Commodity Prices and BRIC and G3 Liquidity: A SFAVEC Approach

Ronald A. Ratti[a*] and Joaquin L. Vespignani[b**]

[a]University of Western Sydney, School of Business, Australia
[b]University of Tasmania, School of Economics and Finance, Australia

Abstract

This paper investigates the influence of liquidity in the major developed and major developing economies on commodity prices. Unanticipated increases in the BRIC countries' liquidity is associated with significant and persistent increases in commodity prices that are much larger than the effect of unanticipated increases in G3 liquidity, and the difference increases over time. Over 1999-2012 BRIC liquidity is strongly linked with global energy prices and global real activity whereas G3 liquidity is not. The impact of BRIC liquidity on mineral and metal prices is twice as large as that of G3 liquidity. BRIC liquidity is significantly connected with global tightening while G3 liquidity is not. Granger casualty goes from liquidity to commodity prices. BRIC and G3 liquidity and commodity prices are cointegrated. BRIC and G3 liquidity and global output and global prices are cointegrated. We constructed a structural factor-augmented error correction (SFAVEC) model.

Keywords: Commodity Prices, BRIC countries, G3, Global liquidity, SFAVEC

JEL Codes: E31, E32, E51, F01, G15, Q43

[*]Ronald A. Ratti; University of Western Sydney, School of Business, Australia; Tel. No: +61 2 9685 9346; E-mail address: r.ratti@uws.edu.au

[**]Corresponding author: Joaquin L. Vespignani; University of Tasmania, School of Economics and Finance, Australia; Tel. No: +61 3 62262825; E-mail address: Joaquin.Vespignani@utas.edu.au

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1. Introduction

The effect of global liquidity on the prices of commodities, goods and assets has been a focus of recent research. Sousa and Zaghini (2007) find that global excess liquidity signals inflationary pressure at a global level. D'Agostino and Surico (2009) demonstrate that global liquidity has predictive power for the US inflation rate. Darius and Radde (2010) show that global liquidity has impact on a commodity price index (but not on equity prices and oil prices). Belke et al. (2010) document that the dramatic increase in global liquidity since 2001 has had impacts on the price of assets in inelastic supply including commodities. Anzuini et al. (2013) find that US monetary expansion has a significant, but modest effect on commodity prices. Ratti and Vespignani (2013) report that increases in global liquidity have had a positive effect on oil prices in recent years. Theoretically increases in liquidity are likely to be associated with a rise in aggregate demand and this will increase the price of most assets including commodity prices.[1]

In this paper we seek to determine the influence of liquidity as it arises from the major developed and major developing economies on commodity prices. Hamilton (2013) notes that the newly industrialized economies have absorbed over two-thirds of the increase in world oil consumption since 1998. Kilian and Hicks (2013) associate the rise in real oil price over 2003-2008 with growth in emerging economies, primarily that in China and India. Radetzki (2006) surmises that in developing Asian countries a dollar added to the GDP uses more than twice the quantity of commodities as does a dollar added to the GDP in OECD countries and notes that between 2000 and 2005, just China's share of global demand growth for petroleum was 28%, for aluminium was more than 50%, for steel was more than 84%, and for copper

[1] Barsky and Kilian (2004) maintain that monetary policy influences commodity prices through expectations of greater growth and inflation. Frankel and Hardouvelis (1985) argue that movement in commodities prices measure the market's assessment of the stance of monetary policy. Frankel (1984) notes that increase in money will raise the real price of commodities because the prices of many other goods are inflexible in the short.

was 95%. Humphreys (2010) notes that industrialization increases demand for metals substantially and that development in the BRIC economies is a major factor in the boom in metal prices from 2003 to 2008. Roberts and Rush (2010) argue that commodity resources are used intensively in traded goods and that this is a part of the demand for commodities by rapidly developing countries. Dungey et al. (2013) find that shocks to Chinese demand result in sustained increase in commodity prices in the Australian mining sector.

In this paper the major developing economies are taken to be the BRIC countries (Brazil, the Russian Federation, India and China). The major developed economies are taken to be the G3, the world's three leading economic blocs - the US, Japan and the European Union (EU). The BRIC countries have become much more important providers of global liquidity in recent years. Information on M2 in US dollars for the BRIC countries and for the G3 over 1999:01-2011:12 is provided in Figure 1. The scale of the right hand side of Figure 1 is for M2 for the BRIC countries and the scale of the left hand side of Figure 1 is for M2 for the G3. Over the fourteen years from 1999:01 to 2012:12 M2 is up approximately by a factor of 13.3 in BRIC countries. In contrast, over 1999:01 to 2012:12, M2 is up by a factor of 2.4 in the G3. BRIC M2 goes from being only about 10% of G3 M2 at the start of the period to being over 50% by the end of the period. Our view is that it greatly matters in assessing the impact of liquidity on commodity prices as to where the innovation in liquidity is originating.

