# FINANCIAL CONDITIONS AND THE

# **BUSINESS CYCLE**

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#### MOTIVATION

"Business Cycle Anatomy" (2020) empirically identified a *single* business cycle shock that moved

unemployment, output, investment, hours worked, and consumption,

but the business cycle shock did **not** relate to

inflation or TFP.

The authors suggest a confidence shock could lead to this result.

We show shocks to financial conditions could also lead to this result.

### PREVIEW OF RESULTS

Using volatility financial conditions index (VFCI) to measure financial conditions, we find...

- Targeting VFCI generates same business cycle shock
- Including VFCI links inflation to the business cycle

### **VAR SETUP**

A **SVAR**(p) model for a vector of variables,  $x_t$ ,

$$B_0x_t = B_1x_{t-1} + \cdots + B_px_{t-p} + \epsilon_t$$

Empirically, only the  $A_i$  matrices and reduced form residuals,  $v_t$ , can be estimated

$$x_{t} = \underbrace{B_{0}^{-1}B_{1}}_{A_{1}} x_{t-1} + \dots + \underbrace{B_{0}^{-1}B_{p}}_{A_{p}} x_{t-p} + \underbrace{B_{0}^{-1}\epsilon_{t}}_{v_{t}}$$

This relates  $v_t$  to an unknown linear mapping of stuctural shocks,  $\epsilon_t$ .

$$v_t = B_0^{-1} \epsilon_t$$

**Identification problem:** determining the  $B_0$  matrix.

# MAX FORECAST ERROR VARIANCE IDENTIFICATION

- 1. Pick one variable to target from  $x_t$  (i.e. unemployment,  $x_t^{(u)}$ )
- 2. Compute the forecast error for target horizon, h

$$F_{t+h} = \underbrace{x_{t+h}^{(u)}}_{realization} - \underbrace{x_{t+h}^{(u)}}_{prediction}$$

3. Choose vector  $B_0^{(u)}$  to maximize the variance of  $F_{t+h}$ 

$$\max_{B_0^{(u)}} Var[F_{t+h}]$$

4. This will identify **one** shock

$$\epsilon_t^u = B_0^{(u)} \hat{\mathbf{v}}_t$$

• Can instead calculate *F* for frequency range  $\{\omega^-, \omega^+\}$ 

# **VOLATILITY FINANCIAL CONDITIONS INDEX (VFCI)**

VFCI can be interpreted as the **price of risk**. Constructed using

- Asset returns
- Forward consumption growth

Can target consumption growth at different forward horizons.

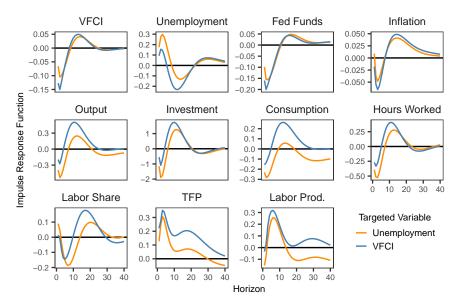
Will use VFCI constructed targeting a horizon of 10 quarters. Why?

- performs well
- lies within the business cycle length

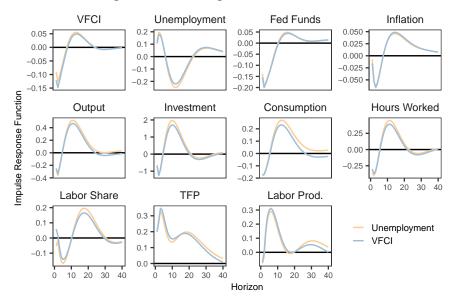
Results are robust to similar horizon targets.



