The Life-Cycle Dynamics of Wealth Mobility

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November, 2025

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- Intergenerational "social" wealth mobility key as context for large wealth inequality
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 - How much? Who moves how? What is behind these mobility patterns?

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Today: Flexibly and non-parametrically characterize lifetime wealth mobility

Possible with Norwegian administrative data on wealth 1993–2017

This paper

- 1. Study individuals as they transition across the wealth distribution over their lives
 - Study individuals' relative and absolute mobility (within-cohort wealth ranks + wealth levels)
 - But: as many different wealth histories as individuals
 - Use clustering techniques to find "typical" trajectories responsible for mobility

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 - Use clustering techniques to find "typical" trajectories responsible for mobility
- 2. Study how "typical" trajectories relate to other observable characteristics
 - Role of heterogeneity in income and returns
 - Lifetime choices and events (portfolio composition, income, etc.)
 - To which extent do individual characteristics at age 30 predict future trajectories?

Contributions

- 1. New evidence on wealth mobility and wealth accumulation: Full life cycle trajectories
 - Add to results for the super wealthy (Gomez; Ozkan, Hubmer, Salgado, Halvorsen), the role of individual factors (Huggett, Ventura, Yaron; Black, Devereux, Landaud, Salvanes), and short-run mobility and race (Hurst, Luoh, Stafford, Gale).
- 2. New facts documenting the distribution of changes in wealth ranks
 - Extensive literature on income (Guvenen, Ozkan, Karahan, Song; Guvenen, Pistaferri, Violante; Arellano, Blundell, Bonhomme; De Nardi, Fella, Paz-Pardo)
- 3. Inter-generational links to full life cycle wealth dynamics
 - Complements "snapshot" links in income (Solon; Aaronson, Mazumder; Chetty, Hendren, Kline, Saez, Turner; Chetty, Grusky, Hendren, Hell, Manduca, Narang) & wealth (Charles, Hurst; Boserup, Kopczuk, Kreiner; Fagereng, Guiso, Malacrino, Pistaferri; Fagereng, Mogstad, Rønning)
- 4. Dimension reduction methods in economics & applications to labor markets
 - K-Means (Bonhomme, Lamadon, Manresa; Gregory, Menzio, Wiczer), Sequence Analysis (Humphries), Hidden Markov (Ahn, Hobijn, Şahin), Finite Mixture

Norwegian Wealth Data

Data: Norwegian Tax Registry 1993 - 2017

- No top-coding + Limited misreporting or measurement error (third-party reporting)
 - Focus on wealth (e.g., don't include public pensions)
 - No transaction data (e.g., changing houses or selling stocks → limited info. on returns)
- We adjust the tax value of real estate to market values (Fagereng, Holm, Torstensen, 2023)
- We focus on wealth at the individual level (additional results for household wealth)
- Key: We link to administrative records (Education, Family, Civil Status, Income)

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Sample selection: Norwegian residents 1993–2017 (no immigrants after 25/2011, no emigrants)

- Focus on birth cohort born between 1960 and 1965 (first observed in early 30s)
 - 292,222 individuals in this sample (279,002 after balancing)

Ranks and Histories

- Compute within cohort ranks as

$$r_{i,t} = 100 \times F_w(w_{i,t}|t, i \in BC(i))$$

- Computed separately for each year and each cohort

- Trajectories: Histories of ranks

$$\mathbf{R}_i = (r_{i,1993}, r_{i,1994}, \dots, r_{i,2016}, r_{i,2017}) \in [0, 100]^{25}$$

We are interested in the distribution of the trajectories \mathbf{R}_i

- Relative mobility in rank \implies absolute mobility in wealth level
 - At the median, +10 ranks ≈ 60 k USD at age 40

Wealth and Income Mobility

Measuring average intra-generational mobility

Relative Mobility Measures

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Relative Mobility Measures

- Rank-rank persistence: $r_{i,t} = \alpha_t + \rho_t r_{i,0} + u_{i,t}$; $M_t^R \equiv 1 \rho_t$.
- Shorrocks Index: Transitions out of quintiles; $M_t^S \equiv 1 \sum_i \mathbb{1}\{Q_{it} = Q_{i.0}\}.$

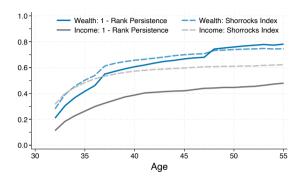
Measuring average intra-generational mobility

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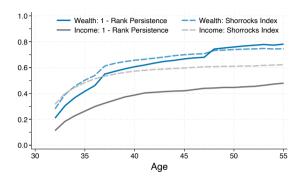
Exercise: Plot intra-generational relative mobility for income and wealth

Wealth is more mobile than income (!)



