\text{Minimize } \sum_i \sum_j \sum_t \lambda_{i,j,t}^e + \sum_j \sum_t \lambda_{j,t}^b + \sum_j \sum_t \lambda_{j,t}^{env}$$

subject to

- Bellman Error:

$$-\lambda_{j,t}^b \leq \Phi(W, J; a_t) - [u(c^*, J) + \beta \Phi(RW - c^*, J; a_{t+1})] \leq \lambda_{j,t}^b$$

- Euler Error

$$-\lambda_{i,j,t}^e \leq u'(c^*, J) - \beta \Phi'(RW - c^*, J; a_{t+1}) \leq \lambda_{i,j,t}^{env}$$

- Envelope Error:

$$-\lambda_{j,t}^{env} \leq \Phi'(W, J; a_t) - R\beta \Phi'(RW - c^*, J; a_{t+1}) \leq \lambda_{j,t}^{env}$$

Appendix: DP with Approximation of the Value Function

- Euler Equations:

$$u'(c_{dN}^*, H) - \beta \eta_{t+1} R V'_{t+1}(RW - c_{dN}^* - \psi + y; H, Q) = 0$$

$$u'(c_{dMhq}^*, h) - \beta \eta_{t+1} R V'_{t+1}(RW - c_{dMhq}^* - \psi - M + y; h, q) = 0$$

- Bellman Equation:

$$V_t(W, H, Q) = \ln \left\{ \exp(\hat{V}_{dN,t}) + \sum_q \sum_h \exp(\hat{V}_{dMhq,t}) \right\}$$

- Envelope Condition:

$$V'_t(W, H, Q) = \Pr(NM|W, H, Q) \cdot \hat{V}'_{dN,t} + \sum_q \sum_h \Pr(Mhq|W, H, Q) \cdot \hat{V}'_{dMhq,t}$$

◀ Go back

- Value Function Approximation

$$V_t(W, H, Q) = \Phi(W, H, Q; a_t, \bar{W}_t) = \sum_{k=0}^7 a_{k+1, H, Q, t} (W - \bar{W}_t)^k$$

- Policy Function Approximation

$$c_{d,t}^*(W, H, Q) = \Phi(W, H, Q; b_{d,t}, \bar{W}_t) = \sum_{k=0}^7 b_{k+1, H, Q, d, t} (W - \bar{W}_t)^k$$

We would like to find coefficients a_t and $b_{d,t}$ such that each time t Bellman equation, along with the Euler and Envelope conditions, holds with the Φ approximation

[◀ Go back](#)

- Euler Errors

$$-\lambda_{i,j,z,t}^e \leq u'(c_{i,j,dN,t}^*, H_{j,t}) - \beta R \Phi'(RW_{i,t} - c_{i,j,dN,t}^* - \psi + y; H_{j,t}, Q_t; a_{t+1}) \leq \lambda_{i,j,z,t}^e$$

$$-\lambda_{i,j,z,t}^e \leq u'(c_{i,j,dMhq,t}^*, H_{t+1}) - \beta R \Phi'(RW_{i,t} - c_{i,j,dMhq,t}^* - \psi - M + y; H_{t+1}, Q_{t+1}; a_{t+1}) \leq \lambda_{i,j,z,t}^e$$

- Bellman Error

$$-\lambda_{j,z,t}^b \leq \Phi(W_{i,t}, H_{j,t}, Q_t; a_t) - \ln \left\{ \exp(\hat{V}_{i,j,dN,t}) + \sum_q \sum_h \exp(\hat{V}_{i,j,dMhq,t}) \right\} \leq \lambda_{j,z,t}^b$$

where

$$\hat{V}_{i,j,dN,t} = u(c_{i,j,dN,t}^*, H_{j,t}) + \beta \eta_{t+1} \Phi(RW_{i,t} - c_{i,j,dN,t}^* - \psi + y; H_{j,t}, Q_t; a_{t+1})$$

$$\hat{V}_{i,j,dMhq,t} = u(c_{i,j,dMhq,t}^*, H_{t+1}) + \beta \eta_{t+1} \Phi(RW - c_{i,j,dMhq,t}^* - \psi - M + y; H_{t+1}, Q_{t+1}; a_{t+1})$$

