## On the Nature of Entrepreneurship

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## Disclaimer

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# This Paper

- Informs theories of entrepreneurship
- How?
  - Assembles novel longitudinal database of business owners
  - Studies patterns of life-cycle income profiles
  - Analyzes determinants of entrepreneurial choice

# Most Previous Work

- Uses surveys with
  - Top-coding
  - Short panels
- Concludes that self-employed (relative to peers)
  - Have flatter life-cycle profiles
  - Enter self-employment with lower past labor income
  - Enter with higher past asset income
- Motivates theories where entrepreneurs
  - Earn large non-pecuniary benefits
  - Are misfits
  - Face liquidity constraints

# In Contrast to Literature

- Use administrative data with IRS vs CPS
  - No Top-coding
  - Long panels
- Conclude that self-employed (relative to peers)
  - Have significantly steeper life-cycle profiles
  - Enter self-employment with higher past labor income
  - Enter with lower past asset income
- Motivate theories where entrepreneurs
  - Make significant investments in business
  - Are not misfits
  - Face few liquidity constraints

# Data

# Sample

• Primary source: administrative IRS data

- Balanced panel of living individuals with US SSN
- Tax years 2000-2015
- Birth cohorts 1950-1975
- Income Measures:
  - Self-employment (SE) income:
    - Schedule C net profits
    - Schedule K-1 ordinary business income
    - W-2 wages of S-corporation owners
  - Paid-employment (PE) income:
    - W-2 wages of non-owners

# **Employment Status**

- Self-employed (SE) in a given year if:
  - $\circ~|{\mbox{SE}}$  income| > 5,000 in 2012\$ and at least one of:
    - |SE income| > PE income or
    - Share of gross profits  $> \mathsf{PE}$  income or
    - Share in business  $\times$  employees  $\ge 1$
- Paid-employed (PE) in a given year if:
  - Not SE
  - PE income > 5,000 in 2012\$
- Non-employed (NE) in a given year if:
  - Not SE or PE

# Skill and Education Measures

Skills:

- Individuals with occupation in e-filing
  - Map entry to SOC code
  - Map SOC to cognitive, interpersonal, and manual skills (as in Lise and Postel-Vinay 2020)
- Individuals with missing codes
  - $\circ~$  Use AI tools and data for peers with codes

Education:

• Use CPS-based classifier

# Life-Cycle Profile Estimation

**Object of Interest** 

# Income(Age | Individual and aggregate factors)

• Statistical model for income:

$$y_{it} = \alpha_i + \beta_{g(i),t} + \sum_{a=a_0}^{a(i,t)} \gamma_{c(i),g(i)}^a + \epsilon_{i,t}$$

- $\circ i \in \mathcal{I}$  is set of individuals
- $\circ \ t \in \mathcal{T}$  is set of calendar dates
- $\circ \ c \in \mathcal{C}$  is set of birth years
- $\circ a \in \mathcal{A}$  is set of ages
- $\circ \ g \in \mathcal{G}$  is set of groups partitioning  $\mathcal I$

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time effects

- $\circ$   $i \in \mathcal{I}$  is set of individuals
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$$y_{it} = \alpha_i + \beta_{g(i),t} + \sum_{a=a_0}^{a(i,t)} \gamma_{c(i),g(i)}^{a} + \epsilon_{i,t}$$
  
age effects

- $\circ i \in \mathcal{I}$  is set of individuals
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- $\circ \ c \in \mathcal{C}$  is set of birth years
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• Estimation of time ( $\Delta\beta$ ), age ( $\gamma$ ) effects:

$$\Delta y_{i,t} = \underbrace{\Delta \beta_{g(i),t} + \gamma_{c(i),g(i)}^{a(i,t)}}_{\text{identification}} + \Delta \epsilon_{i,t}.$$

Identification:

• Assume that age effects are constant across binned cohorts

#### o Normalize time effects to reflect group-specific growth

More details on identification assumptions

# Application: set G with 46,080 subgroups

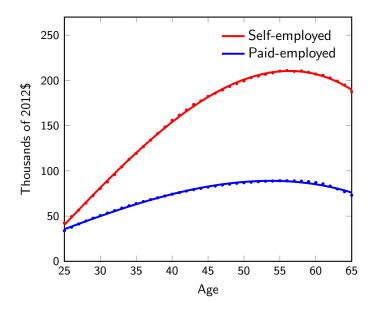
- Time-invariant characteristics include usual ones:
  - Cohort, gender, educated, skilled (cognitively, interpersonally, manually), industry, married, children
- Plus partition sample based on *Employment attachment* 
  - Attached SE, Attached PE, Switchers Definitions