- Declining intra-generational persistence \longrightarrow Increased (cumulative) mobility
 - Wealth: $M_t^R = 0.78$ and $M_t^S = 0.75$ by age 55
 - Income: $M_t^R = 0.48$ and $M_t^S = 0.58$ by age 55

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 - Wealth: $M_t^R=0.78$ and $M_t^S=0.75$ by age 55 Income: $M_t^R=0.48$ and $M_t^S=0.58$ by age 55
- How broad-based is mobility? What (who) drives patterns?

Clustering Wealth Histories

Grouping Individuals Into Typical Histories

Goal: Identify patterns in (ex-post) life cycle paths without restricting to a single statistic

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Method: Agglomerative Hierarchical Clustering to group rank histories

- Start with G = N groups (one for each individual)
- Recursively merge groups by selecting *similar* pairs: $\underset{g,g' \in G, \ g \neq g'}{\mathsf{argmin}} d(g,g').$

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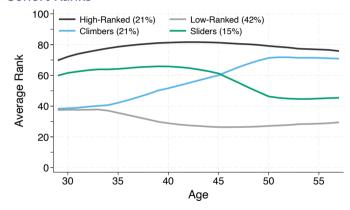
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Result: Hierarchy of partitions ranging from G = N to G = 1.

- Global result with nested clusters (feasible in large datasets)
- Asymptotically consistent as we observe longer trajectories, even for fixed *N* (Borysov, Hannig, Marron, 2014; Egashira, Yata, Aoshima, 2024)
- We use G = 4 in our baseline: exposition + 50% of variation in ranks

Typical Rank Histories

Cohort Ranks

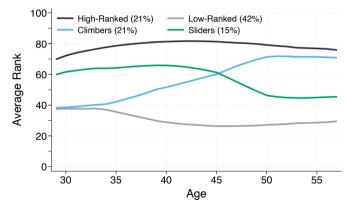


Four largest groups

- Wealthy/High Ranked: always at top of the distribution
- Poor/Low Ranked: always at the bottom of the distribution
- Middle: one group of Climbers and one group of Sliders

Typical Rank Histories

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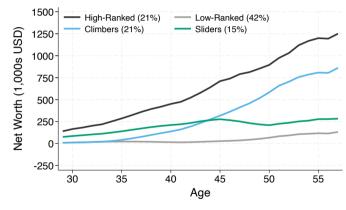


Robust pattern

- Alternative clustering algorithm: same pattern
- Not mechanical: get four horizontal lines if we cluster with income ranks
- Still a lot of within-cluster variation
 - Segmented mobility
 - Sub-cluster analysis
 - Not today!

Wealth Histories Across Segments of the Distribution



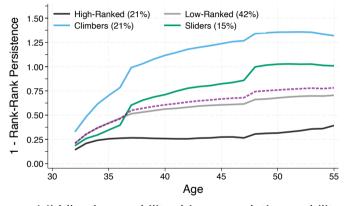


Significant diff. in wealth profiles

- Top: Maintaining rank means level growth (8-10%)
- Bottom: Stay very low
- Climbers: Grow on avg. 18%/y
- Sliders: ahead in 30s + low growth (5%) + Great Recession

Decomposing Mobility

Linear rank-rank persistence:
$$y_{i,t}^k = \alpha_t + \rho_t^{g(i)} y_{i,0}^k + u_{i,t}$$



- Top: Immobile over 25y
- Bottom: Track population movements within segment
- Climbers: Reversal of fortune within 1 decade
- Sliders: No memory in long run

- Middle-class mobility drives population mobility patterns. Climbers are key.

