3 notes on wealth distribution and social mobility

By Shu HU

0 intro

0.1 motivation

- wealth is unequally distrbuted
- high social mobility
 - o intergenerational shorrocks mobility index: 0.88

0.2 econ question

- what factors
 - o drive wealth dynamics
 - o are consistent with the observed
 - cross-sectional wealth distribution
 - social mobility

0.3 methods

- develop a simple ifp model
- match the moments generated by model by empirical moments of
 - the observed wealth distribution
 - social mobility matrix
- use this matched model to quantitatively identify the effects of 3 critical factors on the tail of the distribution (generating thick tail) and social mobility
 - \circ A. skew and persistent distribution of stochastic earnings (:= Y)
 - B. Differential saving rates across wealth levels (:= s/a)
 - higher saving and accumulation rates for the rich
 - forms in literature
 - takes the form of non-homogeneous bequests
 - bequests as a fraction of wealth that are increasing in wealth
 - \circ C. stochastic idiosyncratic returns to wealth/capital income risk (:= r=R-1)
 - allowing rates of return on wealth to be increasing in wealth might also add to the skewness of the distribution

0.4 Conclusions

- capital income risk
- differential savings
- stochastic earnings
 - limited role in filling the tail of wealth distribution (how to understand "filling")
 - o fundamental in inducing enough mobility in the wealth process
- r increases in wealth itself
 - supported in the estimates
 - o improve the fit of the model across the wealth distribution

0.5 drawbacks

• without directly observing return data, the mechanism above is poorly identified

1 model

1.1 setups

- ullet each agent's lives T years
- every period t
 - \circ consumers choose consumption c_t and
 - \circ accumulate wealth a_t
 - subject to a no-borrowing constraint
- consumers leave wealth a_T as a bequest at the end of life T.
- each agent's preferences are composed of
 - \circ a per-period utility from consumption, $u(c_t)$, at any period $t=1,\ldots,T$
 - $u(c_t) = \frac{c_t^{1-\sigma}}{1-\sigma}$
 - a warm-glow utility from bequests at T, $e(a_T)$

$$lacksquare e(a_T) = A rac{a_T^{1-\mu}}{1-\mu}$$

- wealth accumulates from
 - o savings
 - bequests
- idiosyncratic rates of returns r and life-time labor earnings y (w in his paper)
 - o are drawn from a distribution at birth
 - possibly correlated with those of the parent
 - deterministic within each generatioin
 - stochastic over generations only
 - o lifetime earnings processes are hump-shaped, with low earnings early in life
 - o no-borrowing constraints limit how much agents can smoothing lifetime earnings

1.2 notations

- discount rate $\beta < 1$
- ullet present discounted utility of an agent with wealth a_t at the beginning of period t: $V_t(a_t)$
- initial wealth: a_0
- earnings profiles: w
- rate of return: r

1.3 each agent's maximization problem

- $\bullet \ \ V_t(a) = \operatorname{max}_{c,a'} u(c) + \beta V_{t+1}(a')$
- s.t.
 - a' = (1+r)a c + w
 - $c \in [0, a], t = 1, \dots, T 1$
 - $V_T(a) = u(c) + e(a')$

1.4 optimal wealth processes

- $\bullet \ \ a_T = g(a_0; r, w)$
 - \circ CASE 1: if $\mu=\sigma$, then $g(a_0;r,w)=lpha(r,w)a_0+eta(r,w)$
- $\alpha(r, w)$ is the savings rate and independent of wealth
 - \circ under general conditions (what conditions?), $a^n{}_n$ has a stationary distribution whose tail
 - is independent of the distribution of earnings
 - asymptotic to a Pareto law
- CASE 2: if $\mu < \sigma$, then $g''(a_0; r, w) > 0$.
 - \circ keep σ constant, differential savings rate emerge
 - increasing with wealth
 - a stationary distribution might not exist
 - o if stationary dist exist, then it displays a thick tail (proved?)
- stochastic process over generations: r^n, w^n
 - o finite markov chain
 - \circ r^n are w^n independent, but allowed to be serially correlated with transition matrix
 - life-cycle structure
 - $a^n = a_0^n = a_T^{n-1}$
- $\bullet \ a^n = g(a^{n-1}; r^n, w^n)$
 - \circ stochastic process over generations: a^n _n for initial wealth a

1.5 easy to extend to: Markov states of the stochastic process for r to depend on the initial wealth of the agent a

• optimal wealth processes have similar properties to **CASE 2** above.

