Sentiment Booms Go Wrong

Marco Brianti

Vito Cormun

March 2019

Dissertation Workshop, Boston College

Two long Traditions in Macroeconomics

- Changes in expectation as an important driver of economic fluctuations
 - Incentives to anticipate future economic conditions.
 - ⇒ Pigou (1927); Keynes (1936)
 - ⇒ Beaudry and Portier (2004, 2006)

- 2 Endogenous cycle: expansions lead recessions
 - Economic fluctuations are driven by internal forces which favor recurrent periods of boom and bust.
 - ⇒ von Mises (1940); Beaudry, Galizia, and Portier (2018, 2019)
 - ⇒ Minsky (1977); Bordalo, Gennaioli, Shleifer (2018)

This paper

- We empirically estimate sentiment shocks and evaluate their effects on aggregate U.S. variables
 - We define sentiment shocks as changes in expectations uncorrelated with fundamentals
 - Sentiment shocks trigger boom-and-bust dynamics on most macroeconomic variables
 - Sentiments explain up to 40% of output

- We write a general equilibrium model that rationalizes our empirical findings
 - •
 - •

Contributions

- We use Instrumental Variable Local Projection (IV-LP) to estimate sentiments shocks
 - Previous literature estimates DSGE models or employ SVAR
 - ⇒ Milani, 2011; Levchenko and Pandalai-Nayar, 2018
- Uncover new dynamics in response to sentiment shocks
 - Informative for the literature on sentiments
 - ⇒ Angeletos and La'O, 2013; Angeletos et al. (2018)
- New supportive evidence for the literature on credit cycles
 - We proposed structural evidence in favor credit booms with negative macroeconomic consequences
 - ⇒ Lopez-Salido, Stein, and Zakrajsek (2017)
- (Ideally) Theory that displays boom-and-bust dynamics conditional on a specific type of shock
 - Hard to get shock specific boom and busts
 - ⇒ Beaudry, Galizia, and Portier (2019)

- 1. Empirical Strategy
- 2. Empirical Results
- 3. Test
- 4. Model
- 5. Conclusions

Econometric Strategy

A 2-step procedure:

9 Build an **instrument** Z_t correlated with changes in expectations and orthogonal to fundamentals.

Estimate dynamic responses of macroeconomic variables using IV-LP.

Data Treatment on Expectations

Quarterly data from 1982 to 2018 of forecasts on macroeconomic variables, X_t^s , made by **Survey of Professional Forecasters**

Define,

- $E_t^i(X_{t+k}^s)$ as the expectation on X_{t+k}^s given the information set at time t released by professional forecaster i
- $E_t(X_{t+k}^s)$ as the sample mean across i of $E_t^i(X_{t+k}^s)$
- $E_t(\hat{x}_{t+k}^s) = E_t(X_{t+k}^s)/E_t(X_t^s) 1$ as the expectation of the growth rate of X^s from t to t+k given information set t
- $R_{t,k}^s = E_t(\hat{x}_{t+k}^s) E_{t-1}(\hat{x}_{t+k}^s)$ as the revision on expectations from t-1 to t of the growth rate of X^s from t to t+k
- R_t^k is the first principal component of $R_{t,k}^s$

IV-LP Estimator

Dynamic response of endogenous variable Y_{t+h} to R_t is

$$Y_{t+h} = \Theta_h^Y R_t + u_{h,t+h}^Y \tag{1}$$

Because R_t is endogenous, OLS estimation of 1 is not valid. Eq. 1 can be estimated by IV if Z_t satisfies the following conditions

- $E(\varepsilon_{2:N,t}Z_t)=0$ (contemporaneous exogeneity)
- $E(\varepsilon_{1:N,t+j}Z_t)=0$ for $j\neq 0$ (lead-lag exogeneity)

Given the validity of previous conditions, a consistent estimator for Θ_h^Y is defined as

$$\widehat{\Theta}_{h}^{Y} = \frac{\sum_{t=0}^{T-h} Y_{t+h} Z_{t}}{\sum_{t=0}^{T-h} R_{t} Z_{t}}$$