# Main Empirical Results

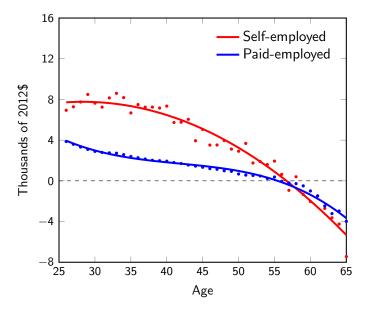
## Income and Growth Profiles

- Attached self-employed
  - o Income similar on average to paid-employed when 25
  - o Growth significantly higher and more persistent
- $\Rightarrow$  Entrepreneurial investment does pay

### Income Profiles: Attached Subsamples



## Growth Profiles: Attached Subsamples



## Volatility Patterns

• Large literature on risk in entreprenurship

• Is SE more risky than PE? By how much?

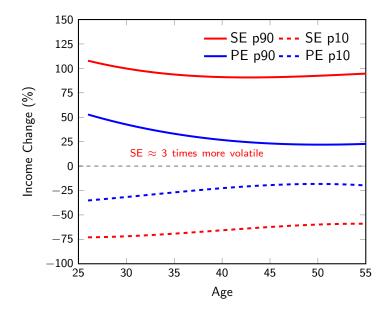
- Are differences in growth driven by increasing risk over age?
- Compute distribution of residuals (net of time-age effects)

$$\Delta \epsilon_{i,a}/|y_{i,a-1}|$$

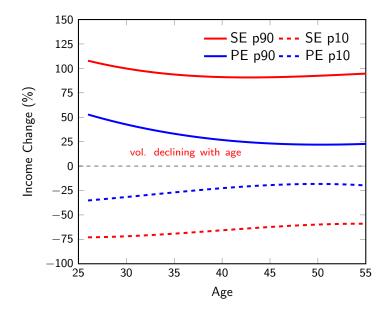
Compare SE and PE

 $\,\circ\,$  Plot  $10^{th}$  and  $90^{th}$  percentiles by age and employment status

### Income More Volatile for Attached SE



### Income More Volatile for Attached SE



## **Entrepreneurial Choice**

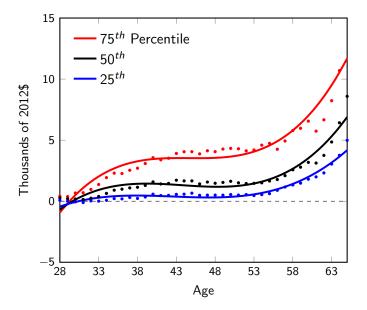
- Entry and exit rates
  - Results similar to surveys
- Use switchers to study
  - Key determinants of choosing self-employment

## Determinants of Self-Employment

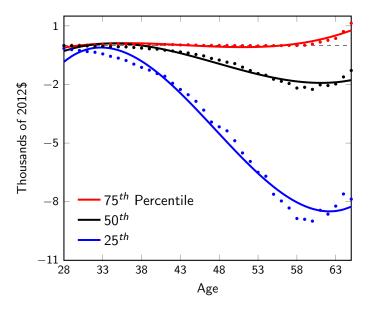
• Compare SE entrants to "similar" peers

- One-time entrants into SE ("Treatment")
- Future switchers with same characteristics ("Control")
- Assess "misfit" hypothesis for SE
  - Compare wage income before entry
- Assess "financial-friction" hypothesis for SE
  - Compare asset income before entry

## Past Wage Incomes Higher for Switchers



## Past Asset Incomes Lower for Switchers



# Informing Theory

# **Empirically-Motivated Features**

- Patterns in the data
  - Hump-shaped and persistent income growth
  - Declining exit rates
  - Volatility decreasing with age
- Empirical results suggest three model features
  - Investment in self-created intangible assets Evidence
  - Incomplete information about entrepreneurial productivity
  - Slow adjustment in achieving optimal size

