

# The Global Effects of U.S. Monetary Policy

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**7<sup>th</sup> Annual Monetary Economics Workshop**

26 November 2019, Pretoria

# Spillovers from US Monetary Policy?

How Does U.S. MP Affect the Rest of the World?

- ▶ **Demand Channel** – demand-augmenting effect
- ▶ **Exchange Rate Channel** – expenditure-switching effect
- ▶ **Financial Channels** – ‘risk taking’ channel, ‘credit’ channel

Implications:

- ▶ **Dilemma vs Trilemma** / Monetary Sovereignty
- ▶ Global Financial Cycle
- ▶ **Policy spillovers** / Policy Coordinations / Currencies Wars

Important question ... yet **limited empirical literature!**

# What do we know?

Paper	Countries	Identification	CPI	IP/GDP	Short-term rates
Dedola et al. (2018)	18 EME, 18 AE	Sign restriction	—	—	AE: —; EME: mixed
Iacoviello, Navarro (2019)	50 countries	Cholesky	na	—	+
Kim (2001)	G7	Cholesky	insign.	—	+
Canova (2005)	Latin America	Sign restriction	+	mixed	+
Mackowiak (2007)	EME	Sign restriction	+	mixed	+
Miniane, Rogers (2007)	26 countries	Zero restrictions	—*	—	+
Di Giovanni, Shambaugh (2008)	IFS countries	none	na	— for peggers	na
Bluedorn, Bowdler (2011)	AE	Romer and Romer	—*	—*	+
Akinci (2013)	6/10 EME	panel VAR + Cholesky	na	—*	na
Rey (2015)	UK, CAN, SWE, NZ	IV SVAR	mixed	mixed	mixed
Passari, Rey (2015)	UK	IV SVAR	insign.	insign.	insign.
Georgiadis (2016)	EME and AE	Sign restriction	na	—	na
Bhattarai et al. (2017)	15 EME	Cholesky	insign.	—	+
Vicondoa (2019)	13 EME	Cholesky	+	—	+
Miranda-Agricciino, Rey (2019)	US, EU, UK	IV SVAR	—	—	—

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# Challenges to Measuring US MP Global Spillovers

Bernanke, Mundell-Fleming Lecture (2015)

1. Policy actions are a **signal** about **US and global fundamentals**
2. Large heterogeneity across countries both in terms of **cyclical position**, financial markets conditions and '**structural features**'
3. Need of high frequency **data on leverage, risk appetite, capital flows, ...**

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# Challenges to Measuring US MP Global Spillovers

## Our Approach

1. Policy actions are a **signal** about **US and global fundamentals**
  - ⇒ High Frequency Identification for conventional MP shocks ...
  - ⇒ ... robust to signalling effects of MP actions
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  - ⇒ **Rich monthly dataset** of **15 ADV** and **15 EME**
  - ⇒ Efficient big data techniques with **bilateral Large BVARs**
  - ⇒ Global macro and financial aggregates
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  - ⇒ **Bayesian Local Projections** and direct study of asymmetries

## Our Results

- ▶ Strong **nominal effects**
- ▶ Sizeable **real effects**
- ▶ Global response of **capital flows** and **risk appetite**
- ▶ Oil and commodities play an important role in CPI movements
- ▶ Some evidence of **asymmetric effects**

## Related Literature

- ▷ **HF Identification MP Shocks:** Rudebusch (1989), Kuttner (2001), Gürkaynak et al (2005), Gertler and Karadi (2015), Caldara and Herbst (2016), Jarociński and Karadi (2018), Cieslak and Schrimpf (2018), Andrade and Ferroni (2016), Miranda-Agrippino and Ricco (2015)
- ▷ **Global Effects US MP:** Kim (2001), Canova (2005), Mackowiak (2007), Miniane, Rogers (2007), Di Giovanni, Shambaugh (2008), Bluedorn, Bowdler (2011), Akinci (2013), Miranda-Agrippino, Rey (2019), Rey (2015, 2016), Passari, Georgiadis (2016), Bhattacharai et al. (2017), Gerko and Rey (2017), Stravakeva and Tang (2018), Dedola et al. (2018), Iacoviello, Navarro (2019), Vicondoa (2019), Gilchrist et al. (2019), Kalemli-Özcan (2019)
- ▷ **Spillovers through Banks:** Correa and Murry (2009), Cetorelli and Goldberg (2012), Temesvary et al. (2017), Correa et al. (2015), Buch et al. (2018), International Banking Research Network (IBRN)
- ▷ **Dominant Currency:** Gourinchas and Rey (2007), Rey (2015), Bernanke (2015), Casas et al. (2017), Gourinchas et al (2017)

## **The Identification of US MP Shocks**

# The HFI of Monetary Policy Shocks

Kuttner (2001), Sack (2004), Gürkaynak, Sack, Swanson (2005)

- ▶ Interest rates futures for agents' expectations

$$p_t^{(h)} = \mathbb{E}_t (i_{t+h}) + \zeta_t^{(h)}$$

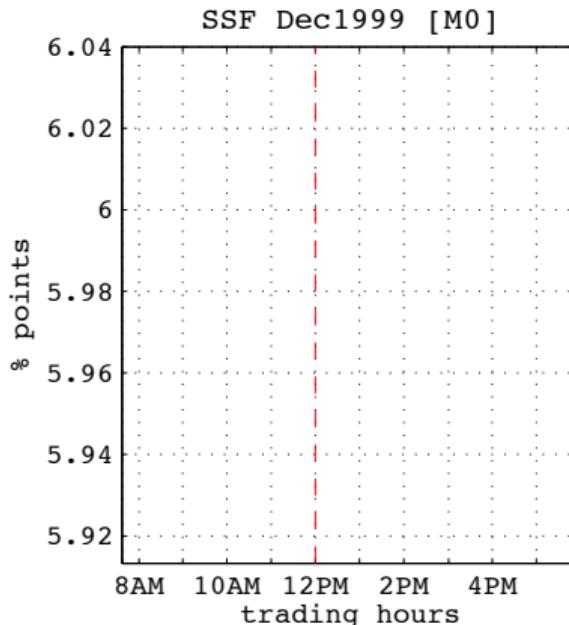
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- ▶ Revision in **30 min window** price around **announcements**



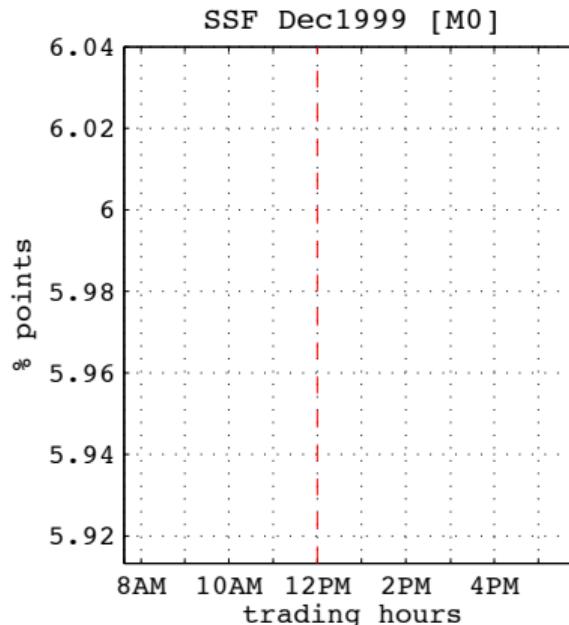
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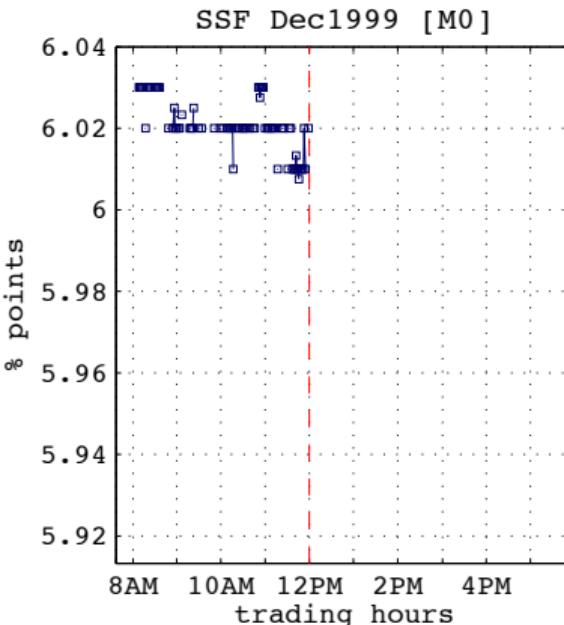
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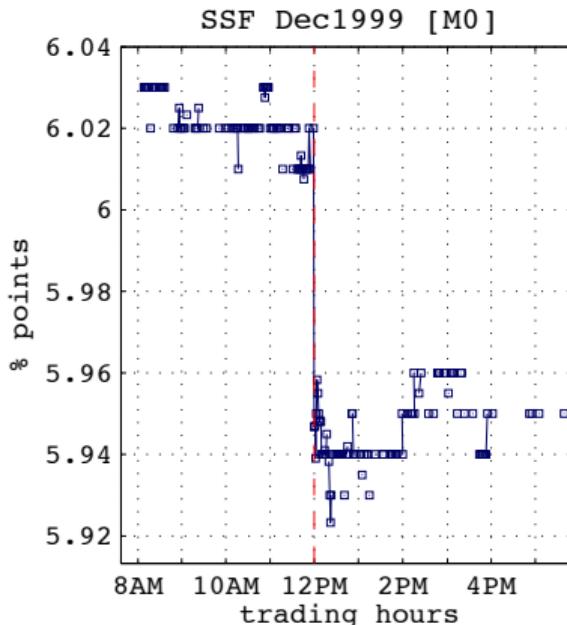
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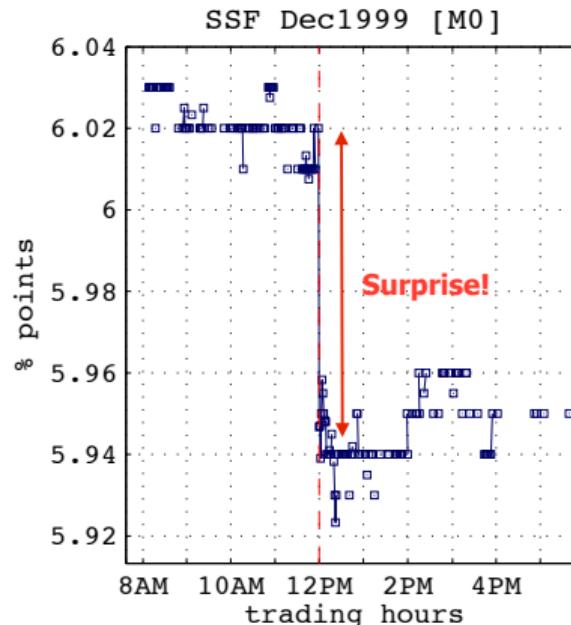
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- ▶ Revision in **30 min window** price around **announcements**  
⇒ **MP surprise**



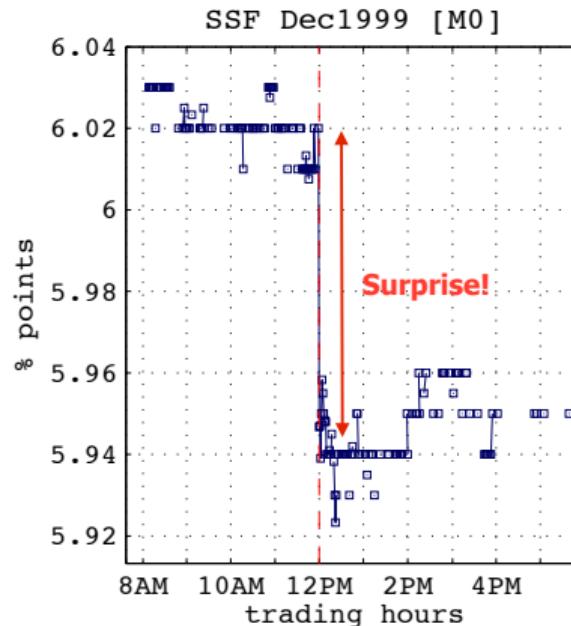
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- ▶ Information or MP shock?