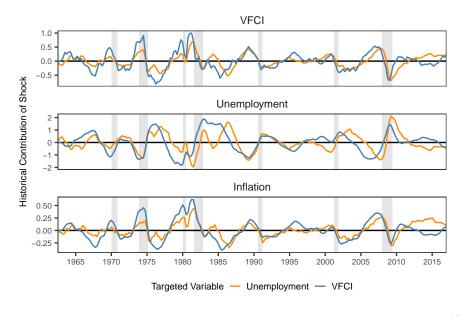
# TARGETING VFCI MATCHES THE BUSINESS CYCLE



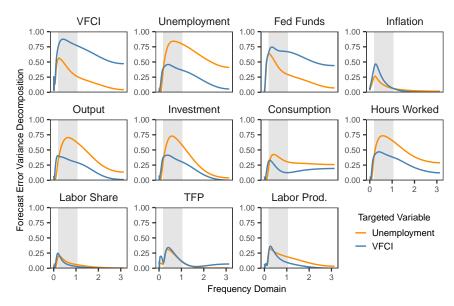
# TARGET FREQ. OF 18:36 Q. ALLOWS PERFECT MATCH



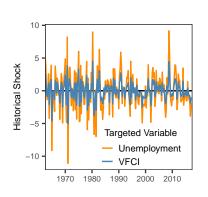
# CONTRIBUTION OF SHOCKS SIMILAR OVER TIME

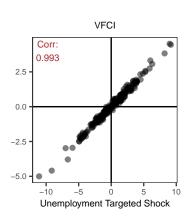


# SHOCKS EXPLAIN LARGE PORTION OF VARIANCE

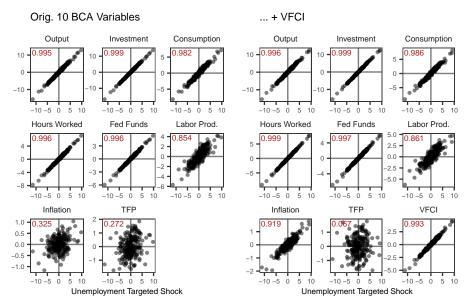


# IDENTIFIED SHOCKS THAT TARGET VFCI AND UNEMPLOYMENT ARE HIGHLY CORRELATED

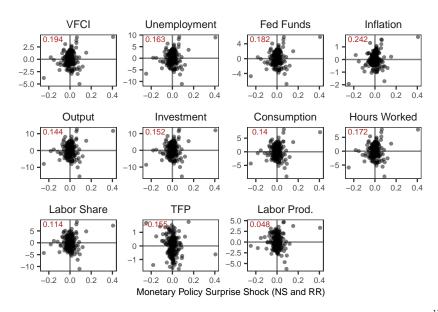




# VFCI LINKS INFLATION TO THE BUSSINESS CYCLE



# SMALL CORR. WITH MONETARY POLICY SURPRISES



# CONCLUSION

#### Results:

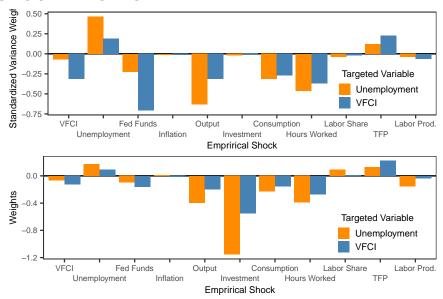
- Targeting VFCI generates same business cycle shock
- Including VFCI links inflation to the business cycle

Defining a business cycle with different frequencies yields even stronger results.

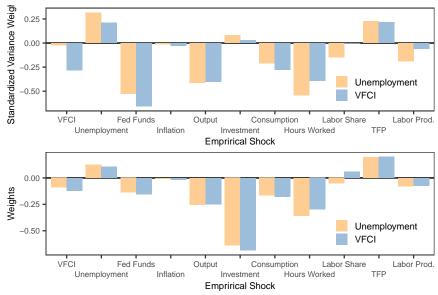
Implies that financial conditions (specifically VFCI) shocks could be driving the business cycle and are at least necessary for understanding it.

