Nonlinear Persistence and Partial Insurance
Income Dynamics and Consumption Insurance in the PSID

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1. to consider *alternative ways of modelling persistence*, and

2. to explore *the nonlinear nature of income shocks and the implications for consumption dynamics and inequality*. 
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In particular,

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2. to explore *the nonlinear nature of income shocks and the implications for consumption dynamics and inequality*.

⇒ key data: PSID, also Norwegian Population Registers ....
New data on consumption and family income sources

I. Newly designed panel data: PSID since 1999.
   - Collection of consumption and assets had a major revision in 1999
     - ~70% of consumption expenditures, [90+% since 2004.]
     - The sum of food at home, food away from home, gasoline, health, transportation, utilities, etc.
   - Earnings and hours for all earners; Assets measured in each wave.
     - e.g. Blundell, Pistaferri and Saporta-Eksten (2016).

II. Administrative linked data: e.g. Norwegian population register.
   - Linked registry databases with unique individual identifiers.
     - Containing records for every Norwegian from 1967 to 2014.
     - Detailed demographic and socioeconomic information (market income, cash transfers). Hours of work. Recent links to real estate and assets; => New consumption measurements.
   - Family identifiers allow to match spouses and children.
     - e.g. Blundell, Graber and Mogstad (2015).
A prototypical “canonical” of (log) family (earned) income $y_{it}$ is:

$$y_{it} = \eta_{it} + \varepsilon_{it}, \quad i = 1, \ldots, N, \quad t = 1, \ldots, T.$$ 

where $y_{it}$ is net of a **systematic component**, $\eta_{it}$ is a **random walk** with innovation $v_{it}$,

$$\eta_{it} = \eta_{it-1} + v_{it}, \quad i = 1, \ldots, N, \quad t = 1, \ldots, T.$$ 

and $\varepsilon_{it}$ is a **transitory shock**.
The prototypical panel data model

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  $$\eta_{it} = \eta_{it-1} + v_{it}, \quad i = 1, ..., N, \quad t = 1, ..., T.$$  

  and $\varepsilon_{it}$ is a transitory shock.

- Consumption growth is then related to income shocks:

  $$\Delta c_{it} = \phi_{t} v_{it} + \psi_{t} \varepsilon_{it} + v_{it}, \quad i = 1, ..., N, \quad t = 1, ..., T.$$  

  where $c_{it}$ is log total consumption net of a systematic component,

  > $\phi_{t}$ is the transmission of persistence shocks $v_{it}$, and

  > $\psi_{t}$ the transmission of transitory shocks;

  - the $v_{it}$ are taste shocks, assumed to be independent across periods.
Covariance Restrictions and Partial Insurance Parameters

Baseline panel data model specification:

\[ \triangle c_{it} = \phi v_{it} + \psi \epsilon_{it} + \nu_{it}, \]

\[ \triangle y_{it} = \nu_{it} + \triangle \epsilon_{it}, \]

Implying covariance restrictions:

\[ \text{var}(\triangle c_{it}) = \phi^2 \sigma^2_v + \psi^2 \sigma^2_\epsilon + \sigma^2_\nu \]

\[ \text{var}(\triangle y_{it}) = \sigma^2_\eta + 2\sigma^2_\epsilon \]

\[ \text{cov}(\triangle y_{it} \triangle y_{it-1}) = -\sigma^2_\epsilon \]

\[ \text{cov}(\triangle c_{it} \triangle y_{it}) = \phi \sigma^2_v + \psi \sigma^2_\epsilon \]

\[ \text{cov}(\triangle c_{it-1} \triangle y_{it}) = -\psi \sigma^2_\epsilon \]

> For \( T > 3 \), BPP include time(age) variation in the \( \sigma^2_* \) and insurance parameters,

> Allow for measurement error and extend to MA(1) transitory shocks,

> BP develop these covariance restrictions for repeated cross-sections.
This “standard” framework implies a set of covariance restrictions,

⇒ BPP show (over-)identification & efficient estimation via nonlinear GMM.
Nonlinear Income Dynamics

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Linearity of the income (or wage) process simplifies estimation,
▷ however, it rules out the nonlinear transmission of shocks and general interactions between assets and shocks in the measures of partial insurance.