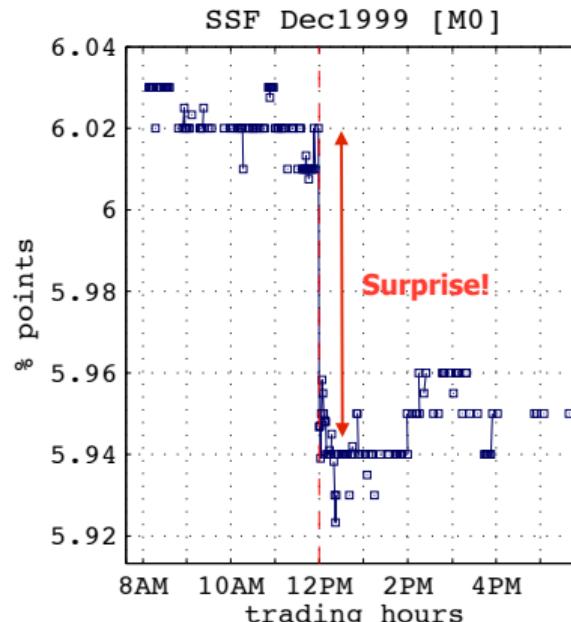
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- ▶ Revision in **30 min window** price around **announcements**
- ▶ Information or MP shock?  
**Interest rate hike**
  - ▶ **MP shock**  
➡ lower output and inflation
  - ▶ **Endogenous reaction** to demand shocks  
➡ higher output and inflation



# An Informationally Robust HFI

Miranda-Agrippino and Ricco (2018)

1. At FOMC frequency  $\implies$  **Signaling Channel**

$$FF4_m = \alpha_0 + \sum_{j=-1}^3 \theta_j F_t^{cb} x_{q+j} + \sum_{j=-1}^2 \vartheta_j [F_t^{cb} x_{q+j} - F_{t-1}^{cb} x_{q+j}] + MPI_m$$

# An Informationally Robust HFI

Miranda-Agrippino and Ricco (2018)

## 2. Monthly aggregation

$$\overline{MPI}_t = \sum_{m \in t} MPI_m$$

# An Informationally Robust HFI

Miranda-Agrippino and Ricco (2018)

3. At monthly frequency  $\implies$  **Slow Absorption of Information**

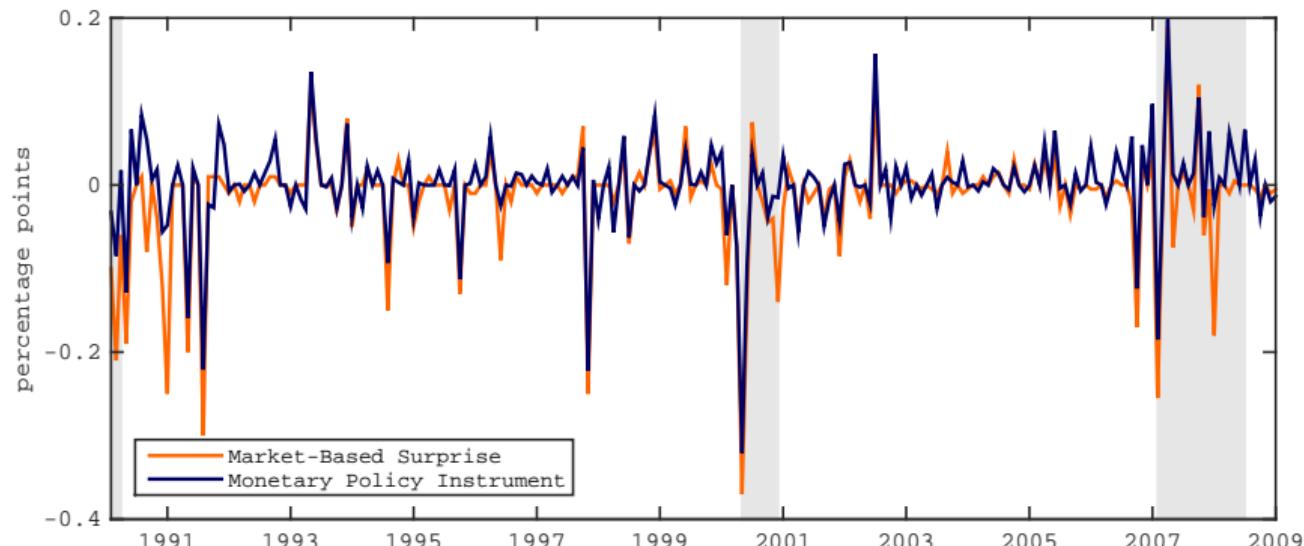
$$\overline{MPI}_t = \phi_0 + \sum_{j=1}^{12} \phi_j \overline{MPI}_{t-j} + \textcolor{red}{MPI}_t$$

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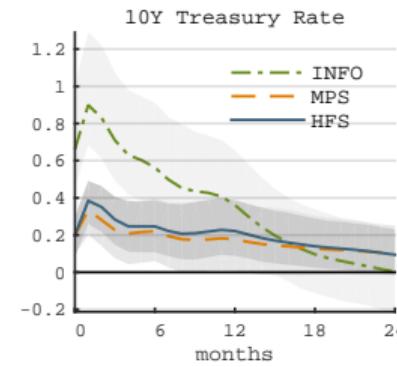
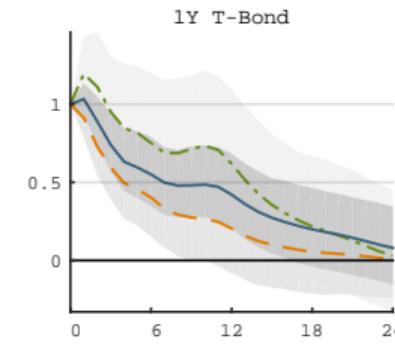
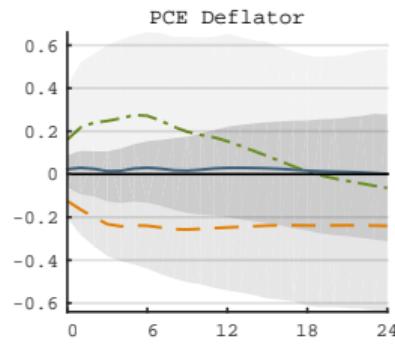
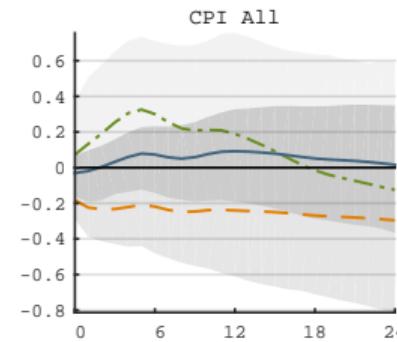
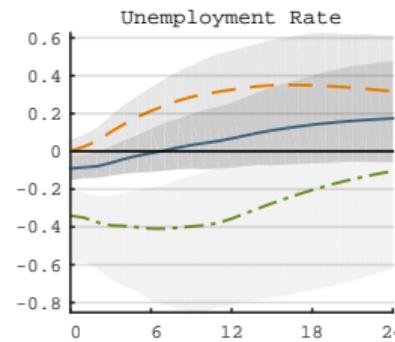
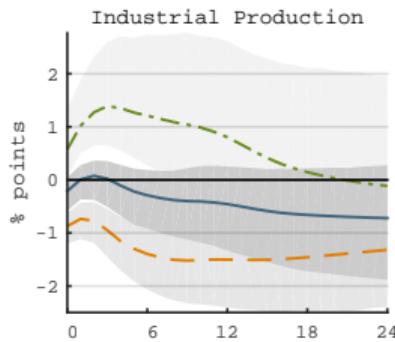
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# Information vs MP Shocks

Miranda-Agrippino and Ricco (2018)



# Large Bayesian VARs

- ▶ Large **Bayesian VAR(12)**
  - ▶ 27 variables in global set
  - ▶ 28 variables in advanced set
  - ▶ 28 variables in emerging set
- ▶ **Estimation Sample:** ~ 1990-2018  
**Instrument:** 1990:2009
- ▶ **IV-SVAR** (Stock and Watson 2012, 2018, Mertens and Ravn 2013)
- ▶ **Standard macro priors** (Normal-Inverse-Wishart)
- ▶ **Optimal prior selection** (Giannone, Lenza, Primiceri, 2015)

# VARs and Local Projections

## VAR-IRFs

$$y_{t+1} = \mathcal{B}y_t + u_{t+1}$$

$$\text{IRF}_h^{VAR} = \mathcal{B}^h A_0^{-1}$$

- ▷ optimal and consistent only if the VAR captures the DGP

# VARs and Local Projections

## VAR-IRFS

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$$\text{IRF}_h^{VAR} = \textcolor{red}{B}^{\textcolor{red}{h}} A_0^{-1}$$

## LP-IRFS

$$y_{t+h} = \widetilde{B}^{(h)} y_t + v_{t+h}$$

$$\text{IRF}_h^{LP} = \widetilde{B}^{(h)} A_0^{-1}$$

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- ▷ robust to misspecification but high estimation uncertainty

# VARs and Local Projections

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## LP-IRFS

$$y_{t+h} = \tilde{\mathbf{B}}^{(h)} y_t + v_{t+h}$$

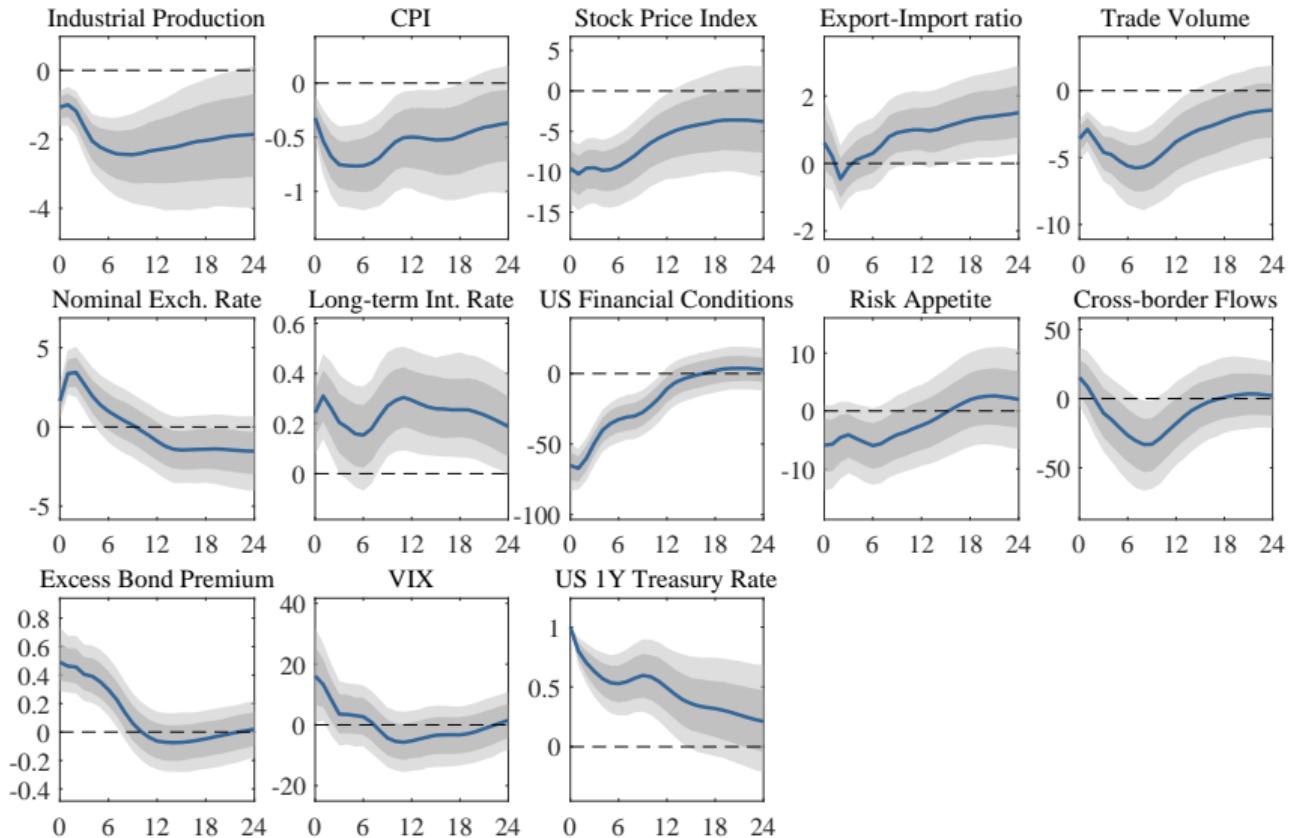
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- ▷ optimal and consistent only if the VAR captures the DGP
- ▷ robust to misspecification but high estimation uncertainty
- ▷ Selecting between the two methods: empirical problem choosing between **bias** and **estimation variance**...
- ▷ **Idea:** Regularise LP with NIW priors centred around VAR coefficients (pre-sample)

$$\tilde{\mathbf{B}}^{(h)} \longleftrightarrow \mathbf{B}^{(\text{VAR}, h)} = \mathbf{B}^h$$