The logs of US dollar commodity price index and commodity price component indices for energy commodities, agriculture commodities, mineral and metal commodities, precious metal commodities, and raw materials commodities are shown in Figure 2. The underlying indices are set at 100 for 2005. From 1999:01 to 2012:12 the commodity price index is up by a multiple of 3.71. Over the same period energy prices, agriculture prices, mineral and metal prices, precious metal prices, and raw materials prices have increased by multiples of about 4.22, 3.19, 3.60, 4.01, and 3.19, respectively.

A structural factor-augmented vector error correction model is employed in the analysis of the effect of innovations in BRIC liquidity and G3 liquidity on global commodity prices.[2] A structural factor-augmented dimension to the SVEC model will capture the dynamic of the information provided by many variables to the analysis of short and long run influence of liquidity on global commodity prices, global industrial production, global inflation and global interest rate. Granger casualty goes from liquidity to commodity prices. BRIC M2 and G3 M2 are cointegrated with commodity prices and with global inflation and global output.

It is found that positive innovations in BRIC liquidity are linked with much larger positive effects on commodity prices than are positive shocks in G3 liquidity. The disparity in the effect of BRIC liquidity compared G3 liquidity on commodity prices grows over time. Positive shocks in BRIC liquidity have much larger effects on energy prices, mineral and metal prices, and raw material prices than do positive shocks in G3 liquidity. Shocks to G3 liquidity have larger effects than shocks to BRIC liquidity on precious metal prices. A positive shock in BRIC M2 is associated with significant increases in global industrial production and global interest rates whereas shocks to G3 liquidity are not. Results are robust to alternative identification schemes in the structural factor-augmented vector error correction model, to different measurement of global variables, to treatment of the global financial crisis, and to variation in lag length.

The data, variables and cointegration are discussed in Section 2. The structural vector error correction model for analysis of liquidity and real crude oil prices is introduced in

[2] The literature on the identification of monetary policy in a VAR framework is expanding in several directions. Bernanke et al. (2005) propose a Factor-augmented VAR (FAVAR) to identify monetary policy shocks. A small number of factors (principal components) can summarize large amounts of information about an economy and be included in the FAVAR. Dees et al. (2007) propose a global VAR (GVAR). The GVAR combines separate models for each of the many economies linking core variables within each economy with foreign variables using quarterly data. The foreign variables external to a domestic economy are trade-weighted.

Section 3. The empirical results are presented in Section 4. Robustness of results is investigated in Section 5. Section 6 concludes.

2. Data, Variables and Cointegration

In this paper we will construct a structural factor-augmented error correction (SFAVEC) model to estimate the impacts of increases in BRIC and in G3 liquidity on global commodity prices. Given that the model will incorporate principal component variables, cointegration vectors and ordering restrictions it is convenient to discuss the data, variables and cointegration vectors in this section before discussing the SFAVEC model.

2.1. The data

The model is constructed with monthly data from January 1999 to December 2012. The starting period is dictated by the creation of the European central bank, the availability of Eurozone interest rate data, and the availability of data at monthly frequency for the BRIC countries. It is necessary to use monthly data since the sample period is unavoidably comparatively short.

The monetary aggregate indicators are M2 for the G3 economies ($G3\,M2_t$), US, Japan and the EU (taken to be the Eurozone and UK), and M2 for the BRIC economies ($BRIC\ M2_t$), Brazil, Russia, India and China. The monetary aggregates are measured in US dollars. Global commodity prices (COM_t), overall, energy, non-energy, mineral and metal, precious metal, and raw materials prices are in US dollars. We will also construct a global interest rate (GIR_t) variable, a global inflation rate ($GCPI_t$) variable, and a global industrial production (GIP_t) variable based the on the interest rate, industrial production and CPI inflation in each of the BRIC and G3 economies. The global commodity price data and commodity price component data are from World Bank and the International Monetary Fund. The M2 data for US, Eurozone, UK, Japan, Brazil and Russian Federation is from International Monetary Fund and the Federal Reserve Bank of St. Louis, while China's and

India's M2 are from People's Bank of China and Reserve Bank of India, respectively.[3] The exchange rate data to convert M2 series from domestic currency into US dollars is obtained from Federal Reserve Bank of St. Louis.[4] All other variables are from the Federal Reserve Bank of St. Louis (FRED data).

2.2. Global interest rate, inflation rate and industrial production

In this paper we will build structural factors such that each factor represents an economic or financial category of variables. We construct global indicators of the interest rate, inflation rate and industrial production. These variables for each of the BRIC and G3 economies clearly play a role in the link between global liquidity and commodity prices. A problem is to find a practical way to compress the information on interest rates, inflation and economic activity in each of the G3 and BRIC economies into a few variables. In this section we will construct global indicators of the interest rate, inflation rate and industrial production based on principal components methodology applied to data for the G3 and BRIC economies. In robustness tests of the results in the paper we will use an alternative method to obtain global indicators of these variables by using nominal GDP weights converted to a single currency (interpolated monthly) applied to the appropriate variable for each individual economy.