The aim in the new work is to step back and take a different tack:
▷ develop an alternative approach to modeling persistence in which the impact of past shocks can be altered by the size and sign of new shocks.

• The idea is to have a framework that allows:
⇒ “unusual” shocks to wipe out the memory of past shocks, and
⇒ future persistence of a current shock to depend on the future shocks.
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• The idea is to have a framework that allows:

⇒ “unusual” shocks to wipe out the memory of past shocks, and
⇒ future persistence of a current shock to depend on the future shocks.

• Nonlinear persistence and the presence of “unusual” shocks is shown to match the data well, and has a key impact on consumption and saving over the life cycle.
Nonlinear Persistence

- Let $y_{it}$ denote log-labor income, net of a systematic component, for $i$ of age $t$
  \[ y_{it} = \eta_{it} + \varepsilon_{it}, \quad i = 1, \ldots, N, \quad t = 1, \ldots, T. \]

- Denote the $\tau$th conditional quantile of $\eta_{it}$ given $\eta_{i,t-1}$ as $Q_t(\eta_{i,t-1}, \tau)$, and specify
  \[ \eta_{it} = Q_t(\eta_{i,t-1}, u_{it}), \quad \text{where} \quad (u_{it} | \eta_{i,t-1}, \eta_{i,t-2}, \ldots) \sim \text{Uniform}(0, 1). \]

- Consider the following measure of persistence:
  \[ \rho_t(\eta_{i,t-1}, \tau) = \frac{\partial Q_t(\eta_{i,t-1}, \tau)}{\partial \eta}. \]

  $\Rightarrow \rho_t(\eta_{i,t-1}, \tau)$ measures the persistence of $\eta_{i,t-1}$ when, at age $t$, it is hit by a shock $u_{it}$ that has rank $\tau$. Measures the persistence of histories.

- Allows general conditional heteroscedasticity, skewness and kurtosis.

- But what is the evidence for such nonlinearities in persistence?
Some motivating evidence: Quantile autoregressions of log-earnings

\[
\frac{\partial Q_{y_t | y_{t-1}}(y_{i,t-1}, \tau)}{\partial y}
\]

PSID data

Norwegian administrative data

Estimates of the average derivative of the conditional quantile function of \(y_{i,t}\) given \(y_{i,t-1}\).
An empirical model for consumption and partial insurance

- Model consumption as a flexible/‘nonparametric’ nonlinear rule.
- Let $c_{it}$ and $a_{it}$ denote log-consumption and assets (beginning of period) net of age dummies. Our empirical specification is based on

$$c_{it} = g_t (a_{it}, \eta_{it}, \varepsilon_{it}, \nu_{it}) \quad t = 1, \ldots, T,$$

where $\nu_{it}$ are independent across periods, and $g_t$ is a nonlinear, age-dependent function, monotone in $\nu_{it}$ taste shifter.
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where $\nu_{it}$ are independent across periods, and $g_t$ is a nonlinear, age-dependent function, monotone in $\nu_{it}$ taste shifter.

> Consistent with the standard life-cycle model, e.g. Kaplan and Violante (2010).

> Consumption responses to $\eta$ and $\varepsilon$ are given by transmission/partial insurance coefficients

$$\phi_t(a, \eta, \varepsilon) = \mathbb{E} \left[ \frac{\partial g_t(a, \eta, \varepsilon, \nu)}{\partial \eta} \right], \quad \psi_t(a, \eta, \varepsilon) = \mathbb{E} \left[ \frac{\partial g_t(a, \eta, \varepsilon, \nu)}{\partial \varepsilon} \right].$$
Similar techniques can be used in the presence of *advance information*, e.g.

\[ c_{it} = g_t \left( a_{it}, \eta_{it}, \eta_{i,t+1}, \varepsilon_{it}, \nu_{it} \right), \]

or *consumption habits*, e.g.

\[ c_{it} = g_t \left( c_{i,t-1}, a_{it}, \eta_{it}, \varepsilon_{it}, \nu_{it} \right). \]

Also cases where the consumption rule depends on lagged \( \eta \), or when \( \eta \) follows a second-order Markov process. (See Section 3 in *ABB*, 2017).
Extensions for habits and unobserved heterogeneity

- Similar techniques can be used in the presence of *advance information*, e.g.