## **Global Spillovers**

# Domestic Effects of U.S. Monetary Policy

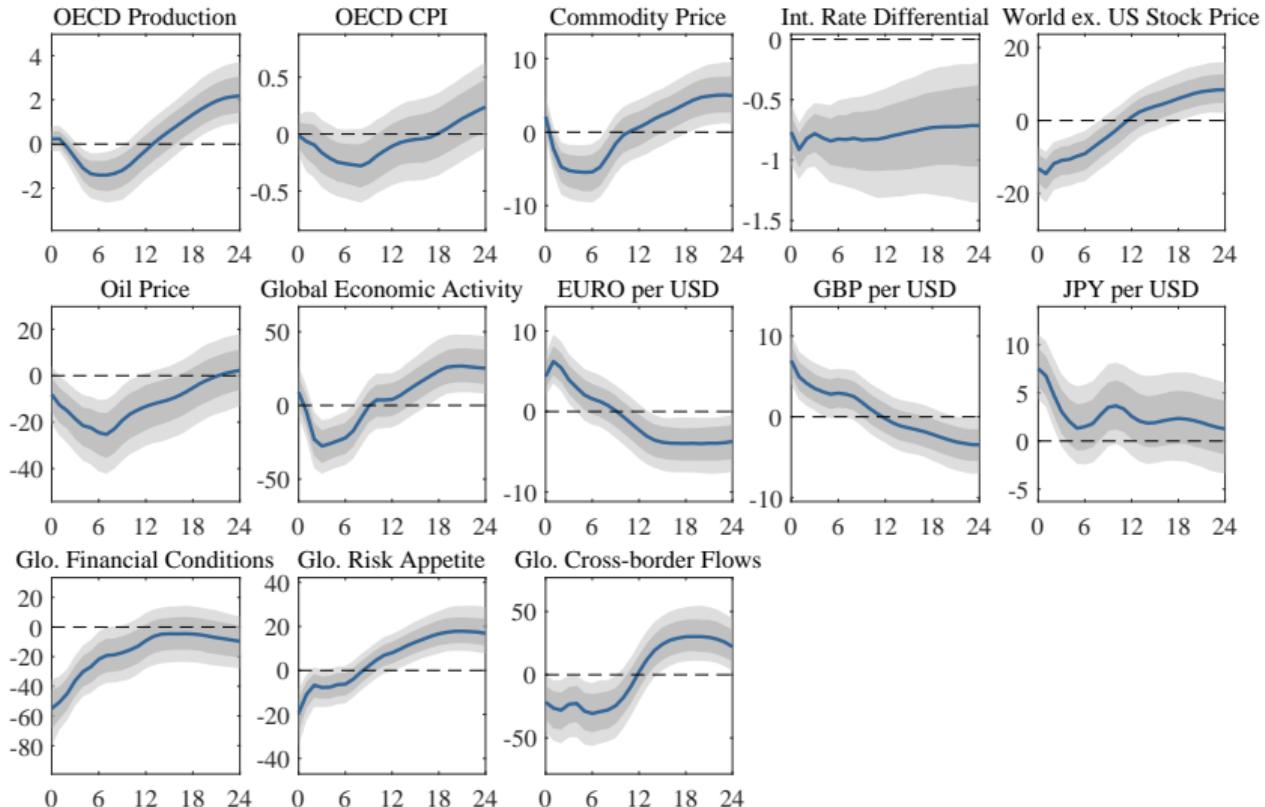


# A Global VAR

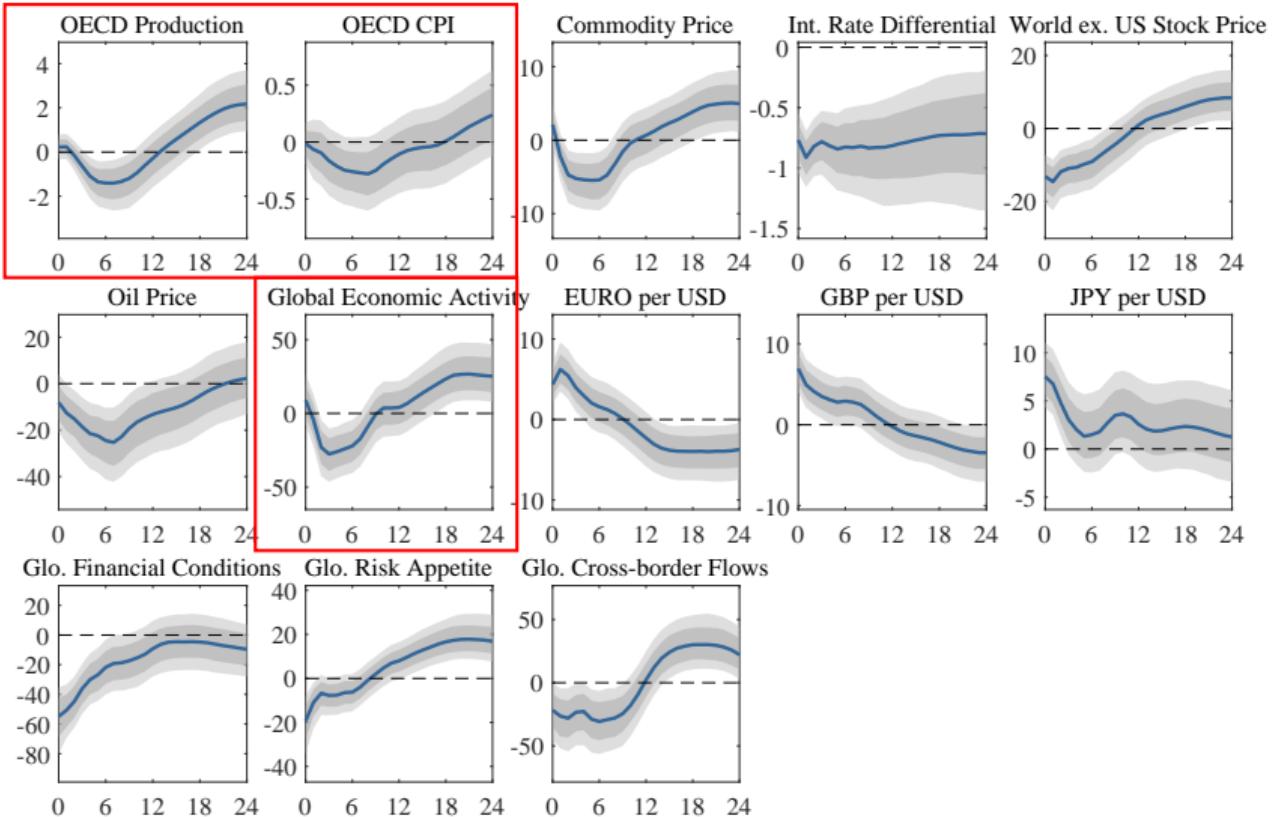
Global set	Log	RW Prior	U.S. set	Log	RW Prior
Industrial Production Index, OECD countries	•	•	US Industrial Production Index	•	•
CPI Index, OECD countries	•	•	US Consumer Price Index	•	•
CRB Commodity Price Index	•	•	US Nominal Stock Price Index	•	•
Short-term Interest Differential			US Export/Import ratio		•
World Stock Price index (excl. North America)	•	•	US Trade Volume	•	•
Global price of Brent Crude (\$/BBL)	•	•	US Nominal Effective Exchange Rate	•	•
Kilian (2019) Global Economic Activity Index			US Financial Conditions Index, CBC	•	
Euro per US Dollar Exchange rate	•		US Risk Appetite, CBC		
British pound per US Dollar Exchange rate	•		US Cross-border Flows Index, CBC	•	
Japanese Yen per US Dollar Exchange rate	•		Gilchrist-Zakrajsek Excess Bond Premium		
Global Financial Conditions Index, CBC	•		CBOE VIX	•	
Global Policy Liquidity Index, CBC	•		US 10-Year Treasury Constant Maturity Rate		
Global Risk Appetite, CBC			US 1-Year Treasury Constant Maturity Rate		
Global Cross-border Flows Index, CBC	•				

Notes: Sample: 1990:1 – 2018:8. OECD variables are weighted by the previous year's GDP (PPP adjusted).

