

# Consumption-Savings with Retirement: Identification in Dynamic Discrete-Continuous Choice Models

---

M. Levy & P. Schiraldi

August, 2021

# Introduction

---

# Motivation

- Agents make complex decisions which involve continuous and discrete decisions: either because of the nature of the goods (notably durables); or because consumption decisions require adjustment or transaction costs (e.g. Chetty and Szeidl, 2007) and therefore are infrequently and discretely adjusted.
- Examples: agents choose
  - which retail stores to visit and how much to buy,
  - which car to buy and how much to drive
  - which investment products to buy and how much to invest,
  - which tasks to perform and how much effort to exert,

# Introduction

Two distinct approaches:

- Euler equation approaches assume existence of single composite good which decision-maker can freely and continuously adjust
- Discrete-choice models: Discount factor is typically not identified plus additional restrictive normalizations on utility (see for example Magnac and Thesmar (2002))
  - Few exceptions: see Levy & Schiraldi (2021).

# Introduction

We make three main theoretical contributions:

- Solve for the selection issue
- Establish new set of (pointwise) identification results of the discount factor and utility function
- Propose a simple two-step method to estimate (semi)parametrically this class of models

# Introduction

We then apply our theoretical results to a lifecycle savings model with retirement

- Use continuous consumption and binary retirement choices in Panel Study of Income Dynamics (PSID)
- Find mean annual discount factor of 0.91, with substantial heterogeneity: while the youngest households have a  $\delta$  approaching one, it falls to 0.863 by age 85.
- At the mean consumption level, we find a RRA coefficient of approximately 0.5 for retired households and 0.3 for working households.

# Model

---

# Setup

We assume an infinite time horizon ( $t = 1, 2, \dots$ ) and a stationary environment.

- Each period, agent chooses a discrete alternative  $j \in \mathcal{J}$  and continuous quantity  $q_j \in \mathbb{R}$
- State variables  $s_{it}, z_{it}, L_{it}, \zeta_{it}, \varepsilon_{it}$ .  $\bar{s}_{it} \equiv (s_{it}, z_{it}, L_{it})$ .
- $s_{it}, z_{it} \in \mathcal{S}$ , and  $L_{it} \in \mathbb{R}$  are observed by the researcher
  - $s_{it}$  is payoff-relevant state drawn from finite state space
  - $z_{it}$  instruments for the discrete choice
  - $L_{it} \in \mathbb{R}$  is current wealth
- $\varepsilon_{it} \equiv (\varepsilon_{i1t}, \dots, \varepsilon_{ijt}) \in \mathbb{R}^J$  represents the vector of individual idiosyncratic random preference shocks
- $\zeta_{it} \in \mathbb{R}$  is an individual level shock to the marginal



# Setup: Assumptions

- The instantaneous utilities are given by, for each  $j \in J$ ,

$$\bar{u}_{jt}(s_t, \zeta_t, q_{jt}, z_t, \varepsilon_{jt}) = u_{ijt}(s_t, z_t, \zeta_t, q_{jt}) + \varepsilon_{jt}$$

- $F(\varepsilon_t)$  and  $G(\zeta_t)$  are known
- Both shocks are observed by agents before making any decisions
- CI assumption similar to Rust

# Selection problem

Challenge in the identification: solve selection problem on unobservable — particularly  $\zeta_t$  — which may affect the conditional continuous choice.

## Assumption

*(Unlimited encouragement)* There exists  $k$  such that for all  $j \in \mathcal{J} \setminus \{k\}$ :

1. The marginal utility  $\partial u_j / \partial q$  is independent of  $z$  for all  $(q, \bar{s}, \zeta)$ .
2. For all  $(s, L)$  there exists a sequence  $\{z_{j,n}\}$  such that 
$$\lim_{n \rightarrow \infty} \Pr(d_t = j | s, L, z_{j,n}) = 1$$

# Unlimited encouragement

It may be viewed as a strengthening of the common encouragement design of experiments with non-random treatment assignment.

- For example, suppose a consumer is choosing between a local bodega in Manhattan and a Costco in New Jersey (which differ in the marginal utility of their goods)
- We require that traffic may be sufficiently bad that the consumer chooses the Manhattan store with arbitrarily high probability — but note that we do not require traffic conditions that force them to New Jersey (and moreover the marginal utility in New Jersey may always depend on traffic).

