

Big Data for Macroeconomists and Policy Makers

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CEPR & TFI Event – Household Welfare: Do We Need Big Data?

A total of five exabytes of information was created between the dawn of civilization through 2003. That much information is now created every two days.

Eric Schmidt, former CEO of Google, 2010

Big data in household finance

1. Why are big data useful to the **policy maker**?
 - **Ex-ante:** to *monitor* household financial fragility
 - **Throughout:** to *target* policy interventions
 - **Ex-post:** to *evaluate* effectiveness of intervention

Big data in household finance

1. Why are big data useful to the **policy maker**?
 - **Ex-ante**: to *monitor* household financial fragility
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 - **Ex-post**: to *evaluate* effectiveness of intervention
2. Are big data always **superior** to survey data?
 - *Some thoughts*

Ex-ante: monitor household balance sheets

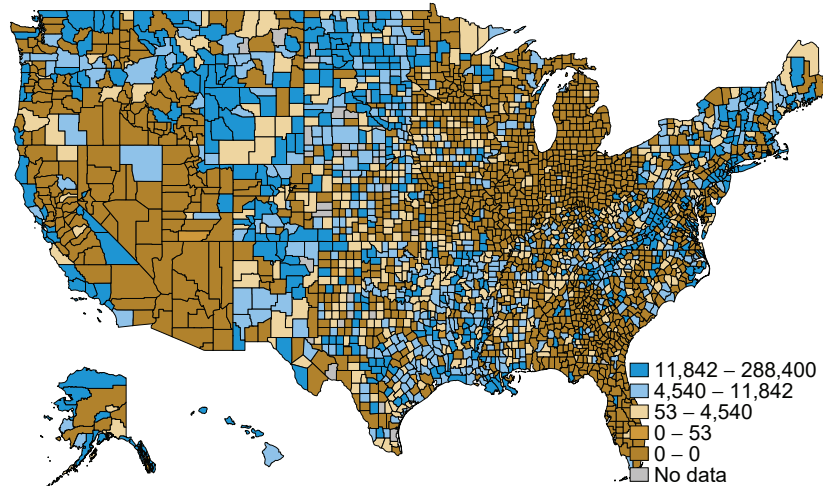
- Lesson from GR: **interconnection** btw hh and bank balance sheets
 - Policy response: stress tests on banks' side
 - Feasible because banks collect big data at their own cost

Ex-ante: monitor household balance sheets

- Lesson from GR: **interconnection** btw hh and bank balance sheets
 - Policy response: stress tests on banks' side
 - Feasible because banks collect big data at their own cost
- Equally important, but absent: **households' stress tests**
 - Need: detailed, high frequency data on hh balance sheets
 - Simulate 'crisis scenarios' and estimate impact of shocks
 - Inform intelligent design of policy
 - Ongoing research with economists at the NY Fed

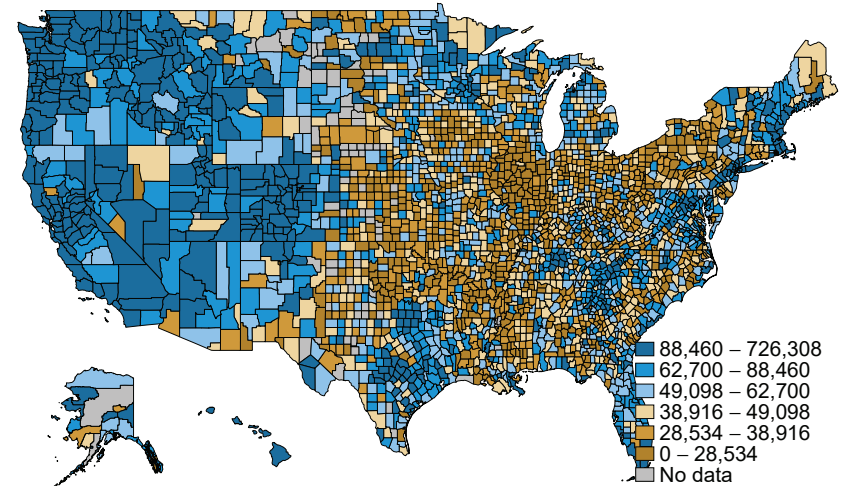
Ex-ante: monitor household balance sheets

Median Non-Negative Tappable Equity, 2011Q1



Source: CRISM. See Fuster et al., 2018, updated.
Non-negative tappable equity defined as $\max(0, 0.9 \times \text{Asset Value} - \text{Total Balance})$.

Median Non-Negative Tappable Equity, 2019Q1



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- Only one determinant of **household shock absorption capacity**
- Challenging measurement (with available data), but important
- You can trace distribution by: area, income, age, credit score, etc.