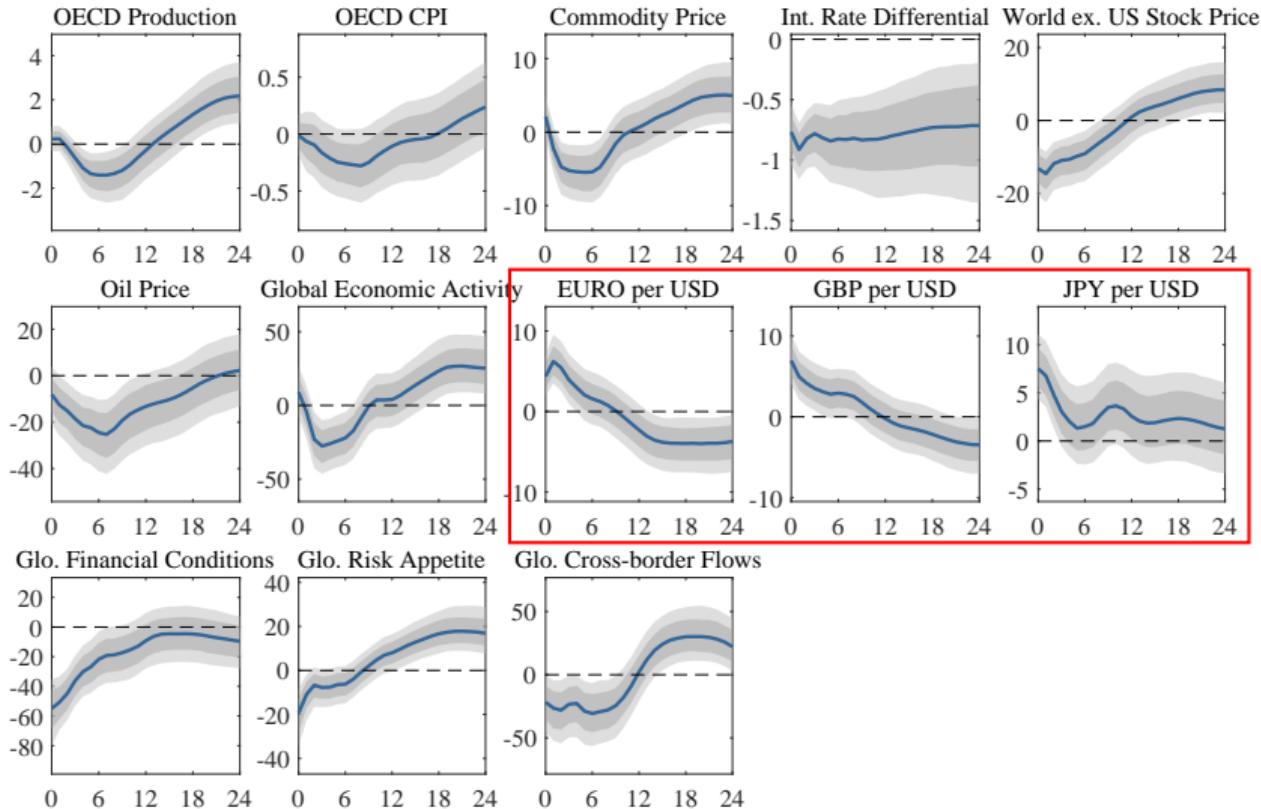
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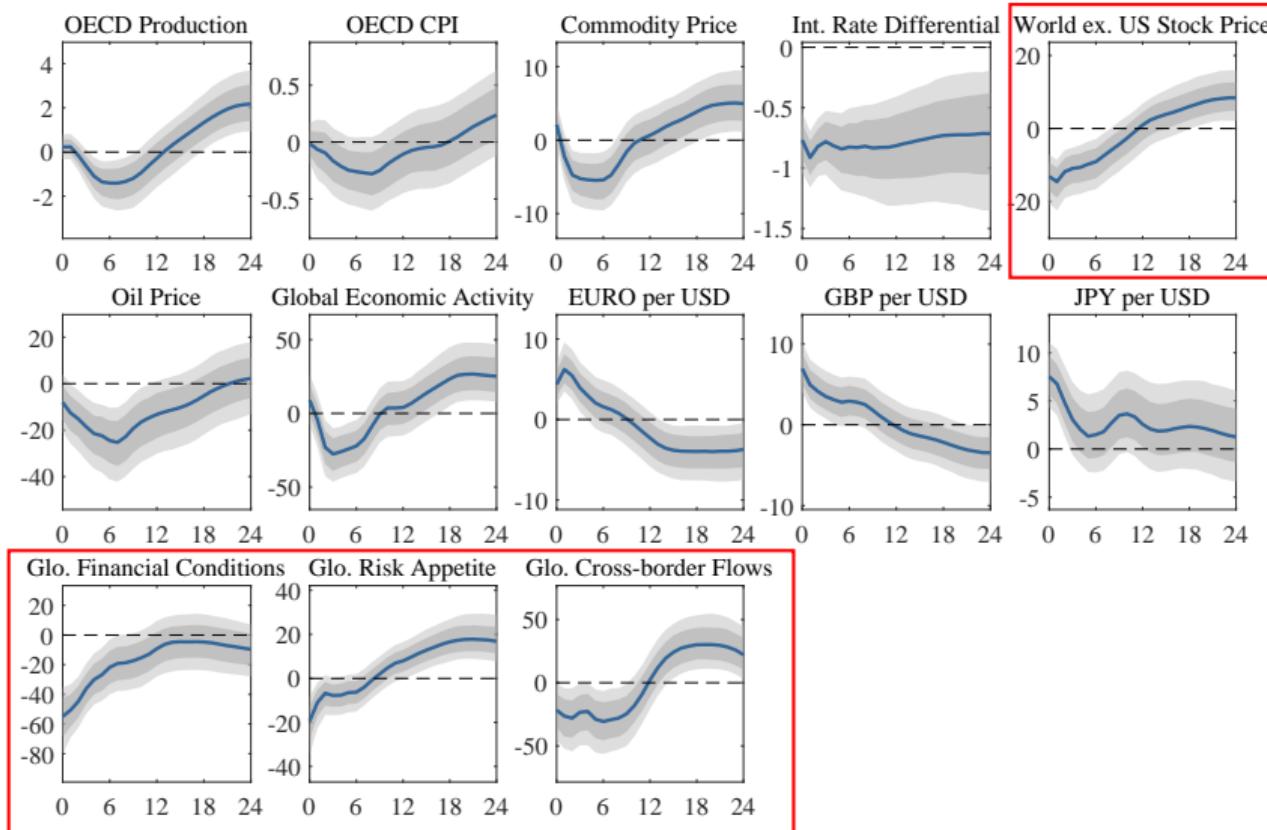
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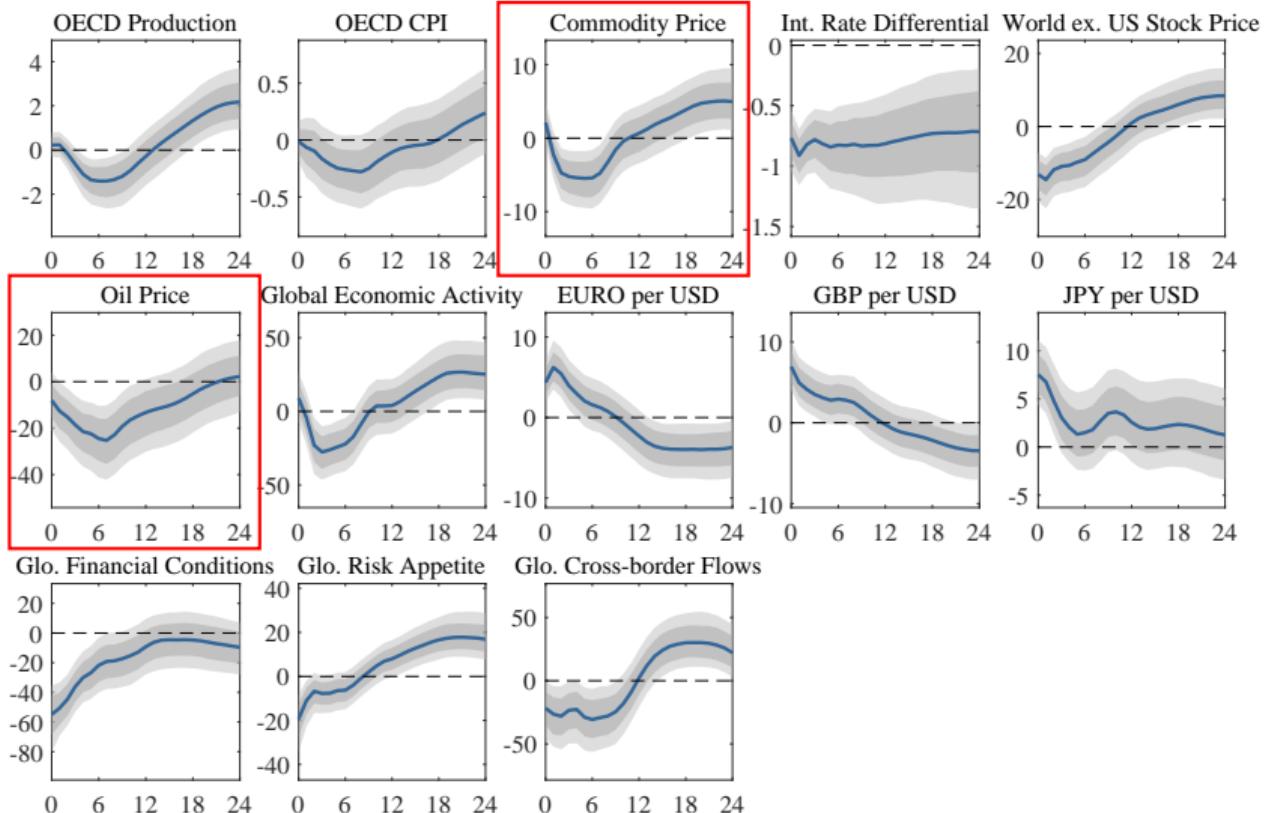
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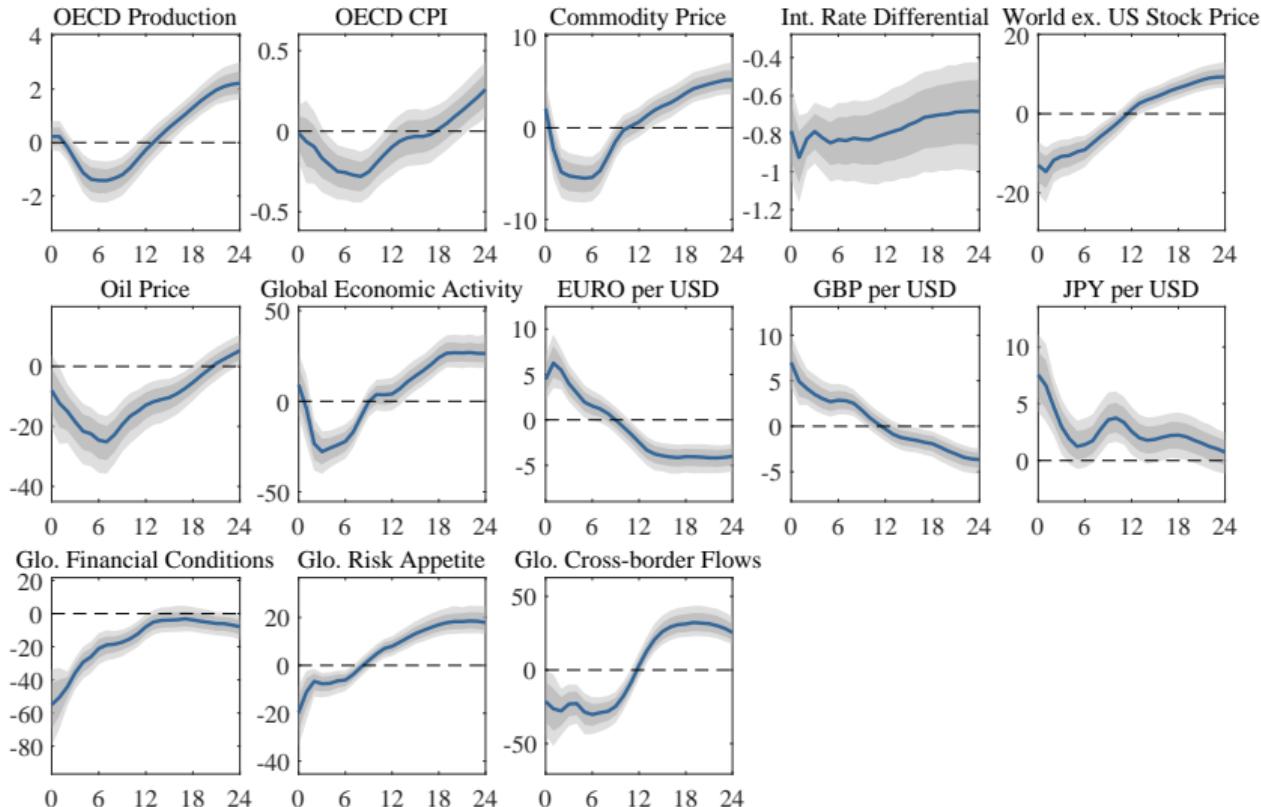
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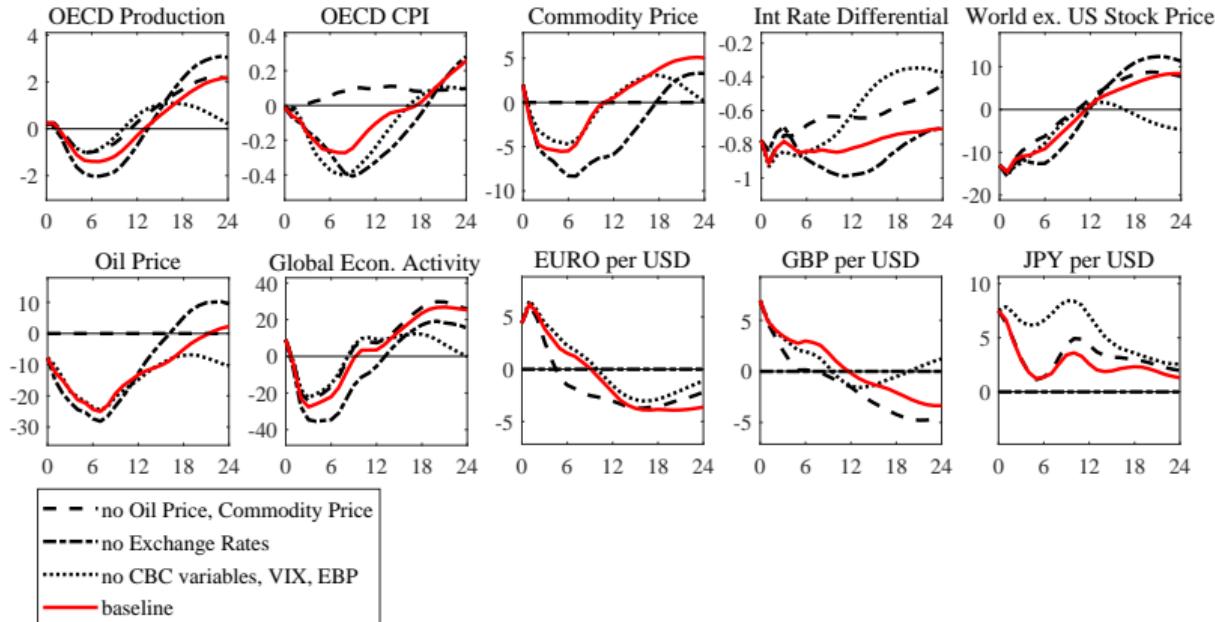
# Global Spillovers



# Global variables (BLP)



# Channels of Transmission



# **Advanced Economies**

# Countries and Data Coverage

<b>Countries</b>	<b>Data coverage</b>
Australia	Jan 1990 - Aug 2018
Austria	Jan 1990 - Aug 2018
Belgium	Jan 1990 - Aug 2018
Canada	Jan 1990 - Aug 2018
Denmark	Jan 1990 - Aug 2018
Finland	Jan 1990 - Aug 2018
France	Jan 1990 - Aug 2018
Germany	Jan 1990 - Aug 2018
Italy	Jan 1990 - Aug 2018
Japan	Jan 1990 - Aug 2018
Netherlands	Jan 1990 - Aug 2018
Norway	Jan 1990 - Aug 2018
Spain	Jan 1990 - Aug 2018
Sweden	Jan 1990 - Aug 2018
UK	Jan 1990 - Aug 2018

# Bilateral VARs

$$Y_{it} = c_i + \sum_{j=1}^p A_{ij} Y_{i,t-j} + u_{it}, \quad p = 12$$

$$Y_{it} = \begin{bmatrix} y_{i,t} \\ y_{US,t} \\ x_t \end{bmatrix}, \quad y_{i,t} = \begin{bmatrix} IP_{i,t} \\ CPI_{i,t} \\ \vdots \end{bmatrix}, \quad x_t = \begin{bmatrix} POIL_t \\ GEAI_t \\ VIX_t \end{bmatrix}$$

**Bilateral VAR** for country  $i$ :

- ▶ Asymmetric Normal-Inverse Wishart priors, i.e. no ‘spillback’ (Carriero et al, 2019)
- ▶ Optimal prior selection (Giannone et al, 2015)
- ▶ IV SVAR identification

# Bilateral VARs

<b>Foreign set</b>	<b>Logs</b>	<b>RW Prior</b>	<b>U.S. set</b>	<b>Logs</b>	<b>RW Prior</b>
Industrial Production Index	•	•	US Industrial Production Index	•	•
Consumer Price Index	•	•	US Consumer Price Index	•	•
Core CPI Index	•	•	US Core CPI Index	•	•
Nominal Stock Price Index	•	•	US Nominal Stock Price Index	•	•
Trade Balance			US Export/Import ratio		•
Nominal USD Exchange Rate	•	•	US Trade Volume	•	•
Short-term Interest Rate			US Nominal Effective Exchange Rate	•	•
Policy Rate			US 10-Year Treasury Constant Maturity Rate		•
Long-term Interest Rate			US Financial Conditions Index, CBC	•	
Financial Conditions Index, CBC	•		US Risk Appetite, CBC		
Policy Liquidity Index, CBC	•		US Cross-Border Flows Index, CBC	•	
Risk Appetite, CBC			BAA spread (corporate minus 10Y Treasury CM)		
Cross-Border Flows Index, CBC	•		CBOE VIX	•	
			US 1-year Treasury constant maturity rate		•
Global price of Brent Crude	•	•			
Kilian (2019) Global Economic Activity Index					
CRB Commodity Price Index	•	•			