2 quantitative analysis

2.1 methodology

- estimate parameters in the model using the method of simulated moments (MSM) estimators
 - fix (or externally calibrate several parameters of the model
 - o select some relevant moments of the wealth process as target in the estimation
 - estimate the remaining parameters by matching the targeted moments generated by the stationary distribution induced by the model and those in the data
- assumption of quantitative exercise
 - wealth and social mobility observed in the data are generated by a stationary distribution

Too strong?

- Simulated model can be solved by using **collocation method**, proposed by Miranda and Fackler (2004)
- quantitative exercise
 - \circ 1. fix σ, T, β
 - lacktriangledown stochastic individual earnings: w^n with transition matrix P (across generations) estimated from data
 - PSID
 - federal income tax records (Chetty et al, 2014)
 - o 2. Target 12 moments
 - bottom 20% + 20-40% + 40-60% + 60-80% + 80-90% + 90-95% + 95-99% + top 1% | wealth shares
 - the diagonal of the (age-independent) social mobility markov chain transition matrix defined over quintiles
 - 3. Estimate 12 parameters
 - preference parameters: μ , A
 - lacksquare parameters of r
 - defined
 - by 5 states r_i
 - 5 diagonal transition probabilities: $P(r_i|r_i)$
 - o In 3-4,
 - lacktriangleright modify r to allow returns to depend on the initial wealth a of the agent
 - experiment with an alternative social mobility matrix
 - defined over the same percentiles of the wealth distribution

2.2 data

labor earnings

- use 10 deterministic life-cycle household-level earnings profiles at different deciles (construct the earning profiles)
 - fix each of 6 age brackets, compute averages of the earnings deciles (Heathcote et al 2010) from PSID, 1967-2002 (Table 1)
 - assume agents stay in the same decile for their whole lifetime.
 - agentss randomly draw one of these earnings profiles at the beginning of life according to an intergenerational transition matrix (10 states) reduced from (from Chetty et al, 2014; 100 states; 1980-1982 us birth cohort and their parental income) (Figure 1)
- wealth distribution (Fig2+Table 2)
 - Survey of Consumer Finances 2007
 - wealth variable is net wealth (:=net financial wealth + housing any debts)
 - skewed-to-right (heavy tail) distribution
 - take the shares from Diaz-Gimenez et al (2011)
- social mobility (wealth transition across generations)
 - o mobility matrix calculated by Charles and Hurst (2003) from PSID (Table 3)
 - constructed by means of pairs of simultaneously alive parent and child of different ages
 - compute transitions from the residuals of the wealth of parents and children after conditioning on age and age squared
 - to eliminate age effects
 - shorrocks index: 0.88
- in 3-4, reproduce estimation
 - using an alternative social mobility matrix, using 2007-09 SCF panel data
 - with transitions computed for a synthetic agent over his/her age profile

3 estimation results

3.1 parameter estimates (baseline: (Table4))

- upper part
 - estimates of preference parameters
- lower part
 - estimated state space
 - \circ diagonal of the transition matrix of the five-state Markov process for r we postulate
 - implied mean and standard deviation of the process
 - \circ autocorrelation computed fitting an AR(1) on simulated data from the estimated process
 - standard errors; obtained by bootstrapping

3.2 model fit (compare targeted simulated moments of the estimated model with counterpart in the data) (Table 5)

- compare moments in data with those in simulated
- compute age-independent social mobility moments after conditioning on age and agesquared

3.3 discussion and interpretation

- differential savings and bequests (Table 6
 - - rich save proportionally more than poor
 - $\circ \mu$ depends on A
 - average saving rates
 - bequests
 - its dist implied by model is very skewed
 - should include inter vivos transfer (? a transfer made during one's lifetime)
- returns to wealth (Table 7)
 - \circ significant standard deviation and wide dispersion in r
 - cautions
 - lacktriangledown is assumed constant over each agent's lifetime, disregarding the whole variation across the life cycle
 - lacktriangle difference in risk composition of investment portfolio might also contribute heterogeneous r in the data
- social mobility (Table 8)
 - fit well compared with Charles and Hurst (2003, Table 2; Table 3 in this paper)
 - slightly overestimate the mobility from the top to the bottom of the distribution vice versa
 - shorrocks index: 0.92