Bernanke et al. (2005) proposed a factor-augmented vector autoregressive model (FAVAR) based on the development of principal components analysis outlined by Stock and Watson (2002). One of the main advantages of this methodology is that a single individual variable or factor can capture the dynamic of a large amount of information contained in many variables. Facing a large number of variables included in this study, we use principal

[3] Note for India the monetary aggregates L2 has been used as a proxy of M2, as the Reserve Bank of India does not report monthly M2 aggregates for this period.
[4] Russian federation exchange rate has been interpolated from the annual series as monthly data for Ruble /US dollars is not available for the full sample period.

component indexes as indicators capturing the effects of global interest rates, global industrial production and global inflation by compressing in turn local information on these variables in the BRIC and G3 economies. The BRIC and G3 economies account for over 75% of global GDP measured by purchase power parity for the full data period. The structural factors or the indicators of global interest rate, global industrial production and of global inflation are the leading principal components of the BRIC and G3 economies' interest rates, industrial production and inflation (in log-level form for industrial production and CPI inflation):

$$GIR_t = [IR_t^{EA}, IR_t^{US}, IR_t^{CH}, IR_t^{Ja}, IR_t^{UK}, IR_t^{In}, IR_t^{Ru}, IR_t^{Br}] \quad (1)$$

$$GIP_t = [IP_t^{EA}, IP_t^{US}, IP_t^{CH}, IP_t^{Ja}, IP_t^{UK}, IP_t^{In}, IP_t^{Ru}, IP_t^{Br}] \quad (2)$$

$$GCPI_t = [CPI_t^{EA}, CPI_t^{US}, CPI_t^{CH}, CPI_t^{Ja}, CPI_t^{UK}, CPI_t^{In}, CPI_t^{Ru}, CPI_t^{Br}] \quad (3)$$

In equation (1), GIR_t is a vector containing the discount rate of the central banks of the Euro area, UK, US, China, Japan, India, Russia and Brazil. Equations (2) and (3) are vectors containing the industrial production and inflation for the same countries, respectively.[5]

The first principal component for the global interest rate, which to economize on notation we will refer to as GIR_t, is drawn in Figure 3a. It captures the collapse in interest rates at the end of 2008 with the onset of the global financial crisis as well as the relatively low interest rates over the period 2002 to 2006. The first principal component for the CPI indices, $GCPI_t$, is shown in Figure 3b. In Figure 3 $GCPI_t$ slopes linearly upward. This indicates an overall flat rate of inflation, consistent with low and moderate inflation in the BRIC and G3 economies over 1999-2012 (the basic data are logs of CPI levels). There are brief periods when inflation seems to flatten out or speed up in line with weaknesses in the global economy and movement in commodity prices.

[5] Discussion of the principal component estimation is presented in Appendix A.

The first principal component for global industrial production, GIP_t, is represented in Figure 3c. Global industrial production has an upward trend until the global financial crisis in 2008. There is a severe correction in GIP_t in 2008-2009, reflecting the global financial crisis, with recovery of global industrial production to early 2008 levels only in 2011. Global industrial production also shows a correction in 2001 coinciding with the March-November 2001 recession in the US.

Information on the correlations between country-specific and global factor for the short-term interest rate (IR), for industrial production (IP), and for the inflation rate (CPI) in turn are reported in the columns in Table 1. The global factors are given by first principal components for the global interest rate (GIR), global industrial production rate (GIP), and global inflation rate (GCPI). The global interest rate correlation with country interest rates is high for most countries and low for India and Brazil (with correlation coefficients of 0.32 and 0.38, respectively). The global industrial production correlation with country industrial production is high for each of the BRIC countries (at 0.92 and above), at 0.61 or 0.62 for the Euro area and the US, and low for Japan (at 0.31) and the UK (-0.62). The global CPI correlation is high with all economies CPI's at 0.95 and above, except for the correlation with Japan's CPI.

2.3. Causality test

In Tables 2 and 3 the Granger causality direction between G3 M2 and commodity prices and between BRIC M2 and commodity prices are reported. The null hypothesis that commodity prices do not Granger cause BRIC M2 and the null hypothesis that commodity prices do not Granger cause G3 M2 cannot be rejected at conventional levels (using most lags structures). These results hold when the test is performed in log-level and in log-difference form.

The null hypothesis of BRIC M2 does not Granger causes commodity prices is rejected to 1% level using both log-level and log-difference form, confirming that causal direction is from BRIC M2 to commodity prices. Results for the G3 M2 and commodity prices are less clear. The null hypothesis that G3 M2 does not granger cause commodity prices is rejected for variables in log-level form but is not rejected in difference-level form. Overall, we conclude that Granger casualty goes from liquidity to commodity prices.

2.4. Stationarity and Cointegration test

In table 4 the stationary properties of the data are reported. For this purpose both the Augmented Dickey-Fuller (ADF) test and the Kwiatkowski–Phillips–Schmidt–Shin (KPSS) are estimated for all variables.[6] Results show that variables are only first difference stationary. In empirical estimation the interest rate is used in levels in line with most macroeconomics studies.

