# Optimism Shock on the Australian Business Cycle

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## Introduction

In this research project, I...

- R package bsvarSIGNs
  - hopefully on CRAN next semester!
- Implements Arias, Rubio-Ramírez, and Waggoner (2018)
- Replicates US data
- Extends to Australian data and more

## bsvarSIGNs

- Bayesian structural vector autoregression (bsvar)
- Simultaneous equations + autoregression
- Identified with many restrictions
  - zero restrictions
  - sign restrictions
  - narrative restrictions

## Question

Does optimism shock drive business cycles?

- ► Theory suggests **yes** (Angeletos, Collard, and Dellas 2018)
- ▶ Optimistic → spend more and work harder → economic boom!
- What does the data say?

## Optimism shock

- Positively affects stock prices
- Zero impact on productivity

 $u_t$  reduced-form errors,  $\varepsilon_t$  structural shocks, \* = no restrictions

# Penalty function approach

Beaudry, Nam, and Wang (2011)

- Studies optimism shock with penalty function approach (PFA).
- Concludes optimism shock drives business cycle
- Since significant boom in consumption and hours worked.

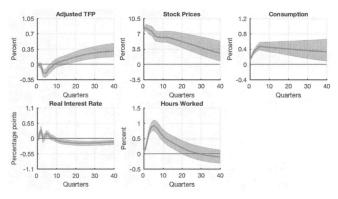


Figure 1: US optimism shock impulse responses with PFA

# Original paper

Arias, Rubio-Ramírez, and Waggoner (2018)

- Proves PFA imposes additional restrictions
- Proposed an importance sampler that gives true solutions
- Concludes optimism shock does not drive business cycle

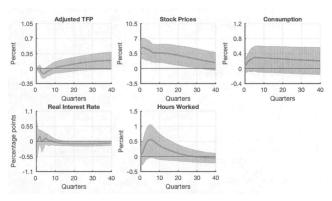


Figure 2: US optimism shock impulse responses with importance sampler

# Importance sampler algorithm

Orthogonal reduced-form parameterization:

$$y'_t = x'_t B + e'_t Q' \operatorname{chol}(\Sigma, \operatorname{upper})$$

- 1. Sample  $(B, \Sigma, Q)$  conditional on the zero restrictions.
- 2. If the sign restrictions are satisfied, keep the draw and compute an importance weight, otherwise discard.
- 3. Repeat steps 1-2 until the desired number of samples is obtained.
- 4. Resample with replacement using the importance weights.

# Replication

#### Same US data, different code

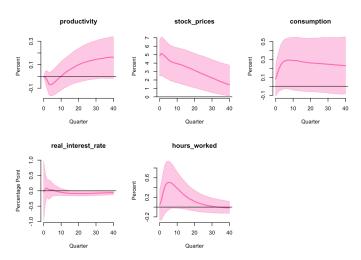


Figure 3: US optimism shock impulse responses using bsvarSIGNs

## Extension 1: Australian data

## No significant impact

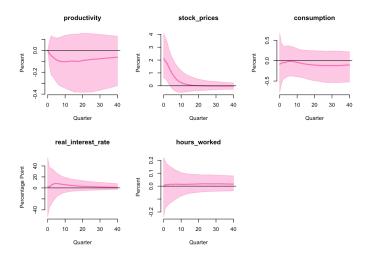


Figure 4: Australian optimism shock impulse responses using bsvarSIGNs

## Extension 2: narrative restriction

Antolín-Díaz and Rubio-Ramírez (2018)

- Algorithm to impose narrative restrictions
- ▶ Sign of structural shocks
- **...**