# Aggregation into Median Economy

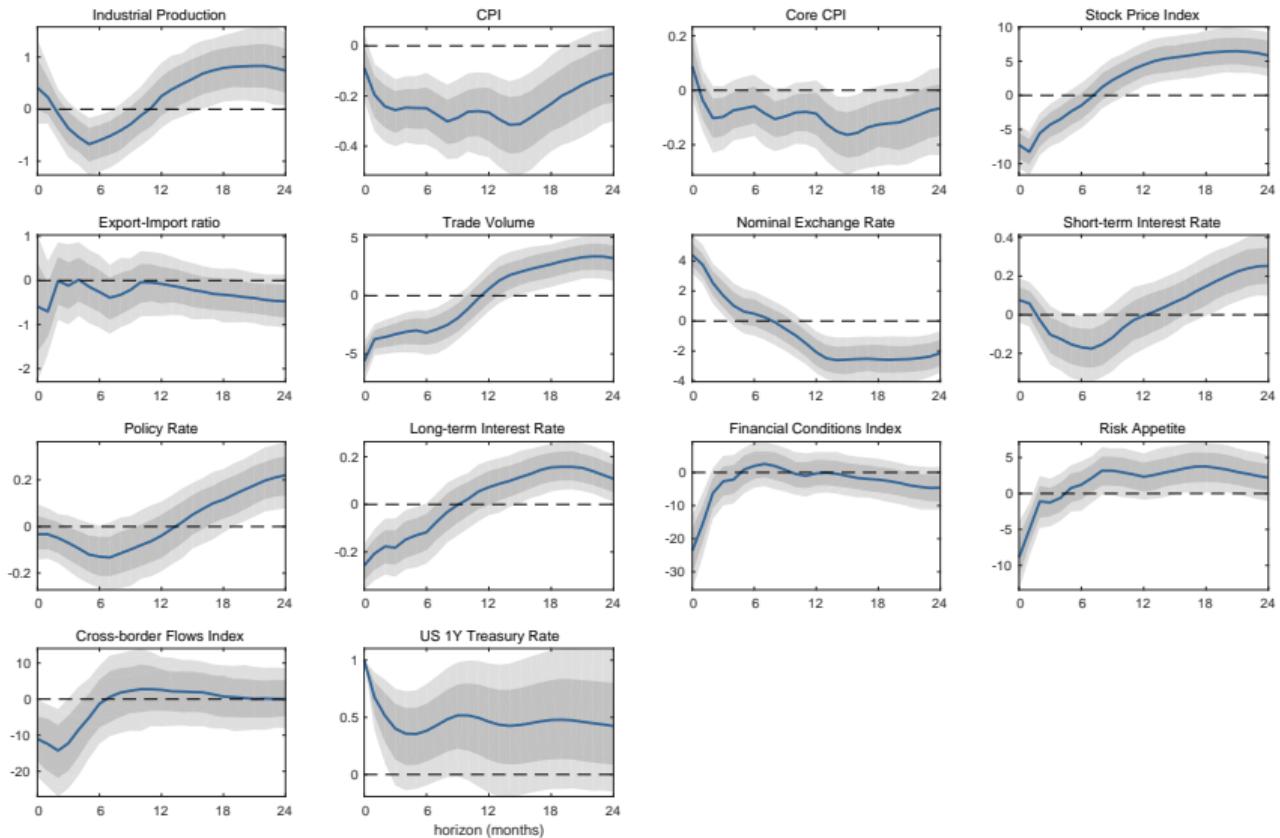
- ▶ Country  $i \in N$  IRFs for variable  $y$  at horizon  $h$