**Lemma 1:**  $\zeta$  and  $\{q_j^*(\bar{s}, z, \zeta)\}_{j \in \mathcal{J}}$  are identified

# Setup: Value functions

The agent's problem can be written recursively as:

$$V(\bar{s}_t, \zeta_t, \varepsilon_t) = \max_{j_t, q_{jt}} \{ u_j(q_{jt}, s_t, \zeta_t, z_t) + \varepsilon_{jt} + \\ + \delta E[V(\bar{s}_{t+1}, L_{t+1}, \zeta_{t+1}, \varepsilon_{t+1}) | j_t, q_{jt}, \bar{s}_t] \}$$

$$\text{s.t.} \quad L_{t+1} = f_j(s_{t+1})(L_t - q_{jt} - \phi_j(s_t, L_t))$$

# Setup: Value functions

The agent's problem can be written recursively as:

$$V(\bar{s}_t, \zeta_t, \varepsilon_t) = \max_{j_t, q_{jt}} \{ u_j(q_{jt}, s_t, \zeta_t, z_t) + \varepsilon_{jt} + \\ + \delta E[V(\bar{s}_{t+1}, L_{t+1}, \zeta_{t+1}, \varepsilon_{t+1}) | j_t, q_{jt}, \bar{s}_t] \}$$

$$\text{s.t. } L_{t+1} = f_j(s_{t+1})(L_t - q_{jt} - \phi_j(s_t, L_t))$$

with conditional value function:

$$v_j(\bar{s}_t, \zeta_t) \equiv u_j(q_{jt}^*, \bar{s}_t, \zeta_t) + \delta E [V(\bar{s}_{t+1}, \zeta_{t+1}) | j_t, \bar{s}_t, L_t, q_{jt}^*]$$

- Note: due to max operator,  $V$  not typically concave

## Setup: A modified Euler equation

Combining these results, we obtain a modified Euler equation for any alternative  $j$ :

$$\frac{\partial u_j(q_t^*, s_t, \zeta_{jt})}{\partial q} = \delta \mathbb{E} \left[ f_j(.) \left( \frac{\partial u_j(q_{t+1}^*, s_{t+1}, \zeta_{t+1})}{\partial q} + \frac{\partial \Phi_j(\bar{s}_{t+1}, \zeta_{t+1})}{\partial L_{t+1}} \right) | j_t, \bar{s}_t, q_{jt}^* \right]$$

- If no discrete choice, reduces to familiar Euler equation
- Presence of discrete choice causes two adjustments:
  - Continuation value must account for future marginal utility *and* marginal impact on future choice surplus

# Results: Identification

## Theorem (1)

$\delta(\cdot)$  and  $\partial u(\cdot)/\partial q$  are point-identified.



# Level of utility

## Assumption (Normalization of utility)

$u_0(0, s, \zeta) = 0$  for all  $\bar{s}$  and  $\zeta$

- some degree of normalization is required given that only differences in utilities affect the decision-maker's choices.
- We impose that the utility of consuming  $q_0 = 0$  for the reference alternative is normalized to zero across states.

## Theorem (4)

*Under assumptions above then  $(u(\cdot), \delta(\cdot))$  is point identified.*

# Consumption, Savings, and Retirement

---

# Consumption, Savings, and Retirement

The canonical lifecycle consumption model is filled with discrete choices

- We focus on (binary) choice of retirement — i.e. exiting labor force
  - Highly implausible that DM obtains same utility when working and retired

# Data

- Primary data source is PSID, 1999-2017.
  - Longitudinal survey of households with bi-annual information about income, employment, consumption, and demographics
  - Greater detail on wealth and consumption added in 1999
- Follow Blundell et al. (2016) to construct measure of consumption.
- Income taxes estimated using NBER Taxsim software
  - Also estimate counterfactual taxes for un-chosen retirement state (incl. social security)
- Returns obtained from Federal Reserve Bank of St. Louis (FRED)

# Summary Statistics

	All	Working	Retired	Difference	
Age	47.23 (15.85)	41.84 (11.60)	65.33 (14.74)	-23.49***	-143.58
Household Size	2.67 (1.47)	2.84 (1.47)	2.09 (1.30)	0.74***	47.70
Years of Education	13.61 (2.63)	13.82 (2.50)	12.89 (2.93)	0.93***	28.27
Health Status	2.45 (1.05)	2.29 (0.97)	2.98 (1.14)	-0.69***	-53.67
Total Income	49646.97 (49138.39)	59272.33 (49018.81)	17338.86 (33155.70)	41933.46***	96.51
Total Wealth	440766.38 (604700.33)	461108.95 (598687.96)	372485.25 (619625.53)	88623.71***	12.40
Consumption	25925.37 (19159.19)	27287.45 (19686.96)	21353.48 (16466.95)	5933.97***	29.59
Observations	41884	32270	9614	41884	

