

Using Consumption Data to Extract Information Sets (Extract II)

by Costas Meghir and Luigi Pistaferri (Extract from Chapter 9 of
Handbook of Labor Economics, Volume 4b, 2011)

James J. Heckman
University of Chicago

Economics 312, Winter 2019



Identifying information set for given insurance configuration

- Can consumption and income data be useful in identifying information set or learn more about the nature of the income process?
- Consider a simple extension of an example used by Browning, Hansen and Heckman (1999).
- Assume that the income process is given by the sum of a random walk ($p_{i,a,t}$), a transitory shock ($\varepsilon_{i,a,t}$) and a measurement error ($m_{i,a,t}$, which may even reflect “superior information,” i.e., information that is observed by the individual but not by an econometrician):

$$y_{i,a,t} = p_{i,a,t} + \varepsilon_{i,a,t} + m_{i,a,t}$$

$$p_{i,a,t} = p_{i,a-1,t-1} + \zeta_{i,a,t}$$

- Written in first differences:

$$\Delta y_{i,a,t} = \zeta_{i,a,t} + \Delta \varepsilon_{i,a,t} + \Delta m_{i,a,t}$$

- One cannot separately identify transitory shocks and measurement error without further information.
- Assume that preferences are quadratic, $\beta(1+r) = 1$ and that the consumer's horizon is infinite for simplicity.

- The change in consumption is given by permanent income equation adapted to the infinite horizon case:

$$\Delta C_{i,a,t} = \zeta_{i,a,t} + \frac{r}{1+r} \varepsilon_{i,a,t} \quad (1)$$

- The component $m_{i,a,t}$ does not enter (1) because consumption does not respond to measurement error in income.
- However, if $m_{i,a,t}$ represented "superior information", then this assumption would have behavioral content: it would be violated if liquidity constraints were binding - and hence $m_{i,a,t}$ would belong to (1).

- Suppose a researcher has access to panel data on consumption and income (a very stringent requirement, as it turns out).
- Then one can use the following covariance restrictions:

$$\begin{aligned} \text{var}(\Delta y_{i,a,t}) &= \sigma_{\zeta}^2 + 2(\sigma_{\varepsilon}^2 + \sigma_m^2) \\ \text{cov}(\Delta y_{i,a,t}, \Delta y_{i,a-1,t-1}) &= -(\sigma_{\varepsilon}^2 + \sigma_m^2) \\ \text{var}(\Delta c_{i,a,t}) &= \sigma_{\zeta}^2 + \left(\frac{r}{1+r}\right)^2 \sigma_{\varepsilon}^2 \end{aligned}$$

- As is clear from the first two moments, σ_{ε}^2 and σ_m^2 cannot be told apart from income data alone (although the variance of permanent shocks can actually be identified - e.g., using $\sigma_{\zeta}^2 = \text{var}(\Delta y_{i,a,t}) + 2\text{cov}(\Delta y_{i,a,t}, \Delta y_{i,a-1,t-1})$, the stationary version of equation above).

- However, the availability of consumption data solves the identification problem.
- In particular, one could identify the variance of transitory shocks from, e.g.

$$\sigma_{\varepsilon}^2 = \left(\frac{r}{1+r} \right)^{-2} [var(\Delta c_{i,a,t}) - var(\Delta y_{i,a,t}) - 2cov(\Delta y_{i,a,t}, \Delta y_{i,a-1,t-1})] \quad (2)$$

- Note also that if one is willing to use the covariance between changes in consumption and changes in income ($cov(\Delta c_{i,a,t}, \Delta y_{i,a,t}) = \sigma_{\zeta}^2 + (\frac{r}{1+r}) \sigma_{\varepsilon}^2$), then there is even an overidentifying restriction that can be used to test the model.