## The GRID Project: An Overview and Some New Insights

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GRID: Global Repository of Income Dynamics

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- harmonized database
- of detailed micro statistics
- **3** on income inequality & income dynamics
- 4 based on panel data
- **5** from administrative records

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- ► 500,000 to 1,500,000 statistics per country.
- Database and user-friendly website recently launched: www.grid-database.org
- ► A Special Issue of *Quantitative Economics* with 13 papers written by country teams was published in November 2022.



#### SPECIAL ISSUE ON GLOBAL INCOME DYNAMICS

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- **Europe:** Austria, Belgium, Finland, Greece, Hungary, Iceland, Ireland, Israel, Netherlands, Poland, Portugal, Switzerland
- Australasia: Australia, India, Japan, New Zealand, Singapore, South Korea, Taiwan
- South America: Chile, Colombia, Costa Rica, Ecuador, Peru, Uruguay.

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► Goal is to have 40 countries in GRID by Summer 2024.

#### Background: Data on Income Inequality

- Several harmonized cross-country databases (of statistics) on income inequality are available:
  - World Inequality Database (WID.world) spearheaded by the work of T. Atkinson, T. Piketty and E. Saez
  - World Income Inequality Database (WIID2) maintained at the United Nations University
  - OECD Income Distribution Database (IDD)
  - Luxembourg Income Study (LIS)

### Why This Project?

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- focused on cross-sectional statistics (e.g. inequality)
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- GRID aims to extend the focus to:
  - Administrative data, now available for many countries
  - The study of income dynamics (mobility, income instability, etc.) and other long-run phenomena
  - Finely-defined subpopulations (examples later)

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  - Income changes characterized by high probability of extreme events:
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- Ability to link data across various sources in some countries (income, expenditures, health, education)
- Caveat: In countries with large informal sectors (BRA, MEX, ARG), coverage limited to formal sector
  - Teams for these countries asked to validate findings using survey data; ongoing work

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- Provides static snapshots of a distribution, but misses dynamics
- Two societies can have similar levels of inequality, but very different chances of mobility
- Panel data:
  - Allows study of several dimensions important for welfare:
    - Income Dynamics (volatility, skewness, persistence, etc.)
    - Mobility (intra- and inter-generationally)
    - ► Tail behavior (probability of extreme wage cuts vs. wage hikes)
    - Inequality in long-run ("permanent") income

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- ▶ In GRID, many statistics are reported for groups defined by
  - the interaction of gender x cohort x age group x "permanent" income.
  - Permanent income based on past 3-year average income. 40 quantile bins plus more at the top.

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- Emphasis on providing flexible quantile-based statistics to allow computation of alternative measures:
  - Inequality: Top shares, Gini, Standard deviation, P90-P10, etc.
  - Volatility: P90-P10 for income changes, etc.
  - Skewness: 3rd centralized moment, Kelley's skewness, etc.

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Statistics calculated for 4 different income measures: raw income, log income, residualized income, "permanent" income.

#### Overview of Available Statistics

#### Income levels (static):

- Time-series statistics (up to 35+ years for some countries)
- Measures of inequality for annual and "permanent" income
- Empirical densities

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- 1-year and 5-year income changes (transitory and persistent changes)
- Log change and arc-percent change (to allow extensive margin—zero income)
- Moments on dispersion, skewness, and kurtosis (tails)
- Short- and Long-run mobility statistics
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- A quick look at the GRID website: https://data.grid-database.org

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#### But most importantly:

 Single master code to generate all statistics from very first step of cleaning raw data to the very end results. New Insights from GRID: Inequality

### Question 1: Is Income Inequality Rising Globally?

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Figure 1: Trends in Overall Income Inequality for GRID Countries



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#### Question 2: How Much Does Average Income Grow with Age?

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Figure 3: Average Life Cycle Profile of Income, Men



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#### Question 3: What Fraction of Top Earners are Women?

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New Insights from GRID: Dynamics of Income

#### Question 4: Has Income Instability Increased Over Time?

Answer crucial for many questions and policy design.

#### Rising income uncertainty in the US became conventional wisdom.

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- Answer crucial for many questions and policy design.
- Seminal paper by Moffitt and Gottschalk (1994) found:
  - Income volatility increased substantially in survey data from 1968 to 1988.
  - 30+ papers confirmed the result and extended to 2010s.

Rising income uncertainty in the US became conventional wisdom.

#### US Income Volatility in Survey Data Up



Figure 4: Variance of Income Growth (2-year)

Moffitt and Zhang (AEA P&P, 2018)

► Variance of income growth in PSID nearly tripled from 1970s to 2010s.

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That income uncertainty is higher today has become conventional wisdom.

#### US Income Volatility Trending UP



#### Left: Moffitt and Zhang (AEA P&P, 2018)

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US Income Volatility Trending Up? or Down?



Left: Moffitt and Zhang (AEA P&P, 2018)

Right: Guvenen, Ozkan, Song (JPE, 2014)

US Income Volatility Trending Up? or Down?



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**Right: GRID Project** 

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#### How About Other Countries? Evidence from GRID

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### Taking Stock

#### Overall, income volatility has been

- declining in several countries (Brazil, Argentina, USA, UK after mid-1980s)
- relatively flat in some others (Canada, Denmark, France, Germany, Mexico, Spain, Sweden), and
- increasing in a few (men in Italy, Norway, and women in UK pre-1990s).

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- We do not see any broad-based increase in income instability as suggested by US survey data.
- GRID contains very granular statistics that allows digging deeper into this question.

### Question 5: How Does Income Volatility Vary by Income Level?

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Figure 5: USA



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#### Question 6: Great Gatsby Curve

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- ▶ GRID Version 2.0 will triple the number of countries in the database.
- ▶ We have future plans to expand the set of variables included in GRID.
- We welcome all feedback and suggestions. Drop us a line at support@grid-database.org
- Follow us on Twitter for future updates: @griddatabase
- Thanks again to WCEG for giving us this opportunity to talk about GRID!

# Thanks!