Throughout: target policies

- **Social insurance**: credit or transfers to households in need
- Crisis: urgency calls for **easy-to-implement** policies
 - Easy to implement: depends on **available data** you have
 - You end up with universal policies with low impact
 - **CARES Act**: Any family of 4 with $AGI < \$150,000$ gets \$3,400
- Big data allow to tailor intervention: **more bang for the buck**
- You channel financial help to those who **really** need it

Throughout: target policies

- **Lockdown** has heterogeneous effects in the population
 - Key: whether you are in a **flexible or rigid** occupation
 - Dingel-Neiman definitions based on O*NET tasks

Occupation	Income	Flexible
Economist	\$100K	Y
Optometrist	\$100K	N
Secondary school teacher	\$50K	Y
Flight attendant	\$50K	N
Telemarketer	\$25K	Y
Cook	\$25K	N

- A transfer **conditional on occupation** could be more generous

Ex-post: policy evaluation

1. Impact evaluation using **diff-in-diff**
 - Big data improve design of the quasi-experiment
 - Big data allow to capture **heterogeneous treatment effects**
 - Cannot capture GE feedbacks (the intercept)

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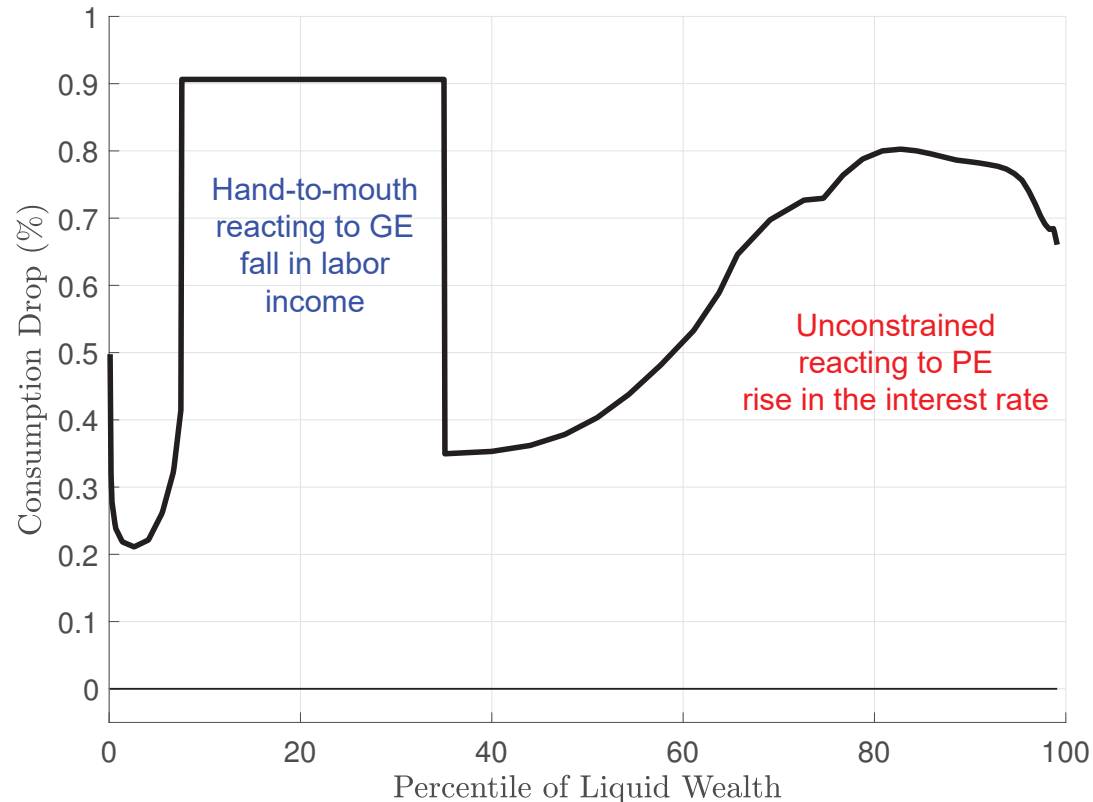
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2. Impact evaluation using: **structural equilibrium models**

- State-of-the-art macro models capture heterogeneity in MPC
- Key for transmission mechanism of shock and policy response
- Big data allow to properly **calibrate** these models
- Assess **distributional impact of policy**

Ex-post: policy evaluation

- Example: **monetary policy tightening** in HANK models



- Do we see this same heterogeneity in (big) data?

Are big data always superior to survey data?

1. Virtues of big data vis-à-vis survey data

- Large-scale
- Contain novel variables (e.g., individual stocks, mobility)
- High-frequency
- Real time
- High quality (less measurement error)

They allow to truly **embrace and measure heterogeneity**

(household balance sheets, preferences, financial literacy, etc...)

Are big data always superior to survey data?

2. Challenges associated with the use of big data

- **Representativeness** of the underlying universe
 - Data that are not meant to be representative
 - Coverage is often time-varying / endogenous
- **Access** to the dataset is often restricted
 - Inequality in access (connections / research budget)
 - Replicability of the findings
- **Privacy** issues
 - It's about protecting: **not an argument against collecting**