$$IRF_{y,i}^h = \frac{\partial y_{i,t+h}}{\partial \epsilon_{US,t}^{MP}}$$

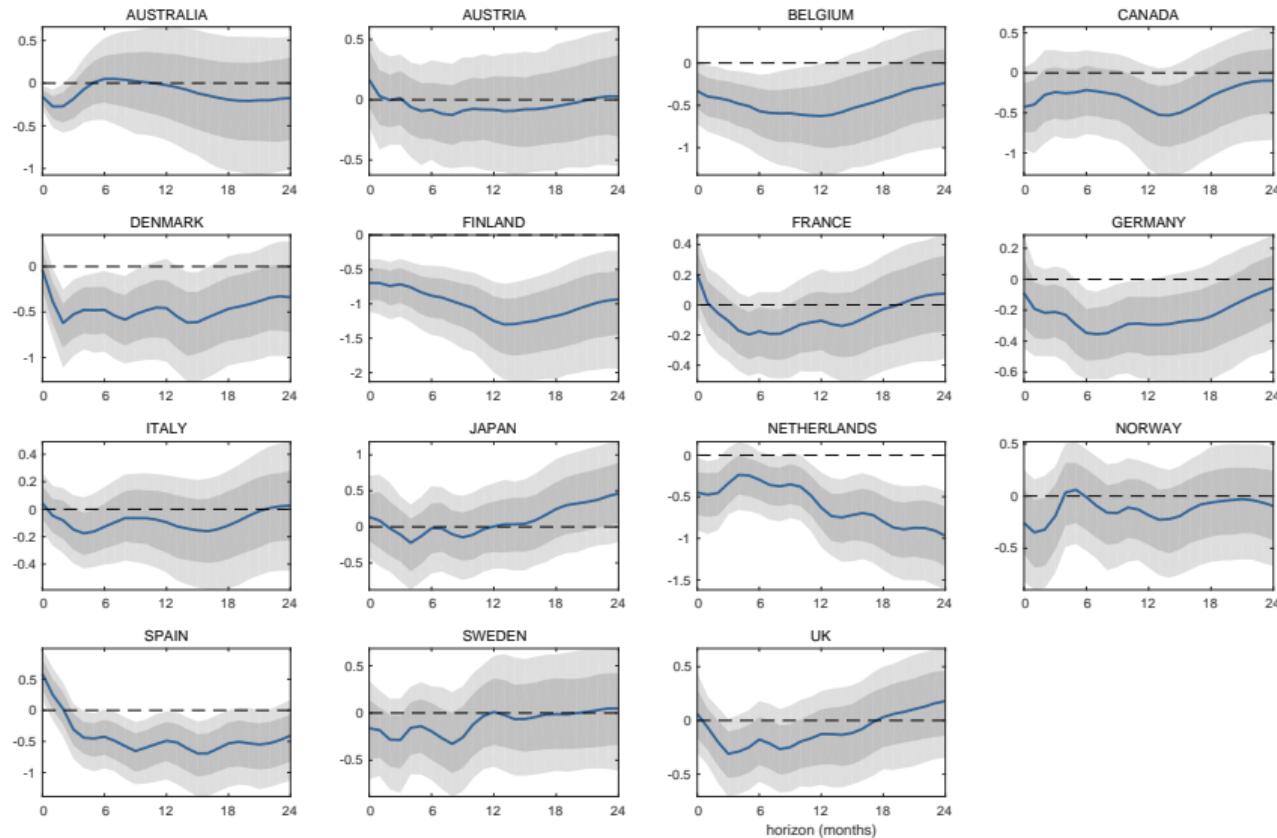
- ▶ Median country response

$$IRF_y^h = Median_i(IRF_{y,i}^h)$$

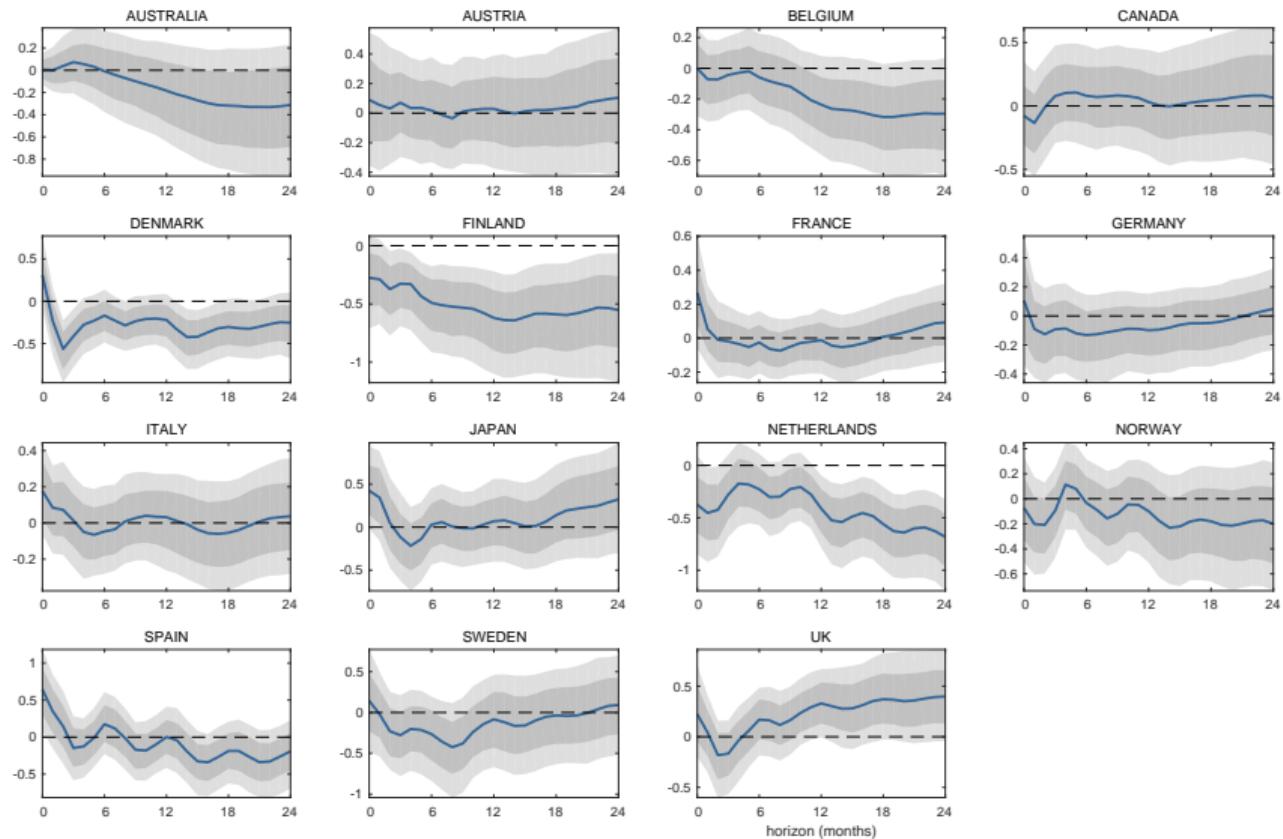
# Median Advanced Economy



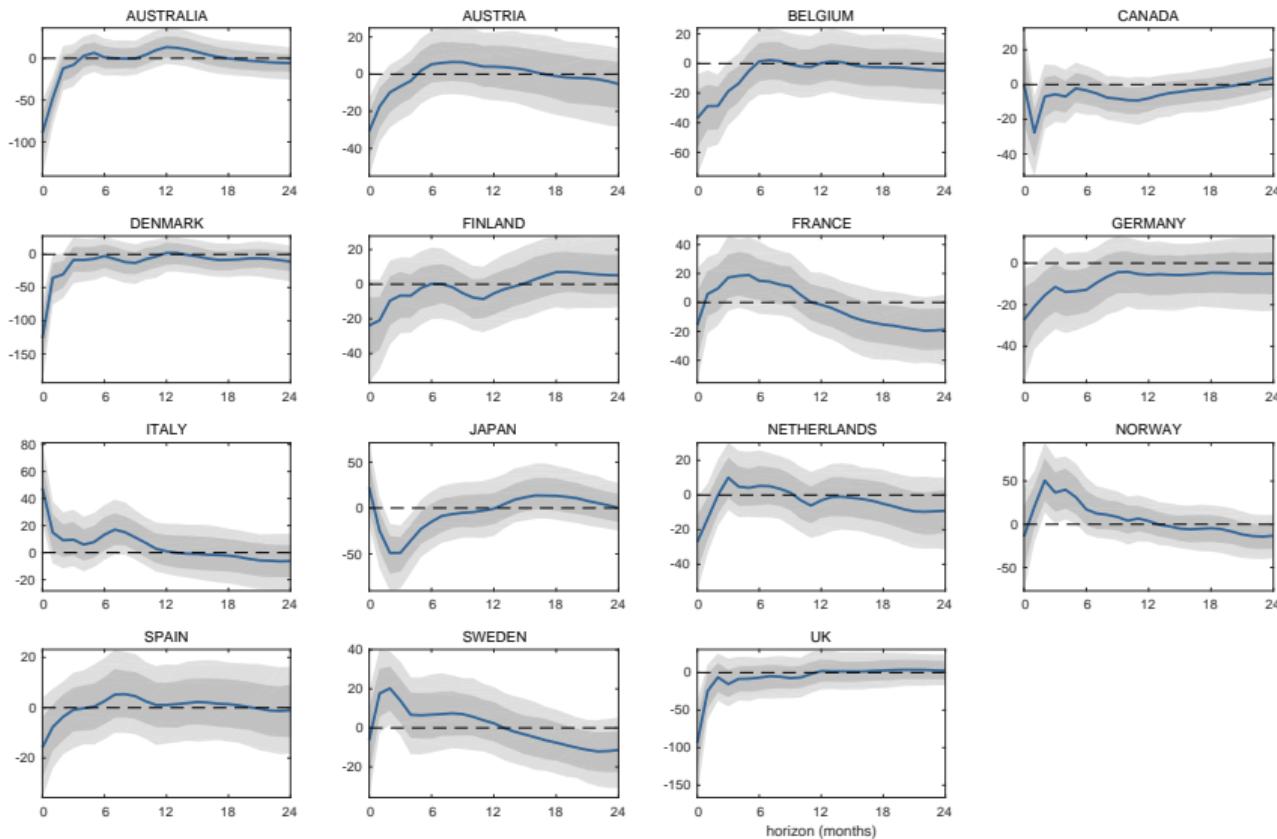
# Individual Countries – CPI



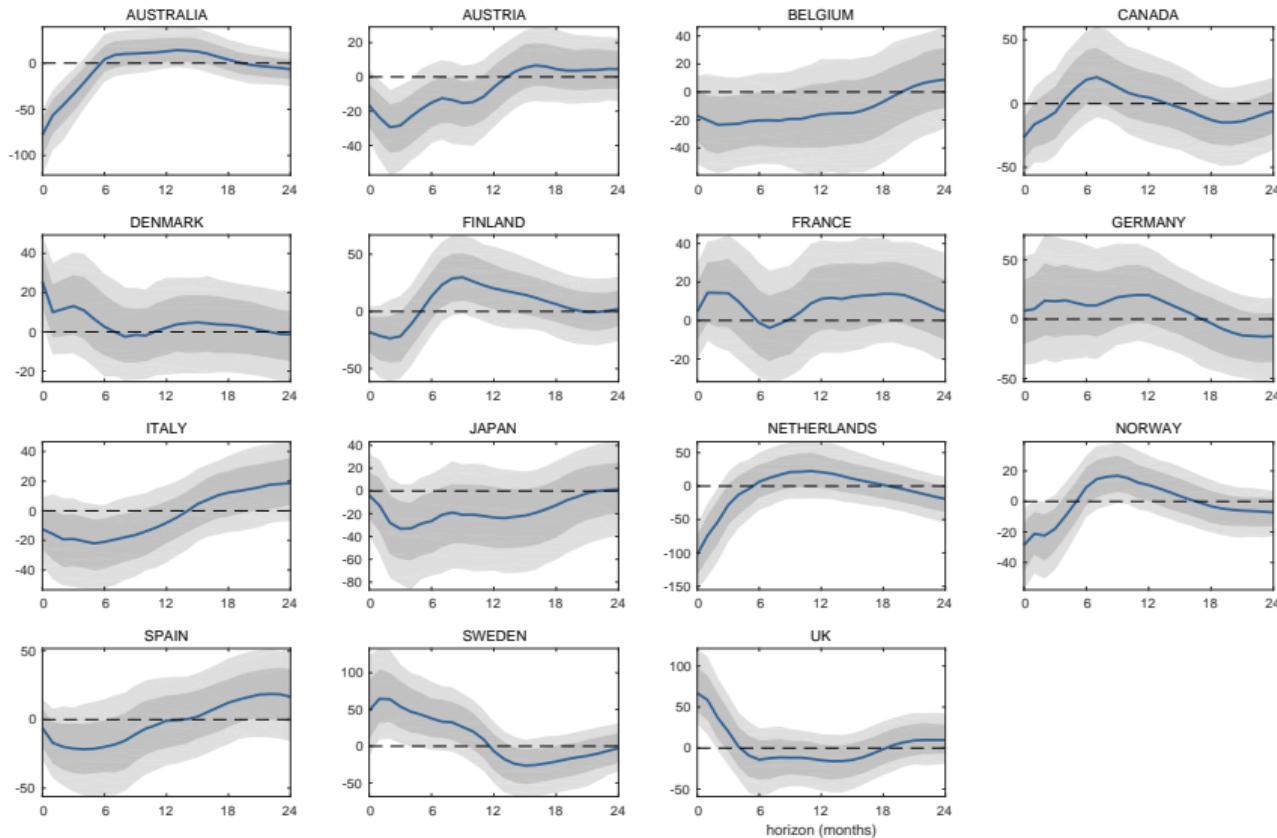
# Individual Countries – Core CPI



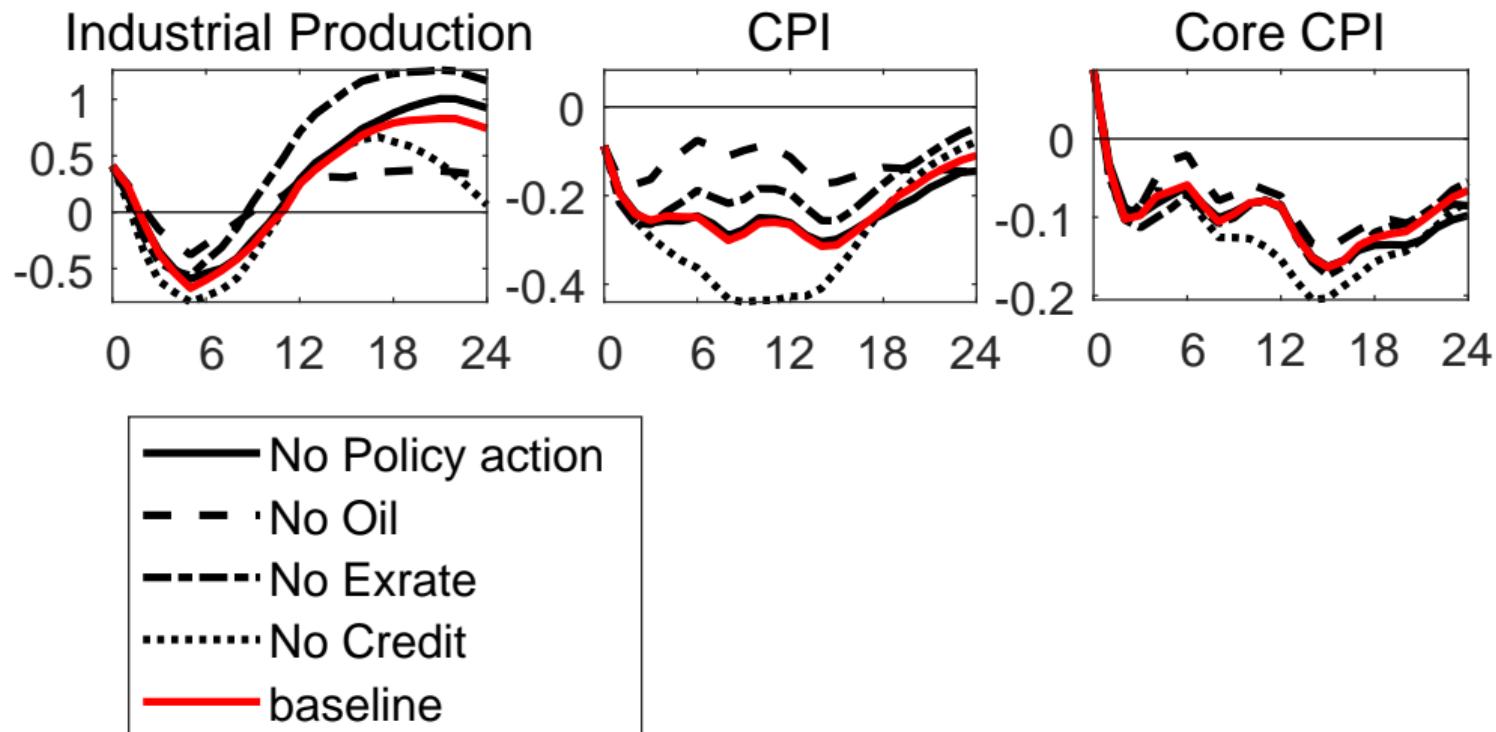
# Individual Countries – Financial Conditions



# Individual Countries – Cross-border Flows



# Channels of Transmission

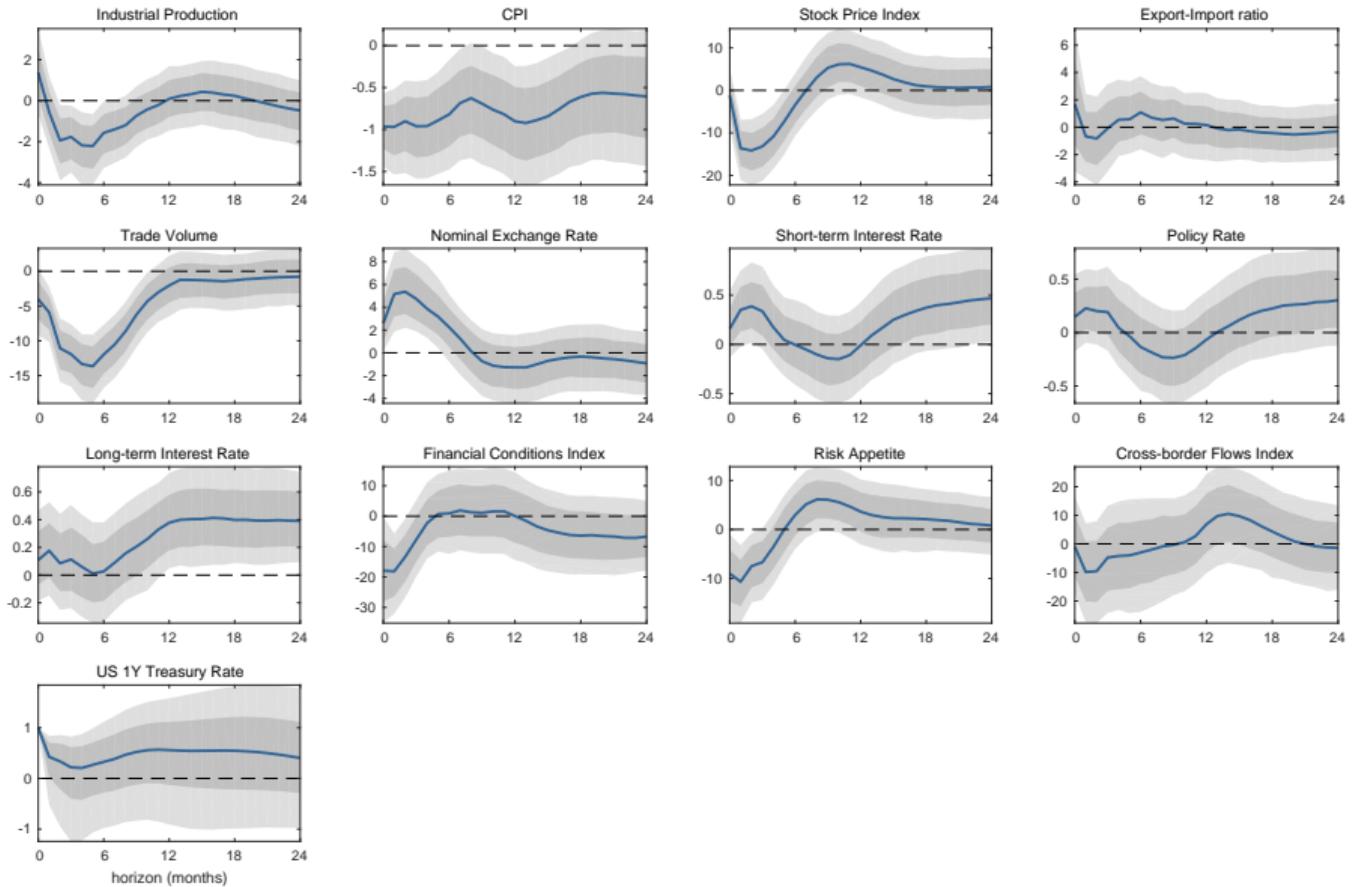


# **Emerging Economies**

# Countries and Data Coverage

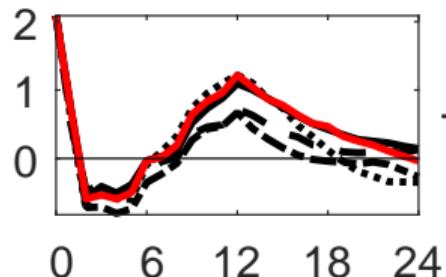
<b>Country</b>	<b>Data coverage w/ Core CPI</b>	<b>Data coverage w/o Core CPI</b>
Brazil	Dec 1999 - Aug 2018	Dec 1999 - Aug 2018
Chile	May 1995 - Nov 2013	May 1995 - Nov 2013
China		Aug 1994 - Aug 2018
Colombia	Sep 2002 - Aug 2018	Sep 2002 - Aug 2018
Czech Republic	Apr 2000 - Aug 2018	Apr 2000 - Aug 2018
Hungary	Feb 1999 - Aug 2018	Feb 1999 - Aug 2018
India		May 1994 - Apr 2018
Malaysia		Jan 1996 - Dec 2017
Mexico	Nov 1998 - Feb 2018	Nov 1998 - Feb 2018
Philippines	Jan 2000 - Feb 2018	Feb 1999 - Feb 2018
Poland	Jan 2001 - Aug 2018	Jan 2001 - Aug 2018
Russia	Jan 2003 - Jun 2018	Jan 1999 - Jun 2018
South Africa	Jan 2002 - Aug 2018	Jan 1990 - Aug 2018
Thailand	Jan 1999 - May 2018	Jan 1999 - May 2018
Turkey	Jun 2000 - Aug 2018	Jun 2000 - Aug 2018