We now examine possible cointegration among the price, monetary aggregate and output variables in the model. This is in keeping with the quantity theory of money as noted by and in empirical studies by Swanson (1998), Bachmeier and Swanson (2005), Garret et al. (2009) and others. The long-run relationship between global liquidity and commodity prices has been also well established in the literature.[7] Here we are interested in cointegration results for two relationships, that for BRIC M2, G3 M2 and global commodity prices, and that for BRIC M2, G3 M2, and global activity and global (consumer) prices. To formally establish the cointegration relationship among these variables, we use the Johansen's cointegration test.[8]

Table 5 reports results for the Johansson cointegration tests. In Table 5.1, results reveal that log of commodity prices, BRIC M2 and G3 M2 are a cointegrated vector when the

[6] Note that the appeal of using both methods is that they have inverse hypothesis. The null hypothesis for the ADF test is the variable has a unit root and the null hypothesis for the KPSS test is that the variable is stationary, this improve the robustness of the results.
[7] See for example Darius and Raddle (2010) and Belke et al. (2013).
[8] For more detail of this test please see Enders (2004), pp. 362.

test is specified with intercept. In table 5.2, the test for cointegration among global inflation, global output and G3 M2 and BRIC M2 shows one cointegration vector when both intercept and linear trend are introduced to the model. Consequently two cointegration vectors are introduced into the model and are specified as follows:

$$ECT1_t = log(COM_t) - \beta - \mu log(BRIC\ M2_t) - \sigma log(G3\ M2_t) \sim I(0) \quad (4)$$

$$ECT2_t = log(GCPI_t) - \beta - \gamma log(GIP_t) - \pi log(BRIC\ M2_t) - \varphi log(G3\ M2_t) - \theta t \sim I(0) \quad (5)$$

3. The model

Following Bernanke et al.'s (2005) idea of incorporating principal component vectors in a simultaneous equation model, we construct a structural factor-augmented error correction model with the following variables: GIR_t, $G3\ M2_t$, $BRIC\ M2_t$, $GCPI_t$, GIP_t and COM_t and two error correction terms specified in equations (4) and (5).[9]

The SFAVEC model can expressed as:

$$B_0 X_t = \beta + \sum_{i=1}^{j} B_i X_{t-i} + \omega ECT1_{t-1} + \rho ECT2_{t-1} + \varepsilon_t \quad (6)$$

where j is optimal lag length, determined by the Schwarz criterion (one lag in this case), X_t is vector of endogenous variables, ECT_t is the error correction term, and ε_t is the vector of structural changes, which is serially and mutually independent.

The vector X_t is expressed as:

$$X_t = [GIR_t, \Delta\log(G3\ M2_t), \Delta\log(BRIC\ M2_t), \Delta\log(GCPI_t), \Delta\log(GIP_t), \Delta\log(COM_t)\ ] \quad (7)$$

3.1. Generalized cumulative impulse response function

Country-specific SVAR studies such as Kim and Roubini (2000), Kim (2001) and Anzuini et al. (2013) use non-recursive contemporaneous restriction in order to identify the model based on economic theory and/or the estimated time of the central bank reaction to

[9] The FAVAR approach in Bernanke et al. (2005) uses a two-step procedure. Factors are estimated by principal components before estimation of the factor-augmented VAR. The number of unobservable factors used for policy analysis may be selected on statistical grounds. We use one factor for global interest rate, global industrial production and global inflation to retain parsimony in the structural factor-augmented VEC approach.

information release. However, in a global study such as this, there is not strong belief on variable ordering and contemporaneous restrictions. In a country-specific study it is possible to infer that the central bank can or cannot observes inflation contemporaneously based on the date at which inflation indicators is released and can change the interest rate accordingly. At the global level, whether global interest rate responds to global inflation is less clear, as the global variables are composed of several country-specific variables.

Consequently, the generalized cumulative impulse response (GIRF) developed by Koop et al. (1996) and Pesaran and Shin (1998) may be appropriate for this study.[10] Unlike conventional impulse response, generalized impulse response analysis approach is invariant to the ordering of the variables which is an advantage in absence of strong prior belief on ordering of the variables.[11] However, to assess the robustness of the results we will also examine outcomes obtained from identification strategies for the global variables that reprise those in the literature that are sensible for analysis of country level variables. These identification strategies that are frequently used in the literature are discussed below.

3.2 Alternative identification strategies

We evaluate different alternative contemporaneous restrictions based on Kim and Roubini (2000) and Anzuini et al. (2013). In these studies the basic variables are a central bank controlled interest rate, a monetary aggregate, consumer price inflation, industrial production and commodity price/oil price. Building on the identification strategies commonly used in the literature, completion of the model requires recognition and specification of the contemporaneous restrictions appropriate for the relationship between G3 M2 and BRIC M2. The four possibilities for whether G3 M2 and BRIC M2 respond or do not respond to each

[10] Note that other non-recursive alternatives are explored in the next section.