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▷ also cases where the consumption rule depends on lagged \( \eta \), or when \( \eta \) follows a second-order Markov process. (See Section 3 in ABB, 2017).

- Households can also differ in their initial productivity \( \eta_1 \) and initial assets, the panel data provide opportunities to allow for additional, *unobserved heterogeneity* in earnings and consumption.

- E.g. consumption rule with *unobserved heterogeneity*:

\[
c_{it} = g_t \left( a_{it}, \eta_{it}, \epsilon_{it}, \zeta_i, \nu_{it} \right).
\]
Data: PSID

- (New) PSID 1999 - 2009, we use 6 waves (every other year), as in BPS.
- $C_{it}$: Information on food expenditures, rents, health expenditures, utilities, car-related expenditures, ..... 
- $A_{it}$: Assets holdings are the sum of financial assets, real estate value, pension funds, and car value, net of mortgages and other debt. (Net worth).
- $y_{it}$ are residuals of log total pre-tax household labor earnings on a set of demographics. Note, $c_{it}$ and $a_{it}$ are residuals, using the same set of demographics as for earnings.

▷ cohort and calendar time dummies, family size and composition, education, race, and state dummies.

- As in BPS, we select married male heads aged between 25 and 59.
- In this talk we focus on a balanced sub-sample of $N = 792$ households.
Specification and estimation: summary

• The quantile function of $\eta_{it}$ given $\eta_{i,t-1}$ is specified as

$$Q_t(\eta_{t-1}, \tau) = Q(\eta_{t-1}, age_t, \tau) = \sum_{k=0}^{K} a^Q_k(\tau) \varphi_k(\eta_{t-1}, age_t),$$

where $\varphi_k$, $k = 0, 1, ..., K$, are polynomials (Hermite). Similarly for $\varepsilon_{it}$ etc.

• Consumption (log) function:

$$g_t(a_t, \eta_t, \varepsilon_t, \tau) = g(a_t, \eta_t, \varepsilon_t, age_t, \tau) = \sum_{k=1}^{K} b^g_k \tilde{\varphi}_k(a_t, \eta_t, \varepsilon_t, age_t) + b^g_0(\tau).$$

1. The first estimation step recovers estimates of the income parameters.
2. The second step recovers estimates of the consumption parameters, given an estimate of the income parameters.

• The estimation algorithm alternates between draws of latent variables from candidate posteriors and quantile regressions using those draws.
RESULTS
Nonlinear persistence of $\eta_{it}$ (PSID):

$$\rho_t(\eta_{i,t-1}, \tau) = \frac{\partial Q_{\eta_t|\eta_{t-1}}(\eta_{i,t-1}, \tau)}{\partial \eta}$$

Note: Estimates of the average derivative of the conditional quantile function of $\eta_{it}$ on $\eta_{i,t-1}$ with respect to $\eta_{i,t-1}$, evaluated at percentile $\tau_{\text{shock}}$ and at a value of $\eta_{i,t-1}$ that corresponds to the $\tau_{\text{init}}$ percentile of the distribution of $\eta_{i,t-1}$. Evaluated at mean age in the sample.
Nonlinear persistence of $y_{it}$

$$\frac{\partial Q_{y_t|y_{t-1}}(y_{i,t-1},\tau)}{\partial y}$$

PSID panel data

Nonlinear model

Note: Estimates of the average derivative of the conditional quantile function of $y_{it}$ given $y_{i,t-1}$ with respect to $y_{i,t-1}$, evaluated at percentile $\tau_{\text{shock}}$ and at a value of $y_{i,t-1}$ that corresponds to the $\tau_{\text{init}}$ percentile of the dist. of $y_{i,t-1}$.
Nonlinear persistence of $y_{it}$

$$\frac{\partial Q_{yt|y_{t-1}}(y_{i,t-1}, \tau)}{\partial y}$$

Norwegian register data

Nonlinear model

**Note:** Estimates of the average derivative of the conditional quantile function of $y_{it}$ given $y_{i,t-1}$ with respect to $y_{i,t-1}$, evaluated at percentile $\tau_{\text{shock}}$ and at a value of $y_{i,t-1}$ that corresponds to the $\tau_{\text{init}}$ percentile of the dist. of $y_{i,t-1}$. 
Nonlinear persistence of $y_{it}$ (cont.)