# **APPENDIX**

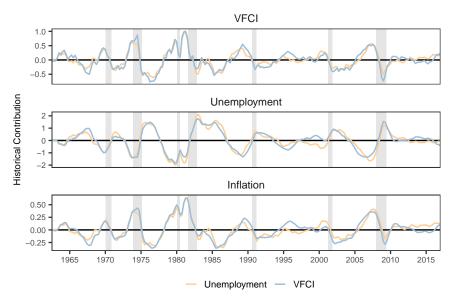
# **SHOCK WEIGHTS**



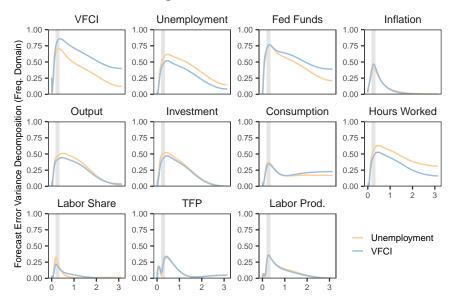
# SHOCK WEIGHTS: 18 TO 36 Q



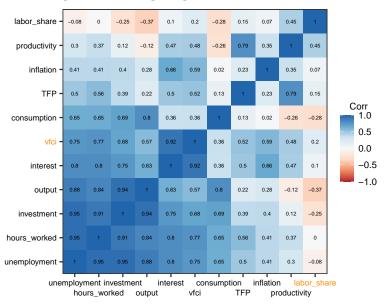
# HISTORICAL CONTRIBUTION: 18 TO 36 Q



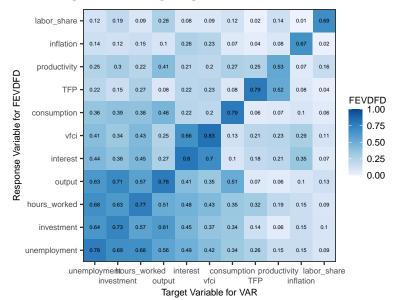
# FEVDFD: 18 TO 36 Q

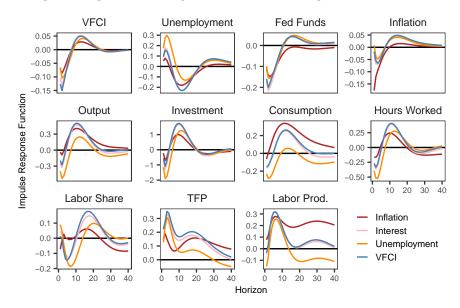


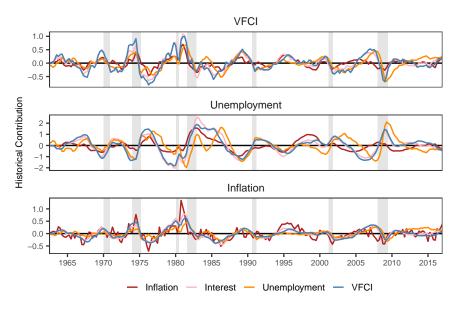
### COMPARING VAR TARGETS

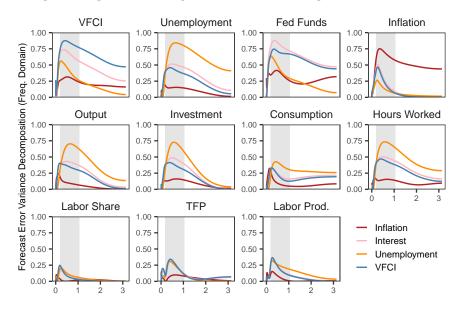


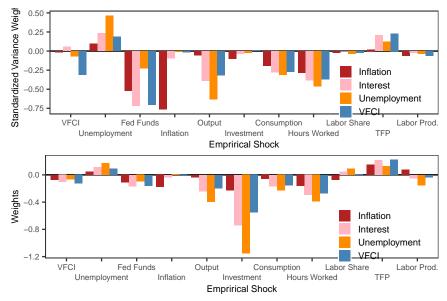
### COMPARING VAR TARGETS











# **VFCI ESTIMATION**

Assuming (1) no arbitrage and (2) a representative agent, a log-linear approximation of the representative agent's FOC relates

- asset prices, R<sub>t</sub>
- future consumption growth,  $\Delta c_{t+1} = \ln c_{t+1} \ln c_t$