[11] In contrast with impulse response functions for structural models, GIRF do not require that we identify any structural shocks. The GIRF considers shocks to individual errors and integrates out the effects of the other shocks using the observed distribution of all the shocks without any orthogonalisation.

other in the same month are presented in equations (8) to (11) when building on the Kim and Roubini (2000) model.

In the Kim and Roubini (2000) model, consistent with Sims and Zha (1995)'s dynamic stochastic general equilibrium model, the monetary policy feedback rule is based on the recognition of information delays that do not allow the monetary policy to respond within the month to price level and output events. The monetary policy rule responds contemporaneously to G3 and BRIC M2. The world price of commodities (or the world price of oil in Kim and Roubini (2000)) influences the first principal component for the global interest rate to capture systematic response to (negative) supply shocks and inflationary pressure.

Following the literature, the M2 monetary aggregates respond contemporaneously to the domestic interest rate, inflation and industrial production implying that real demand for money depends on the interest rate and real income. Restrictions in both the inflation and output equations (four and fifth equations respectively) are standard in the economic literature. It is assumed that firms do not change their output and price contemporaneously in response to unexpected financial signals. The CPI and IP are assumed to be influenced in the same period by oil or commodity prices on the ground that commodities (e.g. oil and gas) are crucial inputs for many sectors.[12] Commodity price (or oil price) is taken to be contemporaneously exogenous to all variables in the model due to information delay.

$$B_o X_t = \begin{bmatrix} 1 & -b_{01} & -b_{02} & 0 & 0 & -b_{05} \\ -b_{10} & 1 & 0 & -b_{13} & -b_{14} & 0 \\ -b_{20} & -b_{21} & 1 & -b_{23} & -b_{24} & 0 \\ 0 & 0 & 0 & 1 & -b_{34} & -b_{35} \\ 0 & 0 & 0 & 0 & 1 & -b_{45} \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} GIR_t \\ \Delta \log(G3\ M2_t) \\ \Delta \log(BRIC\ M2_t) \\ \Delta \log(GCPI_t) \\ \Delta \log(GIP_t) \\ \Delta \log(COM_t) \end{bmatrix} \quad (8)$$

[12] These restrictions have been used by Gordon and Leeper (1994), Sims and Zha (2006), Christiano *et al.* (1999) and Kim (2001).

$$B_oX_t = \begin{bmatrix} 1 & -b_{01} & -b_{02} & 0 & 0 & -b_{05} \\ -b_{10} & 1 & 0 & -b_{13} & -b_{14} & 0 \\ -b_{20} & -b_{21} & 1 & -b_{23} & -b_{24} & 0 \\ 0 & 0 & 0 & 1 & -b_{34} & -b_{35} \\ 0 & 0 & 0 & 0 & 1 & -b_{45} \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} GIR_t \\ \Delta\log(BRIC\ M2_t) \\ \Delta\log(G3\ M2_t) \\ \Delta\log(GCPI_t) \\ \Delta\log(GIP_t) \\ \Delta\log(COM_t) \end{bmatrix} \quad (9)$$

$$B_oX_t = \begin{bmatrix} 1 & -b_{01} & -b_{02} & 0 & 0 & -b_{05} \\ -b_{10} & 1 & 0 & -b_{13} & -b_{14} & 0 \\ -b_{20} & 0 & 1 & -b_{23} & -b_{24} & 0 \\ 0 & 0 & 0 & 1 & -b_{34} & -b_{35} \\ 0 & 0 & 0 & 0 & 1 & -b_{45} \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} GIR_t \\ \Delta\log(G3\ M2_t), \\ \Delta\log(BRIC\ M2_t), \\ \Delta\log(GCPI_t) \\ \Delta\log(GIP_t), \\ \Delta\log(COM_t) \end{bmatrix} \quad (10)$$

$$B_oX_t = \begin{bmatrix} 1 & -b_{01} & -b_{02} & 0 & 0 & -b_{05} \\ -b_{10} & 1 & -b_{12} & -b_{13} & -b_{14} & 0 \\ -b_{20} & -b_{21} & 1 & -b_{23} & -b_{24} & 0 \\ 0 & 0 & 0 & 1 & -b_{34} & -b_{35} \\ 0 & 0 & 0 & 0 & 1 & -b_{45} \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} GIR_t \\ \Delta\log(G3\ M2_t), \\ \Delta\log(BRIC\ M2_t), \\ \Delta\log(GCPI_t) \\ \Delta\log(GIP_t), \\ \Delta\log(COM_t) \end{bmatrix} \quad (11)$$

In the setup in equation (8), G3 M2 influences BRIC M2 contemporaneously and not vice versa. In equation (9), BRIC M2 influences G3 M2 contemporaneously but is not influenced by G3 M2 contemporaneously. In equation (10), G3 M2 and BRIC M2 do not influence each other contemporaneously. Finally, in equation (11), G3 M2 and BRIC M2 influence each other contemporaneously.