# Median Emerging Economy

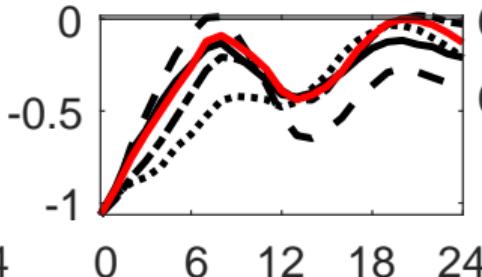


# Channels of Transmission

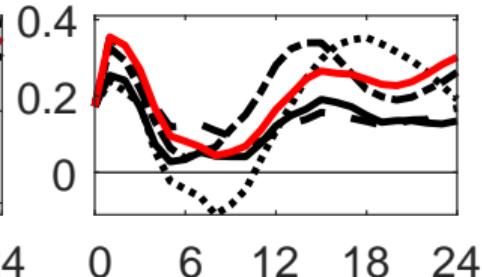
Industrial Production



CPI

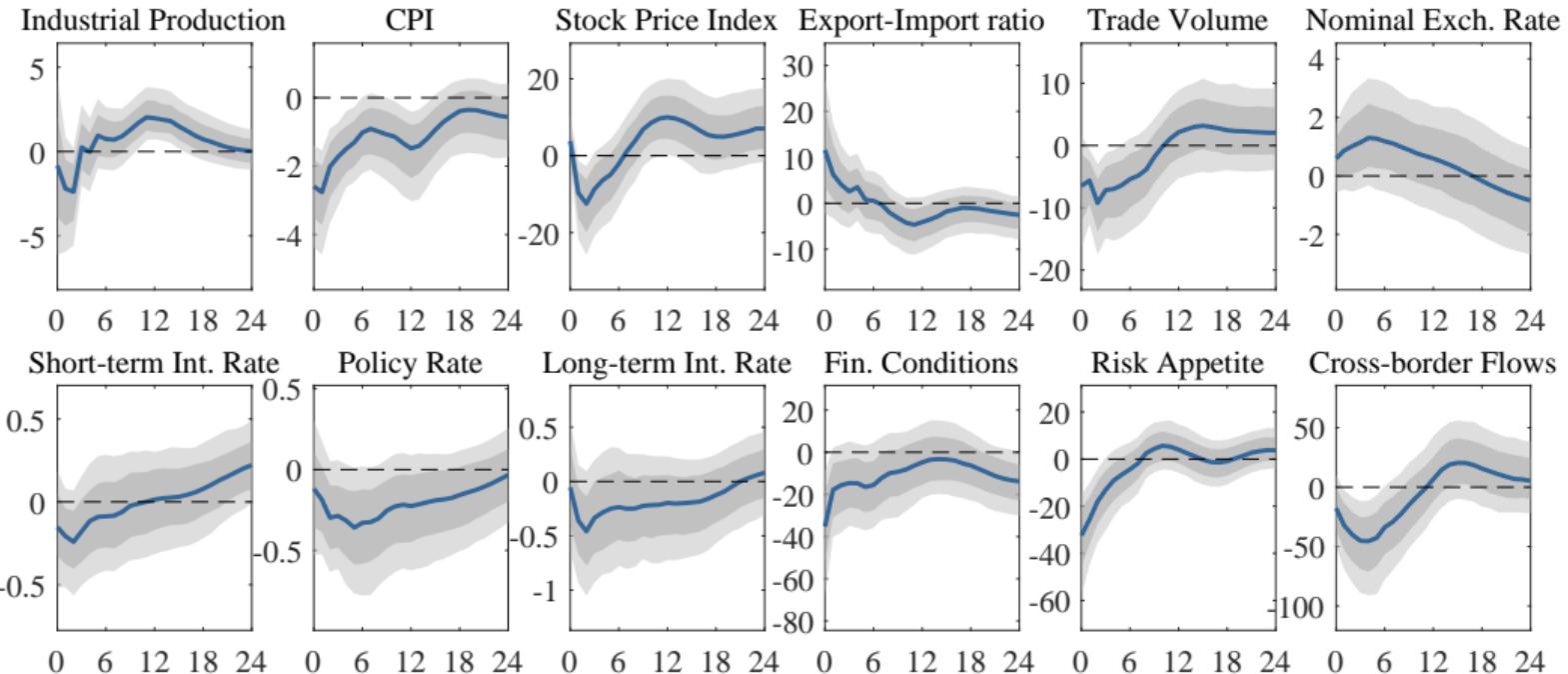


Short-term Interest Rate

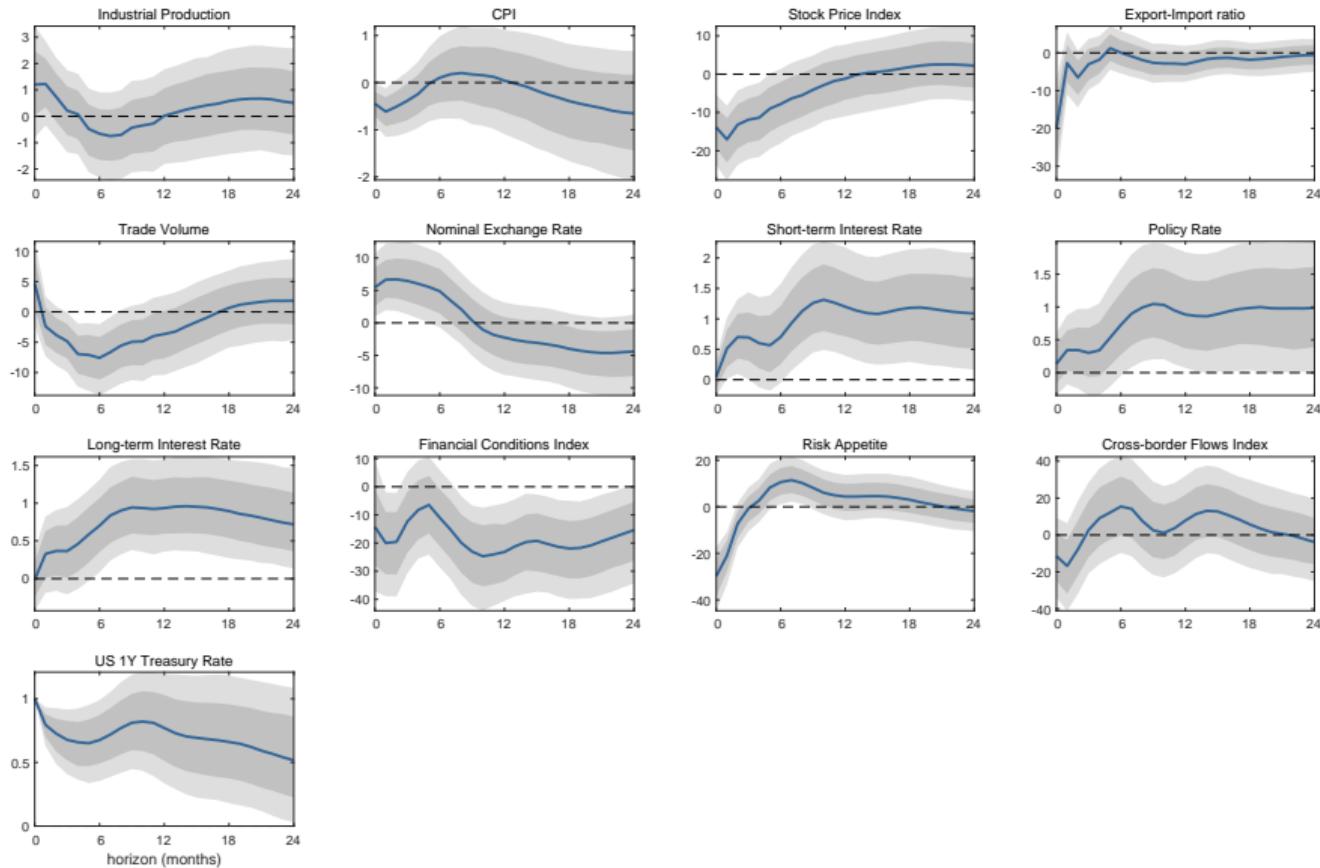


- No Policy action
- - - No Oil
- - - - No Exrate
- ..... No Credit
- baseline

# China



# South Africa



# **Capital Controls**

# Classification

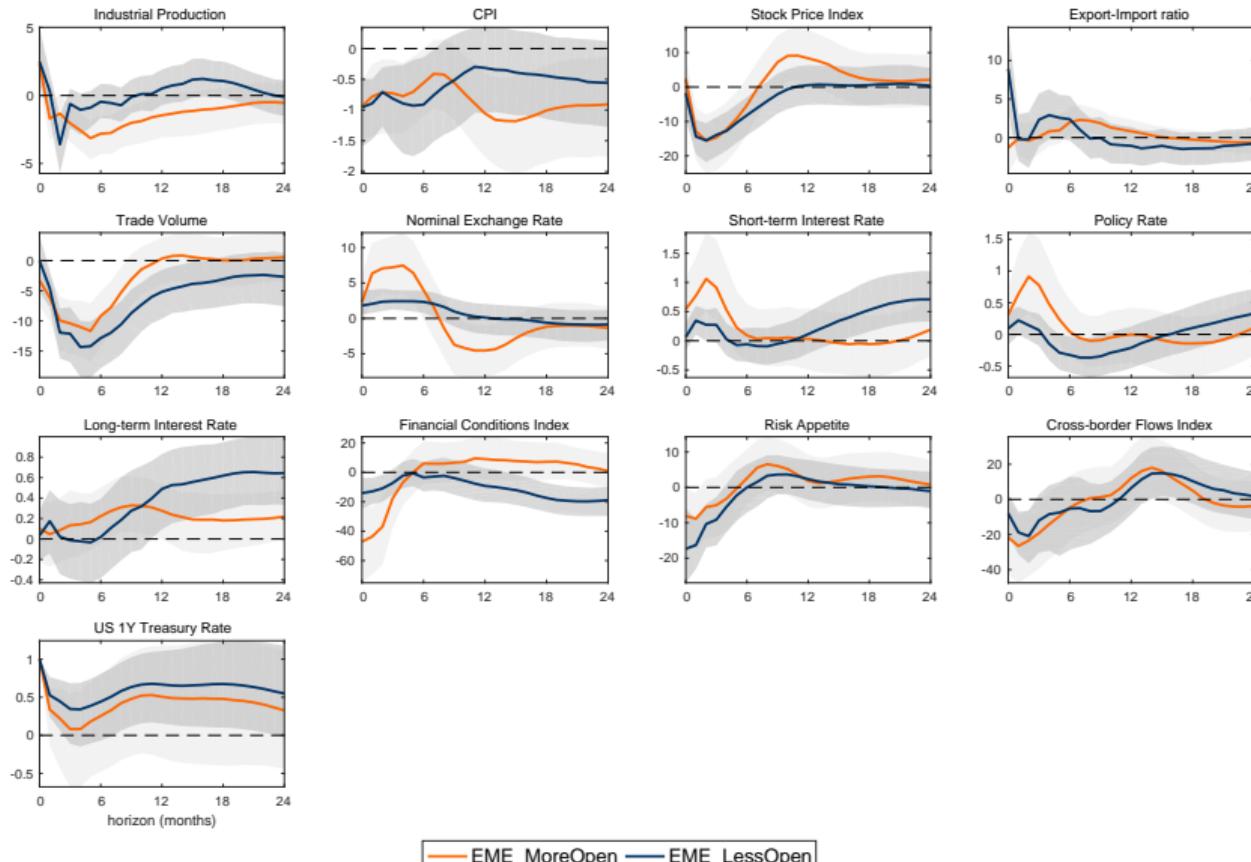
**Chinn-Ito Index, 1990-2015 Average**

<b>ADVANCED</b>	AUSTRALIA	0.814	<b>EMERGING</b>	BRAZIL	0.262
	AUSTRIA	0.965		CHILE	0.517
	BELGIUM	0.965		CHINA	0.147
	CANADA	1		COLOMBIA	0.272
	DENMARK	0.993		CZECH REP.	0.839
	FINLAND	0.965		HUNGARY	0.653
	FRANCE	0.944		INDIA	0.166
	GERMANY	1		MALAYSIA	0.513
	ITALY	0.944		MEXICO	0.640
	JAPAN	0.988		PHILIPPINES	0.393
	NETHERLANDS	1		POLAND	0.315
	NORWAY	0.886		RUSSIA	0.432
	SPAIN	0.898		SOUTH AFRICA	0.169
	SWEDEN	0.942		THAILAND	0.338
	UK	1		TURKEY	0.310
ADVANCED	MEDIAN	0.965	EMERGING	MEDIAN	0.338