Instrument Z_t

We estimate instrument Z_t as the unpredictable component of R_t orthogonal to fundamentals,

$$R_t = c + B(L)\Delta TFP_t + \delta W_t + Z_t$$

where,

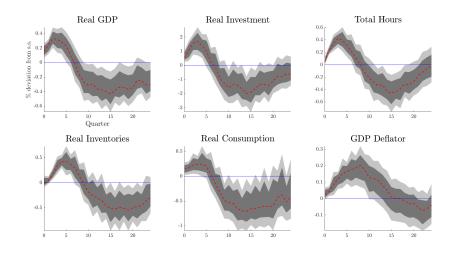
- \bullet ΔTFP is the first difference of utilization-adjusted TFP
- \bullet W_t represents a series of controls
 - Lagged principal components
 - Other structural shocks

Importantly, R-Squared are relatively small (30%-50%)

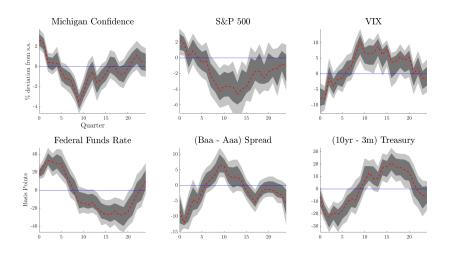
- ⇒ A large part of SPF expectations is unrelated to fundamentals
- $\Rightarrow Z_t$ is a relevant instrument

- 1. Empirical Strategy
- 2. Empirical Results
- 3. Test
- 4. Model
- 5. Conclusions

Impulse Responses (I)



Impulse Responses (I)



Variance Decomposition

	Impact	1 Year	2 Years	5 Years
Real GDP	9.38%	19.81%	16.50%	37.72%
Real Investment Total Hours	4.95%	19.69%	15.06%	35.96%
Total Hours	1.50%	23.38%	14.87%	25.21%
Real Consumption	4.51%	6.70%	5.50%	32.21%

Takeaways

- Sentiment shocks generate cycles of 6 to 7 years in both real and financial variables.
- Sentiments account for the bulk of fluctuations at Business Cycle frequency.
- Technology? Financial variables?

Robustness Checks

- Detrending techniques: first difference, linear, quadratic, Hodrick-Prescott and Band-pass.
- Bivariate VAR(10).
- Choice of lags and controls such as news shocks.
- Use data from Michigan Consumer Survey as a measure of expectations.

- 1. Empirical Strategy
- 2. Empirical Results
- 3. **Test**
- 4. Model
- 5. Conclusions

Test

- 1. Empirical Strategy
- 2. Empirical Results
- 3. Test
- 4. Model
- 5. Conclusions

Model

- 1. Empirical Strategy
- 2. Empirical Results
- 3. Test
- 4. Model
- 5. Conclusions

Conclusions

- 1. Empirical Strategy
- 2. Empirical Results
- 3. Test
- 4. Model
- 5. Conclusions
- 6. Appendix

Technical Details on Empirical Strategy

- Forecast horizon k from SPF data is either 2 or 3
- Forecasted variables X^s are real GDP, nominal GDP, real consumption, real investment, and industrial production
- If Y_t is non-stationary,
 - Detrend Y_t with low-frequency filters
 - Take the first difference of Y_t and $\Gamma_h^Y = \sum_{i=0}^h \Theta_h^Y$ is the response of Y_{t+h}
- Bootstrap method is from Kilian and Kim (2011)

Bootstrapping Technique

- $\textbf{② Create $\Lambda_{h,t,1}^Y$ of the same length of T of $\Lambda_{h,t}^Y$ where $\Lambda_{h,t,1}^Y$ is formed by randomly extracted blocks of length I from $\Lambda_{h,t}^Y$ }$
- **Solution** Estimate $\Theta_{h,1}^Y$ from $\Lambda_{h,t,1}^Y$ using IV-LP estimator
- **1** Redo first 3 steps B=2000 times and get $\Theta_{h,b}^{Y}$ where $b=1,\ldots,B$
- **Select** confidence bands of $\Theta_{h,b}^{Y}$ across b for all h

Impulse Responses to a Surprise Productivity Shocks