Anzuini et al. (2013) have an alternative identification setup and treat oil prices and/or commodity prices as contemporaneously endogenous. They describe equilibrium in the commodity market as the result of arbitrage in a financial market in which all variables have contemporaneous effects on the commodity price. As in Kim and Roubini (2000), Anzuini et al. (2013) assume that the current level of prices and industrial production are not available to the monetary authorities and that the demand for real money balances depends on real activity and the nominal interest rate. Given price stickiness, real activity responds to price and financial signals only with a lag. Anzuini et al. (2013) differ from Kim and Roubini (2000) in assuming that the commodity price index does not to affect real activity or

consumer prices contemporaneously. An identification system analogous to Anzuini et al. (2013) is given in equation (12).

$$B_o X_t = \begin{bmatrix} 1 & -b_{01} & -b_{02} & 0 & 0 & -b_{05} \\ -b_{10} & 1 & 0 & -b_{13} & -b_{14} & 0 \\ -b_{20} & 0 & 1 & -b_{23} & -b_{24} & 0 \\ 0 & 0 & 0 & 1 & -b_{34} & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ -b_{50} & -b_{51} & -b_{52} & -b_{53} & -b_{54} & 1 \end{bmatrix} \begin{bmatrix} GIR_t \\ \Delta \log(G3\, M2_t), \\ \Delta \log(BRIC\, M2_t), \\ \Delta \log(GCPI_t) \\ \Delta \log(GIP_t), \\ \Delta \log(COM_t) \end{bmatrix} \quad (12)$$

In equation (12), G3 M2 and BRIC M2 are treated as contemporaneously exogenous. The model in equation (12) is just identified. With the inclusion of the assumption in equation (12) that commodity price is endogenous to all other variables, assumptions that either G3 M2 or BRIC M2 influences the other contemporaneously would render the model unidentified.

3.3 Log likelihood (LR) over-identifying restrictions test

In this section, we use the log likelihood ratio (LR) test for over-identification restrictions to assess if these restrictions fit the data. Results for this test are presented in Table 6 for restrictions proposed in equation (8) to (12). In Table 6, column 2 shows the number of over identified restriction and column 3 shows the chi-square values. In column 4 the probability values of rejecting the null hypothesis of "restrictions are valid" are presented. The probability values show that this test strongly rejects the restrictions imposed in model 8 to 11 at the 5% level. Among these models, the model in equation (8), the model that only restricts the contemporaneous impact of BRIC M2 on G3 M2, is marginally the best model. In contrast, the restrictions imposed in equation (12) cannot be rejected even at the 30% level, indicating that these restrictions fit the data fairly well compared to the models in equation (8) through (11).

4. The empirical results

4.1. Cumulative impulse responses of global variables to BRIC M2 and G3 M2

We will report the responses of the variables in the SFAVEC model in equations (6)-(7) to one standard deviation generalised cumulative impulse response function in BRIC M2 and in G3 M2. This is because of the difficulty in identifying a system of restrictions that apply to the global variables in the analysis. For the purpose examining the reasonableness of the results, we will also report cumulative impulse responses of the variables in the SFAVEC model in the vector (7) to one standard deviation cumulative impulse response function in BRIC M2 and in G3 M2 for the model with identification restriction in equation (8) and for the model with identification restriction in equation (12). Both these latter models represent restriction on the interaction global variables that are considered to be sensible for variables at a country level.

4.1.1. Generalized cumulative impulse responses of global variables to BRIC M2 and G3 M2

Figures 4a and 5a show the responses of the variables in the SFAVEC model in equations (6)-(7) to one standard deviation generalised cumulative impulse response function in BRIC M2 and in G3 M2.[13] We are using one standard deviation generalised cumulative impulse response function following Pesaran and Shin (1997).[14] The dashed lines represent a one standard error confidence band around the estimates of the coefficients of the cumulative impulse response functions.[15] It is found that positive innovations in the BRIC countries' liquidity lead to significant and persistent increases in global interest rates, global industrial

[13] One standard deviation in BRIC M2 is 0.012 and one standard deviation in G3 M2 is 0.015. If the impulses were normalized, this would reduce the apparent impact of shocks on commodity price of G3 M2 compared to BRIC M2. Thus, the cumulative impulse response results reported in this section understate the influence of BRIC M2 compared to G3 M2 somewhat. For purposes of comparison the standard deviation of the overall commodity price variable is 0.046.

[14] The advantage of this generalized approach is that unlike the "orthogonalized" impulse responses, it is not invariant to the order of vector autoregression (VAR) variables.

[15] The confidence bands are obtained using Monte Carlo integration as described by Sims (1980), where 5000 draws were used from the asymptotic distribution of the VAR coefficient.

production and commodity prices. The rise in commodity prices is sharp in the first three months following the BRIC M2 shock and then gradually continues to increase. Following the BRIC M2 shock, most of the rise in global industrial production occurs in the first four months whereas the global interest rate continues to rise as time goes on. The positive shock in BRIC M2 is associated with a boom in global industrial production and global tightening in monetary policy as indicated by increases in central bank discount rates. Innovations in the BRIC countries' liquidity do not significantly affect global inflation. A positive shock in BRIC M2 is associated with positive increase in G3 M2.