• State vector 
$$s = [a, \kappa, j, \epsilon, z, \mu]$$

$$V_k(s) = \max\{U(c,\ell) + \beta EV(s')\}$$
  

$$a' = (1+r)a + pe^z f_y(\kappa, h_y, k, n) - (r+\delta_k)k - wn - e - c \ge 0$$
  

$$\kappa' = (1-\delta_\kappa)\kappa + f_\kappa(h_\kappa, e)$$
  

$$\ell = 1 - h_y - h_\kappa$$

- Two production technologies:
  - $f_y(\kappa, h_y, k, n)$ : goods and services
  - $f_{\kappa}(h_{\kappa}, e)$ : new intangible assets

• State vector 
$$s = [a, \kappa, j, \epsilon, z, \mu]$$
  
financial assets

$$V_k(s) = \max\{U(c,\ell) + \beta EV(s')\}$$
  

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intangible assets

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- Two production technologies:
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  - $f_{\kappa}(h_{\kappa}, e)$ : new intangible assets

• State vector 
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age

$$V_k(s) = \max\{U(c,\ell) + \beta EV(s')\}$$
  

$$a' = (1+r)a + pe^z f_y(\kappa, h_y, k, n) - (r+\delta_k)k - wn - e - c \ge 0$$
  

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$$\ell = 1 - h_y - h_\kappa$$

- Two production technologies:
  - $f_y(\kappa, h_y, k, n)$ : goods and services
  - $f_{\kappa}(h_{\kappa}, e)$ : new intangible assets

• State vector 
$$s = [a, \kappa, j, \epsilon, z, \mu]$$
  
true and predicted skills

$$V_k(s) = \max\{U(c,\ell) + \beta EV(s')\}$$
  

$$a' = (1+r)a + pe^{z}f_y(\kappa, h_y, k, n) - (r+\delta_k)k - wn - e - c \ge 0$$
  

$$\kappa' = (1-\delta_{\kappa})\kappa + f_{\kappa}(h_{\kappa}, e)$$
  

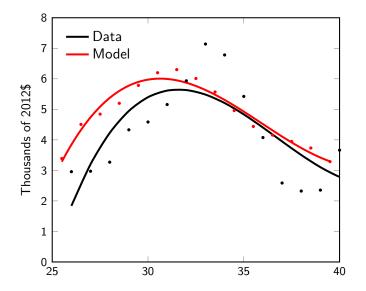
$$\ell = 1 - h_y - h_{\kappa}$$

- Two production technologies:
  - $f_y(\kappa, h_y, k, n)$ : goods and services
  - $f_{\kappa}(h_{\kappa}, e)$ : new intangible assets

# Comparing Growth Profiles

- Choose income shocks consistent with IRS micro data
- Simulate time series over the life cycle
- Aggregate simulations using IRS counts and entry ages
- Construct growth differential for self-employed:
  - $\circ~$  Stayers: attached to self-employment past age 35 ~
  - $\circ~$  Switchers: ran a business at least 5 years but exited by 35

# Growth Differentials for Young Entrepreneurs



# Conclusion

- Assembled novel longitudinal database for business owners
- Estimated life-cycle income profiles for many groups
- Developed prototype model of entrepreneurs
- Studied model predictions for IRS data

# Appendix

# Identification

- Two identifying assumptions
  - Age effects are same across binned cohorts ( $\geq 2$ )
  - Average time effect satisfies (where  $\overline{y}_{g,t_0}$  is avg income for g):

$$rac{\overline{\Deltaeta_g}}{\overline{y}_{m{g},t_0}} = rac{\mu_{m{g}}}{T}\sum_t (1+\mu_{m{g}})^t$$

• Allows flexibility when set  ${\mathcal G}$  large

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# **Employment Attachment**

- Attached (SE or PE) if:
  - Fewer than 2 switches in status during sample
  - No itermediate spells of non-employment
- Mostly switchers if:
  - $\circ~$  In SE or PE for 12+ years
  - No intermediate spells of non-employment
- Any non-employment if:
  - $\circ~$  Switched in/out of NE from SE or PE at least once
  - Or, 5 years of NE during sample

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# Evidence of Business Intangibles

- Business sale is taxable event for buyer and seller
- Forms 8594, 8883 show assets primarily intangible, eg
  - Customer bases, client lists, non-compete covenants
  - Licenses, permits, trademarks, tradenames
  - Workforce in place
  - Goodwill and on-going concern value



# Empirical Moments: IRS vs CPS (Thous. 2012\$)

