Risk Shocks

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Finding

- Countercyclical fluctuations in the cross-sectional variance of a technology shock, when inserted into an otherwise standard macro model, can account for a substantial portion of economic fluctuations.
 - Complements empirical findings of Bloom (2009) and Kehrig (2011) suggesting greater cross-sectional dispersion in recessions.
 - Complements theory findings of Bloom (2009) and Bloom, Floetotto and Jaimovich (2009) which describe another way that increased cross-sectional dispersion can generate business cycles.
- 'Otherwise standard model':
 - A DSGE model, as in Christiano-Eichenbaum-Evans or Smets-Wouters
 - Financial frictions along the line suggested by BGG.

Outline

- Rough description of the model.
- Summary of Bayesian estimation of the model.
- Explanation of the basic finding of the analysis.















Economic Impact of Risk Shock



Five Adjustments to Standard DSGE Model for CSV Financial Frictions

- Drop: household intertemporal equation for capital.
- Add: equations that characterize the loan contract
 - Zero profit condition for suppliers of funds.
 - Efficiency condition associated with entrepreneurial choice of contract.
- Add: Law of motion for entrepreneurial net worth (source of accelerator and Fisher debt-deflation effects).
- Introduce: bankruptcy costs in the resource constraint.

Risk Shocks

• We assume risk has a first order autoregressive representation:

iid, univariate innovation to
$$\hat{\sigma}_t = \rho_1 \hat{\sigma}_{t-1} + \underbrace{u_t}_{t}$$

• We assume that agents receive early information about movements in the innovation ('news').

Risk Shock and News

- Assume iid, univariate innovation to $\hat{\sigma}_t$ $\hat{\sigma}_t = \rho_1 \hat{\sigma}_{t-1} + \hat{u}_t$
- Agents have advance information about pieces of u_t 'signals' or 'news'

$$u_t = \xi_t^0 + \xi_{t-1}^1 + \ldots + \xi_{t-8}^8$$

$$\xi_{t-i}^i \sim \text{iid}, E(\xi_{t-i}^i)^2 = \sigma_i^2$$

 ξ_{t-i}^i ~piece of u_t observed at time t-i

DSGE Baseline 4493.85 DSGE without Signals DSGE with Signals on Equity Shock (γ) and No Signals on Risk Shock (σ)

Marginal likelihood

4098.43

4422.46

 Marginal likelihood

 DSGE Baseline
 4493.85

 DSGE without Signals
 4098.43

 DSGE with Signals on Monetary Policy and No
 4427.59

Marginal likelihood **DSGE** Baseline 4493.85 DSGE without Signals 4098.43 DSGE with Signals on Exogenous Spending Shock (g) and No Signals on Risk Shock (σ) 4096 62

Marginal likelihood DSGE Baseline 4493.85 DSGE without Signals 4098.43 DSGE with Signals on Technology Shocks and No Signals on Risk Shock (σ) 4334 47

Monetary Policy

- Nominal rate of interest function of:
 - Anticipated level of inflation.
 - Slowly moving inflation target.
 - Deviation of output growth from ss path.
 - Monetary policy shock.

12 Shocks

- Trend stationary and unit root technology shock.
- Marginal Efficiency of investment shock (perturbs capital accumulation equation)

$$\bar{K}_{t+1} = (1 - \delta)\bar{K}_t + G(\zeta_{i,t}, I_t, I_{t-1})$$

- Monetary policy shock.
- Equity shock.
- Risk shock.
- 6 other shocks.

Estimation

- Use standard macro data: consumption, investment, employment, inflation, GDP, price of investment goods, wages, Federal Funds Rate.
- Also some financial variables: BAA 10 yr Tbond spreads, value of DOW, credit to nonfinancial business, 10 yr Tbond – Funds rate.
- Data: 1985Q1-2010Q2

Results

- Risk shock most important shock for business cycles.
- Quantitative measures of importance.
- Why are they important?
- What shock do they displace, and why?



Notes: The grey solid line represents the (two-sided) fitted data. The dotted black line is the model simulations.

Why Risk Shock is so Important

- A. Our econometric estimator 'thinks' risk spread ~ risk shock.
- B. In the data: the risk spread is strongly negatively correlated with output.
- C. In the model: bad risk shock generates a response that resembles a recession
- A+B+C suggests risk shock important.



Correlation (risk spread(t),output(t-j)), HP filtered data, 95% Confidence Interval

Notes: Risk spread is measured by the difference between the yield on the lowest rated corporate bond (Baa) and the highest rated corporate bond (Aaa). Bond data were obtained from the St. Louis Fed website. GDP data were obtained from Balke and Gordon (1986). Filtered output data were scaled so that their standard deviation coincide with that of the spread data.

Surprising, from RBC perspective



Looks like a business cycle



Quantity of capital

- Marginal efficiency of investment shock can account well for the surge in investment and output in the 1990s, as long as the stock market is not included in the analysis.
- When the stock market is included, then explanatory power shifts to financial market shocks.
- When we drop 'financial data' slope of term structure, interest rate spread, stock market, credit growth:
 - Hard to differentiate risk shock view from marginal efficiency of investment view.

Conclusion

- Incorporating financial frictions and financial data changes inference about the sources of shocks:
 - risk shock.
- Interesting to explore mechanisms that make risk shock endogenous.
- Models with financial frictions can be used to ask interesting policy questions:
 - When there is an increase in risk spreads, how should monetary policy respond?

Comparison with Bloom (2009)

• Return of entrepreneur *i* at time *t*.

 $r_{i,t+1} = \log(1 + R_{t+1}^k) + \log \omega_{it}$, $\log \omega_{it} \sim \text{Normal with variance, } \sigma_t$

• Go to CRSP data set, 1985 – 2010

$$\hat{\sigma}_{t} = \left(\frac{1}{N_{t}}\sum_{i=0}^{N_{t}} [r_{it} - \log(1 + R_{t}^{k})]^{2}\right)^{1/2}$$
CRSP measure of uncertainty
$$1 + R_{t}^{k} = \frac{1}{N_{t}}\sum_{i=0}^{N_{t}} \exp(r_{it})$$
log, idiosyncratic shock





Smoothed estimate of the risk shock