In Figure 5 shocks to the G3 economies' liquidity are not associated with statistically significant changes in global interest rates, global inflation, global industrial production, or BRIC M2. A positive innovation in G3 M2 does lead to an increase in commodity prices that is statistically significant for ten months. However, positive innovations in BRIC M2 are linked with a positive effect on commodity prices that is three times as large as the effect of unanticipated increases in G3 liquidity on commodity prices after three months. The magnitude of this relatively larger effect of BRIC M2 compared to G3 M2 on commodity prices then slowly grows over time. The one standard deviation generalised cumulative impulse response functions of the variables in the SFAVEC model in equations (6)-(8) to shocks to all the variables is available as Figure 10 in the Appendix.

4.1.2. Cumulative impulse responses of global variables to BRIC M2 and G3 M2 in SVEC models

Figures 4b and 5b show the responses of the variables in the SFAVEC model in equations (6), (7) and (8) to one standard deviation cumulative impulse response function in BRIC M2 and in G3 M2. Figures 4c and 5c show the responses of the variables in the SFAVEC model in equations (6), (7) and (12) to one standard deviation cumulative impulse

response function in BRIC M2 and in G3 M2. The dashed lines represent a one standard error confidence band around the estimates of the coefficients of the impulse response functions.[16]

In both sets of results, positive innovations in BRIC M2 are linked with a positive effect on commodity prices that are much larger than the effect of unanticipated increases in G3 liquidity on commodity prices after two or three months. In Figures 4b and 4c BRIC M2 has a statistically effect on commodity prices. In Figure 5b G3 M2 does not have a statistically effect on commodity prices and in Figure 5c G3 M2 does have a statistically effect on commodity prices, but the effect is only half as large as that of BRIC M2 on commodity prices. In addition, global industrial production responds significantly to BRIC M2 (and not to G3 M2) in both sets of result, in line with the generalized cumulative impulse response result for global industrial production already noted.

Overall, the cumulative impulse response results in Figures 4c and 5c from the model in the equations (6), (7) and (12) are similar to the responses of the variables in the SFAVEC model in equations (6)-(7) to one standard deviation generalised cumulative impulse responses in Figures 4a and 5a. A difference between sets of result is that G3 M2 no longer responds significantly to BRIC M2 in Figure 4c. However, a restriction in the system of equations (6), (7) and (12) underlying the result in Figure 4c, G3 M2 and BRIC M2 are restricted to not influence each other contemporaneously.

Results in Figures 4b and 5b from the model in the equations (6), (7) and (8) have several differences compared to the generalized cumulative impulse responses of the variables in Figures 4a and 5a. This is consistent with the rejection of the restrictions in the model compared to the null hypothesis of the Cholesky decomposition restrictions.

4.2. Cumulative response of commodity prices

[16] The confidence bands are obtained using Monte Carlo integration as described by Sims (1980), where 5000 draws were used from the asymptotic distribution of the VAR coefficient.

The cumulative contributions to commodity price of the structural shocks to G3 M2 and to BRIC M2 are reported in Figure 6a from estimating the SFAVEC model in equations (6)-(8). The cumulative contributions of structural shocks to commodity price are the moving average of the last 12 months to improve the readability of the plot. Contextual information on the behaviour of commodity prices over 1999-2012 is provided in Figure 2. Commodity prices fell during 2001 with recession in the US and fell sharply at the end of 2008 during the global financial crisis. Commodity prices rose particularly sharply over 2006-2008.

In Figure 3 the rapid increase in commodity price leading to a peak in June 2008 is associated with positive structural shocks to BRIC M2. The fall in commodity price from July 2008 to January 2009 is associated with the global financial crisis during late 2008, recession in the US over December 2007 to June 2009, and weak growth in Europe. Figure 3 suggests that BRIC M2 and G3 M2 did not contribute to this decline in commodity price. The cumulative impact of the BRIC countries' M2 on the commodity price is positive in the recovery of commodity price during 2009 and 2010.

Figure 6b shows the difference in the cumulative effect on commodity price of structural shocks to BRIC M2 and G3 M2 (BRIC-G3 M2) over 1999:01-2012:12. A positive (negative) value for BRIC-G3 M2 indicates a larger (smaller) effect of BRICM2 on commodity price than that of G3M2. In Figure 6b the contribution to commodity price of liquidity in BRIC countries relative to that of liquidity in G3 countries is much bigger since 2005. The relative contribution of the BRIC countries' liquidity to commodity price is particularly important during 2006 through 2008 and from the end of 2009 through 2010, in line with the rise in the economic importance of the BRIC economies.

4.3. Generalized cumulative impulse response of commodity price components

We now examine the response of commodity price components to innovations in BRIC M2 and in G3 M2. In the SFAVEC model in equations (6) and (7) the variable

$\Delta \log(COM_t)$ is replaced by the log difference of a commodity price component index (one at a time). The commodity price component indices considered are for energy commodities, agriculture commodities, mineral and metal commodities, precious metal commodities, and raw materials commodities. The responses of the commodity price component indices innovations in G3 M2 and in BRIC M2 are shown in first column and in the second column, respectively, in Figure 7.