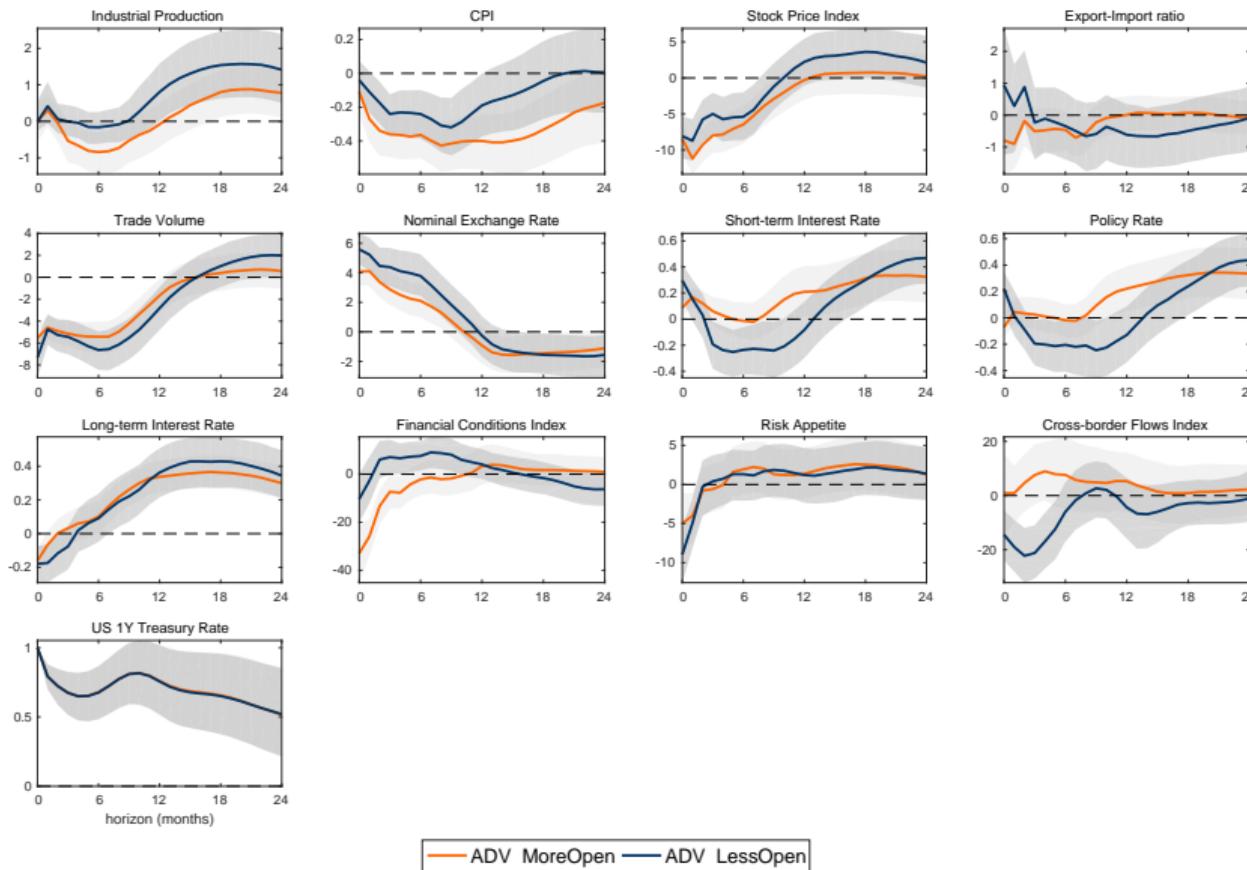
# Classification

Advanced		Emerging	
Open (Top 33%)	Less Open (Bottom 33%)	Open (Top 33%)	Less Open (Bottom 33%)
CANADA	AUSTRALIA	CHILE	BRAZIL
DENMARK	ITALY	CZECH REP.	CHINA
GERMANY	NORWAY	HUNGARY	COLOMBIA
NETHERLANDS	SPAIN	MALAYSIA	INDIA
UK	SWEDEN	MEXICO	SOUTH AFRICA
Average	0.998	0.897	0.632
			0.203

# Relatively Financially Open vs Closed Emerging Economy



# Relatively Financially Open vs Closed Advanced Economy



# **Exchange Rate Regimes**

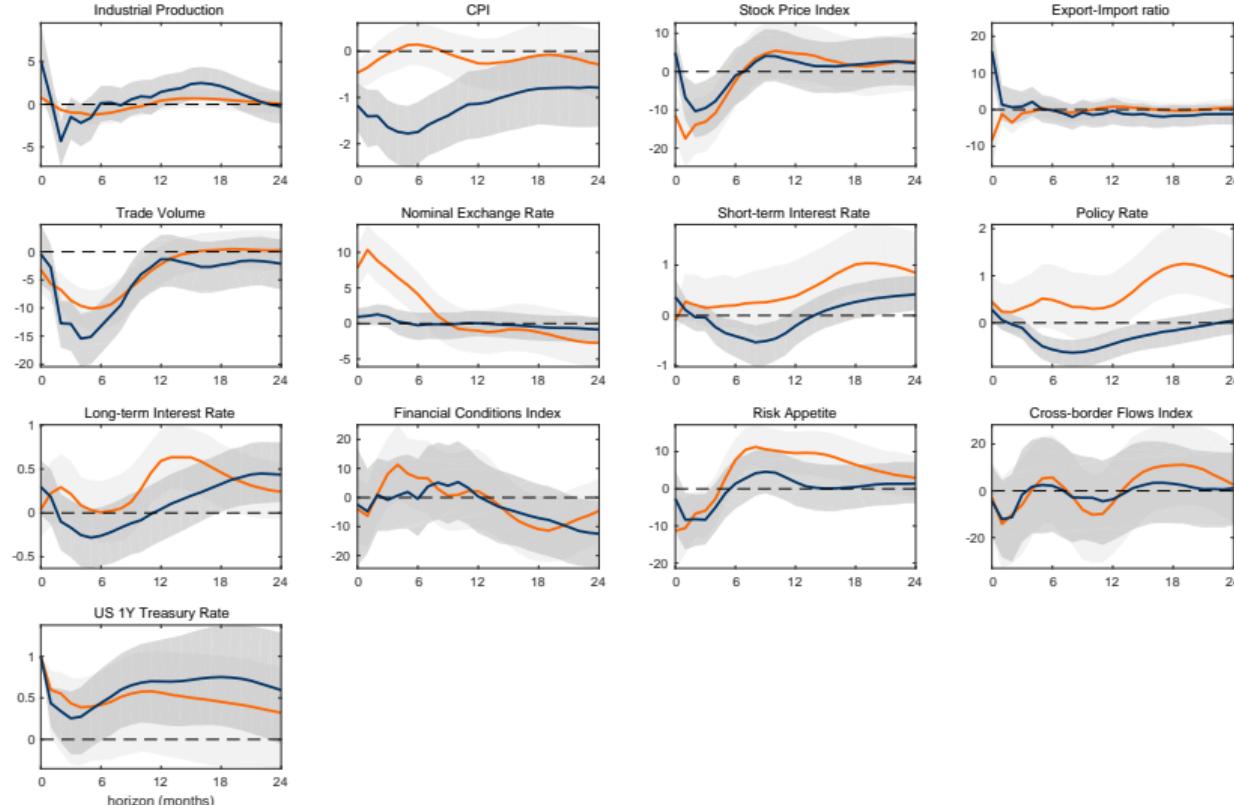
# Classification

## Ilzetzki-Reinhart-Rogoff (2019) Fine Classification

Managed floaters	Median IRR	Crawling pegs	Median IRR
BRAZIL	12	CHINA	5
COLOMBIA	12	INDIA	7
MEXICO	12	PHILIPPINES	10
SOUTH AFRICA	12	THAILAND	11

Notes: Medians across sample period of each country. 12: +/- 5% moving band; 11: +/- 2% moving band; 10: crawling band +/- 5%; 7: de facto crawling peg; 5: pre-announced crawling peg.

# Floating vs Pegging Emerging Economy



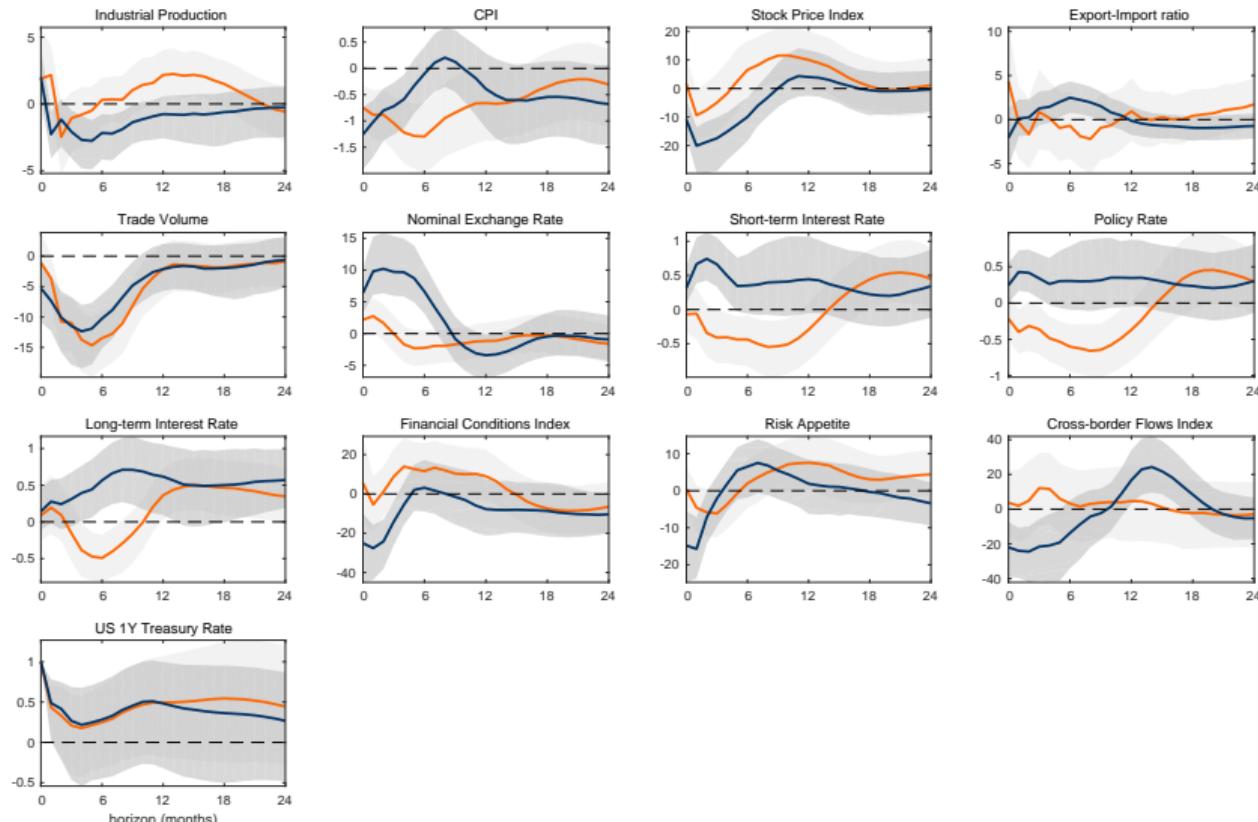
## **USD Trade Invoicing**

# Classification

Trade Invoices in USD (%)			
Highly USD	%	Low USD	%
BRAZIL	0.94	CZECH REP.	0.13
COLOMBIA	0.98	HUNGARY	0.18
INDIA	0.86	POLAND	0.3
MALAYSIA	0.9	TURKEY	0.46
THAILAND	0.82	SOUTH AFRICA	0.52

Notes: Data from Gopinath (2016).

# High vs Low USD trade invoicing



## Conclusions – The Global effects of U.S. Monetary Policy

- ▶ US monetary policy has large and pervasive effects
- ▶ **Spillovers** to both **prices and real economy**
- ▶ Strong evidence of financial channels
- ▶ Affect both **Advanced Economies** and **Emerging Markets**

# **Appendix**

# Standardisation (z-scores)

Method is the ‘standard’ for Financial Stability Boards:

$$z_t = \frac{x_t - MA(x_t, 40)}{SD(x_t, \text{entire period})}$$

- ▶ 40 months moving average trends
- ▶ SD over entire period to account for rare events
- ▶ Expressed in percentiles (50 is trend)

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# Financial Conditions Index

- ▶ AKA: Credit Spreads
- ▶ Index of very short-term credit spreads (e.g. deposit loan spreads)
- ▶ Derived from the z-scores of each sub-variable, which are equally weighted and then normalised
- ▶ Varies between 0 and 100 with 50 indicating ‘neutral’ relative to a 40-month MA

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# Policy Liquidity Index

- ▶ AKA: Central Bank Money
- ▶ Measures size of CB balance sheets and changes in their composition
- ▶ Derived from the z-scores of each sub-variable, which are equally weighted and then normalised
- ▶ Varies between 0 and 100 with 50 indicating 'neutral' relative to a 40-month MA

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## Cross-border Flows Index

- ▶ AKA: Foreign Liquidity
- ▶ Comprises all financial flows into a currency (incl. banking flows, bonds and equities)
- ▶ Estimated from national trade and current account data, foreign exchange reserve movements and (interpolated) quarterly data on net FDI flows
- ▶ Derived from the ‘normalised’ z-score of the value of cross-border flows. The resulting index varies between 0 and 100 with 50 indicating ‘neutral’ relative to a 40-month moving average
- ▶ **In the Global version, US weight is 24.81%**

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# Risk Appetite

- ▶ Difference between the Equity Exposure Index and the Bond Exposure Index
- ▶ **Equity Exposure Index:** AKA Equity Market Sentiment
  - ▶ Based on the balance sheet exposure of all investors (by type) in the asset class.
  - ▶ Derived from a z-score of the current portfolio share away from its 5-year trend, e.g. equity holdings as a percentage of all financial assets
- ▶ **Bond Exposure Index:** AKA Bond Market Sentiment
  - ▶ Equivalent of the equity exposure index for bonds

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