# Estimation

---

## Two-step

- In the first stage, we estimate the policy functions and recover  $\zeta_{it}$ , and the second stage,
- In the second stage, the marginal utility functions and the discount factor are estimated non-parametrically using a pairwise differencing strategy (Honore and Powell, 2005)

# First Stage: discrete choice

- In the first stage, we estimate the policy functions and recover  $\zeta_{it}$ , and the second stage, we estimate the structural parameters.
- We estimate the unconditional probability of individual  $i$  choosing alternative  $j$  by SMLE:

$$LL = \sum_{i,j,t} d_{ijt} \log \overline{Pr}_{ijt}(\bar{s}_{it}; \lambda^d) \quad (1)$$

where  $\overline{Pr}_{ijt}(\bar{s}_{it}; \lambda^d) = \int Pr_{ijt}(\bar{s}_{it}, \zeta; \lambda^d) g(\zeta) d\zeta$



# First Stage: Retired households continuous policy

Estimate continuous policy function and recover  $\zeta_{it}$ .

- Being retire is an absorbing action, the probability of remaining in this absorbing choice is one. It also satisfies the conditional independence assumption as the current choice is a sufficient statistic for the future value of the lagged choice.
- We thus estimate the continuous policy function for retired households by MLE. Specifically, we specify
$$\ln q_{iRt}^* = \mu(\bar{S}_{iRt}^\mu; \lambda_R^{c1}) + \sigma(\bar{S}_{iRt}^\sigma; \lambda_R^{c2}) \cdot \zeta_{it}$$
- As the policy is invertible, we can retrieve the unobserved  $\zeta$  for all households that newly choose to retire or who are already retired, i.e.  $\hat{\zeta}_{it}(q_{iRt}|d_t = R) = q_{iRt}^{*-1}(q_{iRt}; \hat{\lambda}_R^c)$  where  $q_{iRt}$  is the observed quantity consumed.

# First Stage: Working households continuous policy

Retrieve the continuous policy function and unobserved  $\zeta_{it}$  for those households who choose to work.

- As before  $\ln q_{iWt}^* = \mu(\bar{S}_{iWt}^\mu; \lambda_W^{c1}) + \sigma(\bar{S}_{iWt}^\sigma; \lambda_W^{c2}) \cdot \zeta_{it}$ .
- To estimate the unknown parameters  $\lambda_W^{c2}$ , we use a GMM estimator where the set of moment conditions on  $\zeta_{it}$  which is function of the unknown parameters entering the working policy function:

$$\zeta_{it} = q_{iRt}^{*-1}(q_{iRt}; \hat{\lambda}_R^c) \cdot 1_{j=R} + q_{iWt}^{*-1}(q_{iWt}; \lambda_W^c) \cdot 1_{j=W} \quad (2)$$

## Second Stage: Structural Parameters

The estimation follows the proof of Theorem 1.

- Consider 2 households who make different discrete choices (for concreteness, assume  $d_{it} = W$ ), but *would have* chosen the same quantity had both chosen  $d_{i't} = R$
- Once we express the marginal utility of choosing  $W$  in terms of  $R$ , the two equations have the same unknown marginal utility part  $\partial u_R(q_{iRt}^*, s_t) / \partial q_{it}$ .
  - Note that  $\ln(Pr_W / Pr_R) = v_W - v_R$
- $\zeta_{it}$  is generically different from  $\zeta_{i't}$ ,

## Second Stage

we take the difference of the F.O.C.s which determine the continuous choice for each household and we get:

$$\frac{\partial \Psi_W(\bar{s}_{it})}{\partial L} + \zeta_{it} - \zeta_{i't} = \delta_{it} \mathbb{E} \left[ (1 + r_W(\bar{s}_{it})) \mathcal{V}(s_{t+1}, L_W, W, \zeta_{it+1}) - (1 + r_R(\bar{s}_{i't})) \mathcal{V}(s_{t+1}, L_R, R, \zeta_{i't+1}) \right] \quad (3)$$

where  $\mathcal{V}(s_{t+1}, L_{t+1}, d_t, \zeta_{t+1}) = \mathbb{E}_{\zeta_{t+1}} \left[ \frac{\partial \Phi_R(s_{t+1}, L_{t+1}, d_t, \zeta_{t+1})}{\partial L_{t+1}} + \frac{\partial u_R(q_{iRt+1}^*, s_{t+1})}{\partial q_{i,t+1}} + \zeta_{i,t+1} \right]$  is the expected marginal continuation value.