$$\frac{\partial Q_{y_{t}\mid y_{t-1}}(y_{i,t-1},\tau)}{\partial y}$$

PSID data  
Canonical model

Note: Estimates of the average derivative of the conditional quantile function of $y_{it}$ given $y_{i,t-1}$ with respect to $y_{i,t-1}$, evaluated at percentile $\tau_{\text{shock}}$ and at a value of $y_{i,t-1}$ that corresponds to the $\tau_{\text{init}}$ percentile of the dist. of $y_{i,t-1}$.
Nonlinear persistence, 95% confidence bands

(a) Earnings, PSID data          (b) Earnings, nonlinear model

Note: Pointwise 95% confidence bands. Parametric bootstrap, 500 replications.
Consumption response to $\eta_{it}$, by assets and age

$$\bar{\phi}_t(a) = \mathbb{E} \left[ \frac{\partial g_t(a, \eta_{it}, \varepsilon_{it}, \nu_{it})}{\partial \eta} \right], \text{ nonlinear model}$$

Note: Estimates of the average consumption response $\bar{\phi}_t(a)$ to variations in $\eta_{it}$, evaluated at $\tau_{\text{assets}}$ and $\tau_{\text{age}}$. 
Consumption responses to $y_{it}$, by assets and age

$$\mathbb{E} \left[ \frac{\partial}{\partial y} \bigg|_{y_{it}} \mathbb{E} \left( c_{it} \big| a_{it} = a, y_{it} = y, age_{it} = age \right) \right]$$

**Note:** Estimates of the average derivative of the conditional mean of $c_{it}$ given $y_{it}$, $a_{it}$ & $age_{it}$ with respect to $y_{it}$, evaluated at values of $a_{it}$ & $age_{it}$ corresponding to their $\tau_{assets}$ & $\tau_{age}$ percentiles, and averaged over the values of $y_{it}$. 
What about individual wages?

Measured Nonlinear Persistence in the PSID Male Wage Data:

Notes: Log male wages, Age 30-60 PSID 1999-2013 (US). Estimates of the average derivative of the conditional quantile function. Source: Authors calculations.
Results for hourly male wages from the nonlinear model - find an important role for unusual shocks and nonlinear persistence.

(a) Wages, nonlinear model

(b) Persistent component $\eta_{it}$

Notes: Log male wages, Age 30-60 PSID 1999-2013 (US). Simulation of the average derivative of the estimated conditional quantile function from the $\eta$ and $\varepsilon$. 
SIMULATIONS
Impulse responses, by age and initial assets

**Earnings**

\[ \tau_{init} = .9, \ \tau_{shock} = .1 \]

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\[ \tau_{init} = .1, \ \tau_{shock} = .9 \]

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**Consumption**

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Notes: Initial assets at age 35 (for “young” households) or 51 (for “old” households) are at percentile 0.10 (dashed curves) and 0.90 (solid curves). Linear assets accumulation rule. Assets are constrained to be non-negative.
Summary

• New framework to shed new light on income persistence and the nonlinear transmission of income shocks to consumption.

• Show the complementarities between ‘big’ administrative data, like the Norwegian registers, and purpose designed panel surveys, like the PSID.

• Exploit the important new measurements for consumption and assets in the PSID.
Summary

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• Exploit the important new measurements for consumption and assets in the PSID.

▷ A Markovian permanent-transitory model of household income, which reveals asymmetric persistence of unusual shocks.
▷ An age-dependent nonlinear consumption rule that is a function of assets, permanent income and transitory income.
Summary

- New framework to shed new light on income persistence and the nonlinear transmission of income shocks to consumption.
- Show the complementarities between ‘big’ administrative data, like the Norwegian registers, and purpose designed panel surveys, like the PSID.
- Exploit the important new measurements for consumption and assets in the PSID.
  - A Markovian permanent-transitory model of household income, which reveals asymmetric persistence of unusual shocks.
  - An age-dependent nonlinear consumption rule that is a function of assets, permanent income and transitory income.
- This framework for nonlinear persistence in income dynamics leads to new empirical measures of the degree of partial insurance and the link between income and consumption inequality.
- Nonlinearities are also found to hold in individual male wages.
Next steps

1. Explore the impact of nonlinear persistence in wages on family labour supply and consumption smoothing, building on Blundell, Saporta-Eksten and Pistaferri (2010).