3.4 extensions and robustness

- re-estimate model
 - to allow for r dependent on wealth
 - o to match an alternative social mobility matrix constructed using 2007-09 panel data
- rate of return positively dependent on wealth
 - \circ positive correlation between r and a does not implies r increases with wealth a
 - think about average
 - this add to the skewness of the distribution
 - \circ re-estimate the model using a r dependent on a
 - Assume $r = r_0 + b \times p(a)$
 - where
 - p(a) = 1, 2, ..., 8 moments
 - r_0 is a five-state Markov process
 - results

- better parameters fits
- better wealth distribution fits
 - top 20%
- worse social mobility fits
 - dependence of r on wealth a is compensated by a reduced dependence of savings?
- see Fagereng et al (2017)
- alternative social mobility matrix
 - baseline social mobility matrix
 - constructed by means of pairs of simultaneously alive parents and child
 - does not account for any transition induced by bequests
 - only available for wealth transitions between quintiles, but we care about the heavy tails
 - transitions in and out of the top 1 percent are one of the most relevant characteristics of the stochastic process of the wealth accumulation
 - re-estimate the model using a new social mobility matrix (Table11)
 - method
 - intergenerational transition matrix: each element of the matrix take the form $Pr(a_0^n \in p|a_0^{n-1}p')$, where p, p' are generic percentiles of the wealth distribution
 - lacktriangle intra-generational transition matrix: $Pr(a_T^{n-1} \in p|a_0^{n-1} \in p')$
 - divide agent's lifetime *T* into *k*-periods age groups
 - use the Markov assumption (?) to obtain $Pr(a_T^{n-1} \in p|a_0^{n-1}p')$ from the observation of $Pr(a_k^{n-1} \in p|a_0^{n-1} \in p')$, $Pr(a_{2k}^{n-1} \in p|a_k^{n-1} \in p')$ and so on for all age groups.
 - first construct age-dependent two-year transition matrices for age groups running from 30-31 to 66-67
 - multiply these age-dependent two-year transition matrices for all age groups
 - transitions computed for the same percentiles they use as wealth distribution moments
 - it produces spurious mobility due to measurement error in wealth
 - results
 - display substantial social mobility (shorrocks index: 0.98 vs 0.88 (baseline))
 - re-estimate model
 - gives estimates close to baselines (Table 12)
 - also give fits close to baselines (Table 13)

4 counterfactual estimates (by shutting down one factor at a time)

4.1 motives

- understand which mechanism mostly affects which dimension of the wealth distribution and mobility
- interpret the counterfactuals as informal tests of identification of these mechanism
 - lack of identification implies that shutting down one or more of the mechanism has limited effects on the fit for the targeted moments

4.2 re-estimation

- parameters estimates (Table 14)
- model fit (Table 15)

4.3 re-estimation results (see conclusion in intro part)

- no capital income risk (constant *r*)
- no stochastic earnings (constant w)
- homogeneous saving rates ($\mu = 2$)

5 transitional dynamics of the wealth distribution

5.1 assumption relaxation

- assumptions above: the observed distribution of wealth is a stationary distribution
 - our estimates are obtained by matching the data with the moments of the stationary distribution generated by the model.
- relax the stationary assumption on wealth distribution
 - try to match the transitional dynamics of the distribution of wealth

5.2 method

- use observed SCF 1962-1963 wealth distribution as initial condition
- estimate the parameters of the model by matching the implied distribution after 72 years (two iterations of the model) with (see Table 16)
 - o observed SCF 2007 distribution
 - transition matrix adopted in previous analysis

5.3 results (see Table 17)

- such a dramatic increase in wealth inequality can be obtained by exploiting the power of capital income risk and differential savings
- A larger bequest motive

- a rate of return process with higher mean and volatility and much more auto-correlation
 - o induce a simulated wealth distribution for 2007 is even more skewed at the top
 - the bottom 40% of the wealth distribution is very well matched
- not well fits on social mobility by overestimating mobility all across the distribution
 - mobility is the probability that children move away from their parents' wealth cell