## Second Stage

- If we remove the expectation and add the the expectational error term  $\eta_{it}$ , we can write the equation above as:

$$\Upsilon_{ii't} = \delta_i X_{ii't} + \eta_{ii't} + \delta_i \left( (1 + r_{d_{it}}(\bar{s}_{it})) \frac{\partial u_R(q_{iRt+1}^*, s_{it+1})}{\partial q_{Rt+1}} - (1 + r_{d_{i't}}(\bar{s}_{i't})) \frac{\partial u_R(q_{i'Rt+1}^*, s_{i't+1})}{\partial q_{Rt+1}} \right) \quad (4)$$

- $\Upsilon_{ii't} \equiv \frac{\partial \Psi_{d_{it}}(\bar{s}_{it})}{\partial L_t} + \zeta_{it} - \zeta_{i't}$
- $X_{ii't} \equiv \left( (1 + r_{d_{it}}) \frac{\partial \Phi_{iR}(s_{it+1}, L_{it+1}, d_{it}, \zeta_{it+1})}{\partial L_{it+1}} - (1 + r_{d_{i't}}) \frac{\partial \Phi_{i'R}(s_{i't+1}, L_{i't+1}, d_{i't}, \zeta_{i't+1})}{\partial L_{i't+1}} \right)$
- the discount factors and the marginal utility are the only unknown objects.

## Second Stage: Pairwise differencing

- in practice to write the equation above we need to match two individuals with the same unknown marginal utility at time  $t$
- two individuals to have the same (unknown) marginal utility is that they have the same pay-off relevant state variables and the same optimal quantity chosen.
- We match on the discrete pay-off relevant states (family size and age) and a continuous choice  $q_t$
- we then use a Gaussian kernel to assign weights to each potential match on the basis of differences in their optimal consumption conditioned to household's discrete choice

## Second Stage: Continuous Choice

- We model the marginal utility as a linear spline for each state
- The equation (4) can be estimated as a WLS/weighted non-linear LS

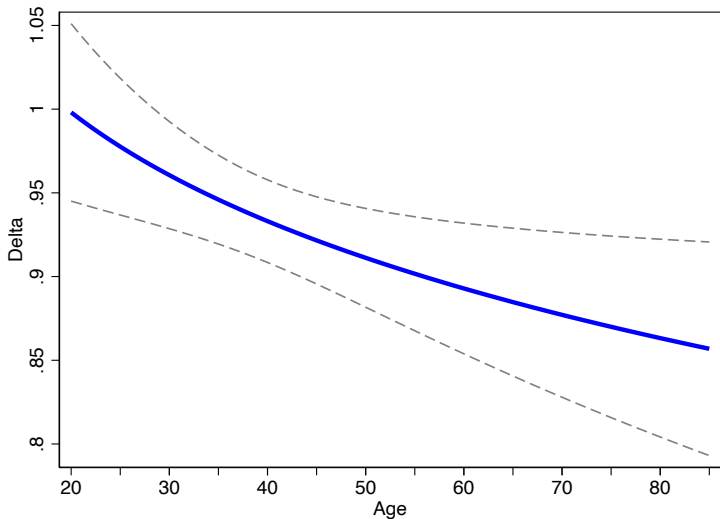
# Results

---



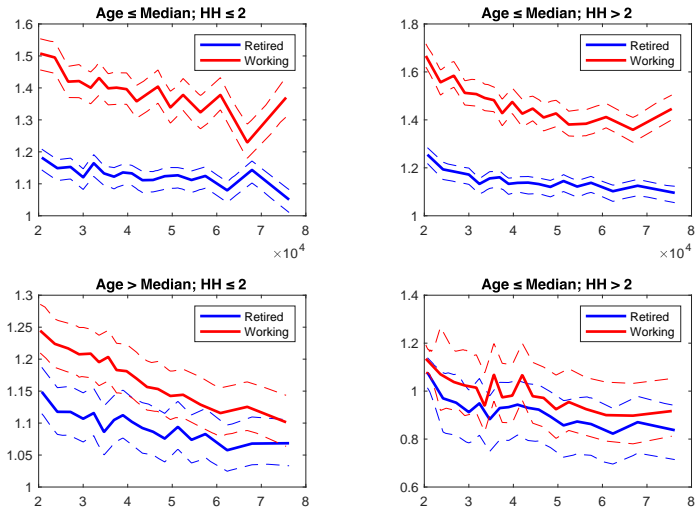
# Discount Factors

Figure 1: Estimated Discount Factors



# Marginal Utilities

Figure 2: Estimated Marginal Utilities



- We can use these estimates to calculate a coefficient of relative risk aversion.
- At the mean consumption level, we find a RRA coefficient of approximately 0.5 for retired households and 0.3 for working households.

# Conclusion

---

# Conclusion

- We show conditions under which  $\delta$  and utility are identified
- We use a two-step semi-parametric estimator to estimate a model of lifecycle consumption/retirement
- Heterogeneity in discount factor based on age