This relationship can be estimated empirically:

$$\Delta c_{t+1} = \beta R_t + \varepsilon_t \tag{1}$$

$$\ln \text{Vol}[\varepsilon_t] = \lambda R_t + \upsilon_t \tag{2}$$

VFCI is predicted value from eq. (2). Interpreted as the "price of risk".

$$VFCI_t \equiv ln \widehat{Vol[\epsilon_t]}$$

# **VFCI ESTIMATION**

Assuming (1) no arbitrage and (2) a representative agent, a log-linear approximation of the representative agent's FOC relates

- asset prices, R<sub>t</sub>
- future consumption growth,  $\Delta c_{t+2} = \ln c_{t+2} \ln c_{t+1}$

This relationship can be estimated empirically:

$$\Delta c_{t+1} = \beta R_t + \varepsilon_t \tag{1}$$

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$$\ln \text{Vol}[\varepsilon_t] = \lambda R_t + \upsilon_t \tag{2}$$

VFCI is predicted value from eq. (2). Interpreted as the "price of risk".

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# **VFCI ESTIMATION**

Assuming (1) no arbitrage and (2) a representative agent, a log-linear approximation of the representative agent's FOC relates

- asset prices, R<sub>t</sub>
- future consumption growth,  $\Delta c_{t+1+h} = \ln c_{t+1+h} \ln c_{t+1}$

This relationship can be estimated empirically:

$$x\Delta c_{t+1} = \beta R_t + \varepsilon_t \tag{1}$$

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$$\ln \text{Vol}[\varepsilon_t] = \lambda R_t + \upsilon_t \tag{2}$$

VFCI is predicted value from eq. (2). Interpreted as the "price of risk".

$$VFCI_{t,h} \equiv ln \widehat{Vol[\epsilon_t]}$$

Can also consider longer forward growth horizons,  $h \in [1, \infty)$ .

# VAR SETUP

1. Run the empirical VAR

$$A(L)X_t = u_t$$

with 
$$A(L) \equiv \sum_{\tau=0}^{p} A_{\tau} L^{\tau}$$
,  $A_0 = I$ , and  $\mathbf{E}[u_t u_t'] = \Sigma$ 

2. Orthogonalize the residuals, S = Choleskey decomposition of  $\Sigma$ 

$$u_t = S\epsilon_t$$

with 
$$\mathbf{E}[\epsilon_t \epsilon_t'] = I$$

3. Denote all possible rotations, Q, of structural shocks,  $\epsilon_t$ 

$$S = \tilde{S}Q$$

where Q is any orthonormal (QQ' = I) rotation matrix.

4. This is the identification problem. Which Q to choose?

### MAX VARIANCE IDENTIFICATION

1. Write out the VMA( $\infty$ ) representation of a VAR(p)

$$X_t = B(L)u_t$$

with 
$$B(L) \equiv \sum_{\tau=0}^{p} B_{\tau} L^{\tau}$$
 and  $B(L) = A(L)^{-1}$ 

2. Substitute in rotations of structural shocks,  $u_t = \tilde{S}Q\epsilon_t$ 

$$X_t = C(L)Q\epsilon_t$$

with  $C(L) = B(L)\tilde{S}$  and  $\Gamma(L) = C(L)Q$  stores the IRF.

#### MAX VARIANCE IDENTIFICATION

1. The forecast error variance (FEV) for time horizon of T

$$FEV_T = \sum_{t=0}^{I} \Gamma(t)' \Gamma(t)$$
$$= \sum_{t=0}^{T} Q' C(t)' C(t) Q$$

2. The forecast error variance for a frequency range,  $\{\underline{\omega}, \bar{\omega}\}$ 

$$FEV_{\underline{\omega},\bar{\omega}} = \int_{\underline{\omega}}^{\bar{\omega}} Q'C(e^{-i\omega})'C(e^{-i\omega})Q d\omega$$
$$= Q'\left(\int_{\underline{\omega}}^{\bar{\omega}} C(e^{-i\omega})'C(e^{-i\omega}) d\omega\right)Q$$

. . .