Assume additionally that the optimism shock

is **negative** when Covid-19 hits Australia in 2020 Q1

$$\varepsilon_{2020Q1}^{\mathrm{optimism}} < 0$$

## Extension 2: Covid-19

## A pessimism shock

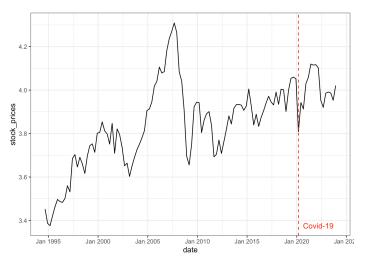


Figure 5: ASX 200 index

# Extension 2: pseudo proof

(To my knowledge) no paper has combined the two algorithms<sup>1</sup>

Both papers use importance sampler

- ▶ Suppose the importance weights are  $w_1$  and  $w_2$
- ► To **combine** the two algorithms, resample with weights *w*
- Where  $w = w_1 \times w_2$

A more rigorous proof (with no guarantee of correctness) is available here

<sup>&</sup>lt;sup>1</sup>Arias, Rubio-Ramírez, and Waggoner (2018) and Antolín-Díaz and Rubio-Ramírez (2018)

# **Extension 2: history**

Historical values of optimism shock

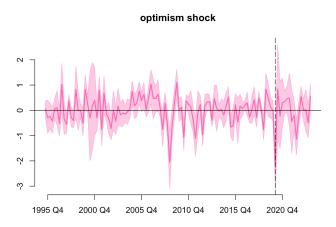


Figure 6: Australian optimism shock

## Extension 2: result

#### Still no significant impact

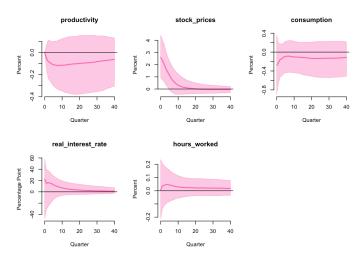


Figure 7: Australian optimism shock impulse responses with narrative restriction

## Summary

- Optimism shock does not drive business cycle
- Result holds for both US and Australia
- ▶ Result is robust to restriction on Covid-19

# Appendix: simulation study

Suppose the true structural model is

$$\begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} y_{1,t} \\ y_{2,t} \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} y_{1,t-1} \\ y_{2,t-1} \end{bmatrix} + \begin{bmatrix} \varepsilon_{1,t} \\ \varepsilon_{2,t} \end{bmatrix}, \begin{bmatrix} \varepsilon_{1,t} \\ \varepsilon_{2,t} \end{bmatrix} \sim \mathcal{N}(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix})$$

equivalently, we can simulate 1,000 observations from the reduced-form

$$\begin{bmatrix} y_{1,t} \\ y_{2,t} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} y_{1,t-1} \\ y_{2,t-1} \end{bmatrix} + \begin{bmatrix} u_{1,t} \\ u_{2,t} \end{bmatrix}, \begin{bmatrix} u_{1,t} \\ u_{2,t} \end{bmatrix} \sim \mathcal{N}(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix})$$

Putting zero and sign restrictions on the inverse of the structural matrix

$$\begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix}^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \Rightarrow \begin{bmatrix} 0 & + \\ + & + \end{bmatrix}$$

Posterior mean of 1,000 draws of the structural matrix is

## References

- Angeletos, George-Marios, Fabrice Collard, and Harris Dellas. 2018. "Quantifying Confidence." *Econometrica* 86 (5): 1689–1726.
- Antolín-Díaz, Juan, and Juan F Rubio-Ramírez. 2018. "Narrative Sign Restrictions for SVARs." *American Economic Review* 108 (10): 2802–29.
- Arias, Jonas E, Juan F Rubio-Ramírez, and Daniel F Waggoner. 2018. "Inference Based on Structural Vector Autoregressions Identified with Sign and Zero Restrictions: Theory and Applications." *Econometrica* 86 (2): 685–720.
- Beaudry, Paul, Deokwoo Nam, and Jian Wang. 2011. "Do Mood Swings Drive Business Cycles and Is It Rational?" National Bureau of Economic Research.