

Information-driven business cycles

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Motivation of the paper

Main idea:

- Propose a methodology for solving and estimating DSGE models with incomplete information.
- relies on solving the aggregate model with full information
- without solving for the disaggregate incomplete information equilibrium

Overview

- Prototypical NK model where agents have incomplete information.
- Introduce informational “wedges” instead of usual expectational errors
 - e.g. partial information for agents i , $\mathcal{J}_{it} \subset \mathcal{I}_t$ (full information): aggregate expectational error solves

$$\epsilon_t = E \left[\cdot \int \mathcal{J}_{it} di \right] - E [\cdot | \mathcal{I}_t]$$

- (aggregate) wedge solves

$$\tau_t = \int E [\cdot | \mathcal{J}_{it}] di - E [\cdot | \mathcal{I}_t]$$

- Paper proposes an estimation methodology to that does not involve solving for incomplete information equilibrium
 - minimal exclusion restrictions for information sets of individual agents
 - mix bootstrapped GMM + simulations + priors on parameters
- Study the impact of wedges on business cycle dynamics

The paper

DSGE models where agents have incomplete information

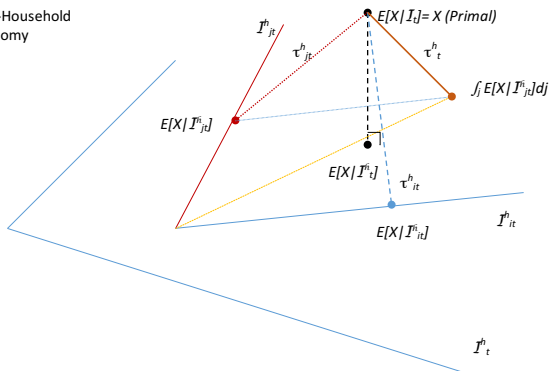
- Agents i of type $a \in \{h, f, b\}$ possess information \mathcal{I}_{it}^a with aggregate $\mathcal{I}_t^a \subset \mathcal{I}_t$
- 3-equation aggregate new Keynesian model derived with expectations conditional on each set of agents' information
 - PC involves $E[\cdot | \mathcal{I}_{it}^h]$, IS involves $E[\cdot | \mathcal{I}_{jt}^f]$, Taylor rule $E[\cdot | \mathcal{I}_t^b]$.
- Related “Primal” economy under complete information $E_t[\cdot] \equiv E[\cdot | \mathcal{I}_t]$
- Informational “wedges” (where complete information $\mathcal{I}_t = \{\mathcal{I}_{it}^h\} \cup \{\mathcal{I}_{jt}^f\} \cup \{\mathcal{I}_t^b\}$.)

$$\begin{bmatrix} \tau_t^h \\ \tau_t^f \\ \tau_t^b \end{bmatrix} \equiv \begin{bmatrix} \int_i E[\lambda' X_t | \mathcal{I}_{it}^h] di \\ \int_j E[\mu' X_t | \mathcal{I}_{jt}^f] dj \\ E[v' X_t | \mathcal{I}_t^b] \end{bmatrix} - \begin{bmatrix} E[\lambda' X_t | \mathcal{I}_t] \\ E[\mu' X_t | \mathcal{I}_t] \\ E[v' X_t | \mathcal{I}_t] \end{bmatrix}$$

- Conditional expectations as projections to understand Theorem 1

Informational wedges and Theorem 1

Two-Household
Economy



Given the Primal eqm X and the individual wedges τ_{jt}^h , candidate individual information sets must be orthogonal to the wedges.

Questions on the Primal/Wedge decomposition

- What is the rationale in using minimum assumptions on agent's information? (robustness?)
i.e. What do wedges represent
 - if postulated $\mathcal{I}_t \not\subseteq$ **actual** full info (news shocks....), does the misspecification appear in the wedges?
 - if postulated individual information $\not\subseteq \mathcal{I}_t^h$: wedges artificially correlate with 'true' info (e.g. historic aggregates) as in your case?
- How are the inclusion restriction imposed? (link with the Wold decomposition?)
- How is the aggregate economy different from complete info + serially correlated errors? (e.g. Milani, Fanelli)

Econometric Methodology

- Propose a generating process for $(a_t, \tau_t^h, \tau_t^f, \tau_t^b)$ and estimate model parameters ψ as

$$\hat{\psi} = \operatorname{argmin}_{\psi} \left(\hat{\Omega} - \Omega(\psi) \right)' V^{-1} \left(\hat{\Omega} - \Omega(\psi) \right)$$

Ω is simulated vectorized autocovariance matrix (up to k lags) of $(y_t, \pi_t, n_t, i_t, a_t)'$,

$\hat{\Omega}$ empirical estimate, V diagonal with bootstrapped variance estimates of $\hat{\Omega}$.

Issues to clarify about econometrics & empirics

■ Econometrics

- why V diagonal? artificially imposes identification? (Gospodinov et al, 2016)
- $a_t \equiv y_t - \alpha n_t$ induces reduced rank in (nondiagonal) V ?
- heteroscedasticity? (Bloom, Jurado et al. 2015...)
- What if *wedges* are persistent?
e.g. risk preference and/or habits? or simply aggregation of AR(2), see Granger (1980)
→ you may replace orthogonality restriction on τ_t with that on the corresponding innovations
- Ω obtained by simulation → what distribution for the innovations to a_t, τ_t

■ Empirics

- HP filtered: assumes individual agents have perfect foresight of aggregate? (as opposed to e.g. Lorenzoni (2009), Blanchard et al. 2013).
- Measure of fit + estimation uncertainty?
- past aggregates are not part of agents' information set?
- impact of weak identification:
→ clarify the link between rank of V , # of orthogonality conditions, weak role of wedges in some IRFs, and the bivariate VMA structure for expectational shocks u_t

A PROMISING FIRST DRAFT!