2. Include firm to firm transitions and lay-offs.

3. Separate housing equity and allow a role for local labour markets.

4. Health and other types of (partially insured) shocks.

5. and more...... [would be great to have links with soc sec earnings etc]

Congratulations PSID at 50!

Thanks for everything, for all those many micro-data innovations and looking forward to the next 50 years!
Descriptive statistics PSID (means)

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<tr>
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<th>1999</th>
<th>2001</th>
<th>2003</th>
<th>2005</th>
<th>2007</th>
<th>2009</th>
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</thead>
<tbody>
<tr>
<td>Earnings</td>
<td>85,001</td>
<td>93,984</td>
<td>100,281</td>
<td>106,684</td>
<td>119,039</td>
<td>122,908</td>
</tr>
<tr>
<td>Consumption</td>
<td>30,182</td>
<td>35,846</td>
<td>39,843</td>
<td>47,636</td>
<td>52,175</td>
<td>50,583</td>
</tr>
<tr>
<td>Assets</td>
<td>266,958</td>
<td>315,866</td>
<td>376,485</td>
<td>399,901</td>
<td>501,590</td>
<td>460,262</td>
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</tbody>
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Notes: Balanced subsample from PSID, $N = 749$, $T = 6$.

- Compared to BPS (12), households in our balanced sample have higher assets, and to a less extent higher earnings and consumption.
Note: *Skewness measured as a nonparametric estimate of*

\[
\frac{Q_{y_t|y_{t-1}}(y_{i,t-1}, .9) + Q_{y_t|y_{t-1}}(y_{i,t-1}, .1) - 2Q_{y_t|y_{t-1}}(y_{i,t-1}, .5)}{Q_{y_t|y_{t-1}}(y_{i,t-1}, .9) - Q_{y_t|y_{t-1}}(y_{i,t-1}, .1)}.
\]
Figure: Densities of persistent and transitory earnings components (PSID)

(a) Persistent component $\eta_{it}$

(b) Transitory component $\varepsilon_{it}$

Note: Nonparametric estimates of densities based on simulated data according to the nonlinear model, using a Gaussian kernel.
Simulating the responses to income shocks

- Simulate life-cycle earnings and consumption after a shock to the persistent earnings component (at age 37).

- We report the difference between:
  - Households that are hit by a “bad” shock ($\tau_{\text{shock}} = .10$) or by a “good” shock ($\tau_{\text{shock}} = .90$).
  - Households that are hit by a median shock $\tau = .5$.

- Age-specific averages across 100,000 simulations. At age 35 all households have the same persistent component (percentile $\tau_{\text{init}}$).
Impulse responses, earnings

Bad shock: $\tau_{shock} = .1$

$\tau_{init} = .1$

$\tau_{init} = .5$

$\tau_{init} = .9$

Good shock: $\tau_{shock} = .9$

$\tau_{init} = .1$

$\tau_{init} = .5$

$\tau_{init} = .9$
More specifically, to account for the impact of income shocks on the evolution of consumption inequality we introduce *transmission* or *partial insurance* parameters, writing consumption growth as:

\[
\Delta \ln C_{it} \approx \gamma_{it} + \Delta Z_{it}' \varphi + \phi_t \nu_{it} + \psi_t \varepsilon_{it} + \xi_{it}
\]

\(\phi_t\) and \(\psi_t\) provide the link between the consumption and income distributions - \(\nu_{it}\) the permanent and \(\varepsilon_{it}\) the transitory shock to income.
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\]

\(\phi_t\) and \(\psi_t\) provide the link between the consumption and income distributions - \(\nu_{it}\) the permanent and \(\varepsilon_{it}\) the transitory shock to income.

For a simple benchmark intertemporal consumption model for consumer of age \(t\), can show

\[
\phi_t = (1 - \pi_{it}) \quad \text{and} \quad \psi_t = (1 - \pi_{it}) \gamma_{Lt}
\]

where

\[
\pi_{it} \approx \frac{\text{Assets}_{it}}{\text{Assets}_{it} + \text{Human Wealth}_{it}}
\]

and \(\gamma_{Lt}\) is the annuity value of a temporary shock to income for an individual aged \(t\) retiring at age \(L\).

[ Easily extend to ARMA processes for income. ]