◀ Go back

- Envelope Error

$$\begin{aligned}
 -\lambda_{j,z,t}^{env} &\leq \Phi'(W_{i,t}, H_{j,t}, Q_t; a_t) - \{f_{i,j,dN,t} \cdot \Phi'(RW_{i,t} - c_{i,j,dN,t}^* - \psi + y; H_{j,t}, Q_t; a_{t+1}) \\
 &+ \sum_q \sum_h [f_{i,j,dMhq,t} \cdot \Phi'(RW_{i,t} - c_{i,j,dMhq,t} - \psi - M; H_{t+1}, Q_{t+1}; a_{t+1})]\} \leq \lambda_{j,z,t}^{env}
 \end{aligned}$$

where

$$f_{i,j,d,t} = \Pr(d|W_{i,t}, H_{j,t}, Q_t) = \frac{\exp(\widehat{V}_{i,j,d,t})}{\exp(\widehat{V}_{i,j,dN,t}) + \sum_q \sum_h \exp(\widehat{V}_{i,j,dMhq,t})}$$

- Policy Function Error

$$-\lambda_{i,j,z,d,t}^{cons} \leq \Phi(W_{i,t}, H_{j,t}, Q_t; b_t) - c_{i,j,d,t}^*(W_{i,t}, H_{j,t}, Q_t) \leq \lambda_{i,j,z,d,t}^{cons}$$

◀ Go back

Dynamic Programming

■ Minimize the Sum of Errors

$$\Lambda = \sum_t \sum_i \sum_j \sum_z \lambda_{i,j,z,t}^e + \sum_t \sum_j \sum_z \lambda_{j,z,t}^b + \sum_t \sum_j \sum_z \lambda_{j,z,t}^{env} + \sum_t \sum_i \sum_j \sum_z \sum_d \lambda_{i,j,d,z,t}^{cons}$$

subject to:

$$\begin{aligned} -\lambda_{i,j,z,t}^e &\leq EulerEquation_{i,j,z,t} \leq \lambda_{i,j,z,t}^e \\ -\lambda_{j,z,t}^b &\leq BellmanEquation_{j,z,t} \leq \lambda_{j,z,t}^b \\ -\lambda_{j,z,t}^{env} &\leq EnvelopeCondition_{j,z,t} \leq \lambda_{j,z,t}^{env} \\ -\lambda_{i,j,d,z,t}^{cons} &\leq PolicyFunction_{i,j,d,z,t} \leq \lambda_{i,j,d,z,t}^{cons} \end{aligned}$$

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Loglikelihood

- Measurement Error in Consumption

$$\Pr(c_{n,t} | d_{n,tp}^H, W_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(c_{n,tp}^{data} - c_{n,tp}^{pred})^2}{2\sigma^2}}$$

- Discrete Choice Probability

$$\Pr(d_{n,tp}^H | W_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) = \frac{e^{V_{d,n,tp}}}{\sum_m e^{V_{m,n,tp}}}$$

- Joint Probability of Housing and Consumption Choice

$$\Pr(d_{n,tp}^H, c_{n,tp} | W_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) = \Pr(d_{n,tp}^H | W_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}) \cdot \Pr(c_{n,t} | d_{n,tp}^H, W_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data})$$

- Log-Likelihood

$$\mathcal{L}(\theta) = \sum_{n=1}^N \sum_{tp=1}^{TP} \log \Pr(d_{n,tp}^H, c_{n,tp} | W_{n,tp}^{data}, H_{n,tp}^{data}, Q_{n,tp}^{data}, \theta)$$

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Solving DP and Estimation with the MPEC

$$\text{Max}_{\theta, a, c} \mathcal{L}(\theta) - \text{Penalty} \cdot \Lambda$$

subject to:

Euler error

Bellman error

Envelope error

Policy function error

where

$$\mathcal{L}(\theta) = \sum_{n=1}^N \sum_{tp=1}^{TP} \log \Pr(d_{n,tp}^H, c_{n,tp} | W_{n,tp}^{data}, H_{n,tp}^{data}, \theta)$$

We assume that there is a measurement error in consumption $\sim N(0, \sigma^2)$

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Empirical Evidence about Reverse Mortgagees

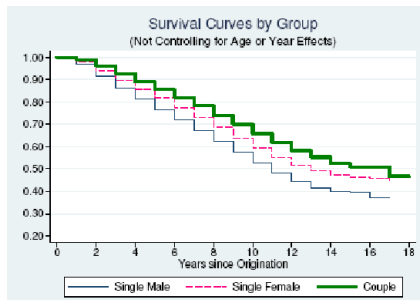


Figure: Survival Curves of HECM Loans for Single Males, Single Females, and Couples (Bowen et al.,2008)