Interpreting Mobility Groups

Wealth mobility in models of wealth dynamics

Two exercises to contextualize our results:

- 1. Buffer-stock models of savings (Zeldes, 1989; Deaton, 1991; Carroll, 1992; Straub, 2019)
 - Skill differences (alone) cannot generate observed wealth+income dynamics
 - Analytical results building on Straub (2019)

Wealth mobility in models of wealth dynamics

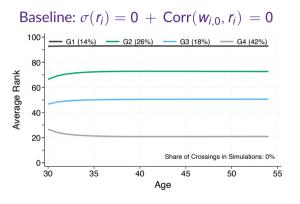
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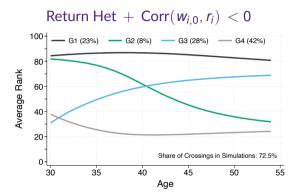
- 1. Buffer-stock models of savings (Zeldes, 1989; Deaton, 1991; Carroll, 1992; Straub, 2019)
 - Skill differences (alone) cannot generate observed wealth+income dynamics
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- 2. Statistical models of wealth (Benhabib & Bisin, 2018; Gomez, 2023)
 - Approximation for broad class of life-cycle models
 - Vary parameters of two-equation model

$$w_{i,t+1} = (1 + r_i) w_{i,t} + s y_{i,t};$$
 $\log y_{i,t+1} = \rho \log y_{i,t} + \varepsilon_{i,t}^{y};$ $\varepsilon_{i,t}^{y} \sim N(0, \sigma(\varepsilon^{y}))$

Wealth mobility in models of wealth dynamics

From literature: s=0.25 (Fagereng, Holm, Moll, Natvik, 2019) and ρ , $\sigma(\epsilon^Y)$ (Fagereng, Holm, Natvik, 2021)

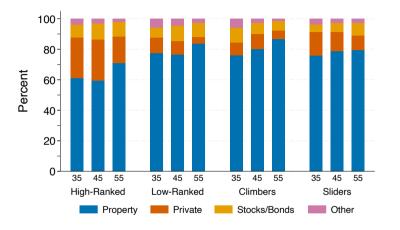




- Return or savings rate heterogeneity is key along with distribution of initial conditions
- Challenge: Marginal distribution of wealth by age is off by a lot in standard models!

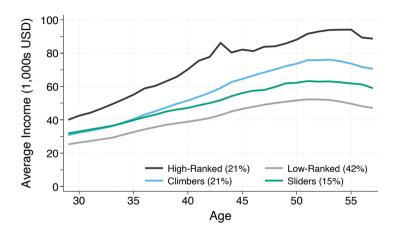
Drivers of Wealth Accumulation

Portfolio composition: Mostly housing (except at the top!)



- Private business wealth more important at the top and for sliders

Income Histories Across Segments of the Distribution



- Distribution of individual income across clusters compressed relative to wealth

Towards Determinants of Trajectories

Hereditary Advantage: Wealth vs Human Capital

Goal: Understand role of different circumstances/characteristics in determining trajectories

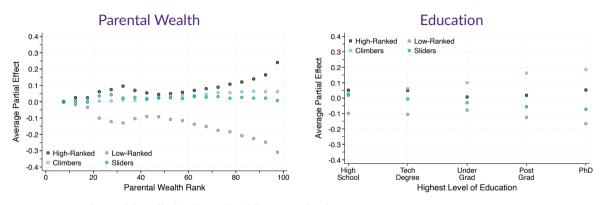
Hereditary Advantage: Wealth vs Human Capital

Goal: Understand role of different circumstances/characteristics in determining trajectories

$$\Pr\left(g = j\right) = F\left(\alpha_0^j + \beta_{q(i)}^j + \gamma_{\textit{educ}(i)}^j + \delta_{\textit{subj}(i)}^j + \lambda_{\textit{male}(i)}^j + \mu_{\textit{bcounty}(i)}^j\right)$$

- $\beta_{q(i)}^{j}$: Indicators for 1993 parental wealth (cohort rank by ventile)
- $\gamma^{j}_{educ(i)}$, $\delta^{j}_{subj(i)}$: Indicators for education level and subject (only for higher ed.)
- $\lambda_{male(i)}^{j}$: Indicator for sex
- $\mu_{bcounty(i)}^{j}$: Indicator for birth location

Non-Linear Effects of Parental Wealth and Education



- Parental wealth tells high-ranked/low-ranked apart
- Education tells climbers/sliders apart

Summary

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Contribution: Flexibly and non-parametrically characterize lifetime wealth mobility

Key takeaways:

- 1. Find evidence of substantial changes in wealth ranks over a quarter century
- 2. Mobility driven by selected groups in the middle of the distribution
- 3. Simulations point to differences across groups beyond income (returns, savings)
- 4. Parental background and education predict distinct wealth trajectories