It is found that positive innovations in the BRIC countries' liquidity lead to statistically significant and persistent increases in global energy prices. The rise in energy prices is very steep in the first two months and then energy prices continue to rise. Shocks to G3 liquidity have a small positive effect on global energy prices that is not statistically significant. Positive innovations in both G3 and BRIC liquidity have positive and statistically significant effects on agricultural prices that persist over time. The size of the effects of G3 and BRIC liquidity on agricultural prices are similar in the first few months, with a tendency for the BRIC effect to grow larger over time while the G3 effect does not (after the first four months).

Positive shocks in G3 and BRIC liquidity have positive and statistically significant effects on mineral and metal prices that persist over time. The size of the impact of BRIC M2 on mineral and metal prices is over 60% larger than that of G3 M2 three months after the shock. The BRIC liquidity effect on mineral and metal prices continues to grow larger over time and after twenty months is over twice the size of the effect of increases in G3 liquidity. Positive innovation in BRIC M2 accompanies statistically significant and growing increase in raw materials prices. Positive shocks to G3 M2 have a positive effect on global raw materials prices that is statistically significant over a two to six month window. Increases in both G3 and BRIC liquidity have positive and statistically significant effects on precious metal prices

that persist over time. On precious metal prices, the size of the effects of G3 M2 is twice as large as that of the effects of and BRIC M2.

The overall conclusion of this section in that positive shocks in BRIC M2 have much larger effects on commodity prices, energy prices, mineral and metal prices, and raw material prices than do positive shocks in G3 M2. Shocks to G3 liquidity did not have a statistically significant effect on global energy prices. It is only on precious metal prices that shocks to G3 M2 have larger effects than shocks to BRIC M2. The results for the effects of structural innovations in BRIC M2 and in G3 M2 on commodity price component indices is consistent with the result in the previous sub-section that the magnitude of the positive effect of positive BRIC M2 innovations compared to positive G3 M2 innovations on commodity prices is much larger and that the disparity grows over time.

5. Robustness analysis

In this section we evaluate the robustness of our model by exploring outcomes when using different indicators for global interest rates, industrial production and inflation, alternative identification restrictions, and lag structures.

5.1. Global variables: alternative global weights

In the earlier analysis the influence of global interest rate, global industrial production and global inflation is captured by the leading principal components from interest rates, industrial production and inflation in the BRIC and G3 economies. Beyer et al. (2000), Giese and Tuxen (2007), and Belke et al. (2013) aggregate global variables by using nominal GDP weights converted to a single currency (interpolated monthly). We constructed a global indicator of interest rate, inflation and industrial production by using nominal GDP relative to total GDP (G3 and BRIC GDP) weights for the US, Eurozone, UK, Japan, Brazil, Russia, India and China economies. These three global indicators are substitute for GIR_t,

$\Delta \log(GCPI_t)$ and $\Delta \log(GIP_t)$ in equations (5) to (7). The cumulative impulse responses are re-estimated from the reconstituted SFAVEC model.

Figures 8 and 9 shows the response in global variables to one standard deviation generalised impulses in BRIC M2 and in G3 M2, respectively. Positive innovations in the BRIC countries' liquidity lead to significant and persistent increases in global interest rates, global industrial production and commodity prices. As before, following the BRIC M2 shock, the rise in commodity prices is steep in the first few months and then continues to increase. A difference from before (reported in Figures 4a and 5a) is that now positive innovations in BRIC M2 significantly affect global inflation. However, by comparing Figures 8 and 9, it can be seen that increases in BRIC M2 are associated with increases in global inflation that are over twice as large as those linked with increases in G3 M2.

In Figure 9 shocks to the G3 economies' liquidity are not associated with statistically significant changes in global interest rates or global industrial production. A positive innovation in G3 M2 does lead to an increase in commodity prices that is larger than in the earlier principal component model and is statistically significant for twenty months. However, positive innovations in BRIC M2 continue to be linked with a positive effect on commodity prices that is larger than is the effect of unanticipated increases in G3 liquidity on commodity (although not by as great a margin). The magnitude of this relatively larger effect of BRIC M2 compared to G3 M2 on commodity prices slowly grows over time.

5.2. Alternative lag-lengths

An alternative lag selection can be selected using the Akaike information Criterion (AIC) rather than SC criterion used in our previous estimation. The AIC select two lags (rather than one) in estimation of the SFAVEC model described in equations (6) and (7). The model is re-estimated with two lags and generalized cumulative impulse response results are very similar to those obtained in Figures 4a and 5a. While results stay statistically significant

the confidence bands around point estimates tend to increase slightly (results are available under request).

5.3. Global financial crisis

The global financial crisis was associated with dramatic changes in commodity prices. To deal with the global financial crisis we introduce a dummy variable that takes the value 1 from July 2008 to December 2008 and 0 otherwise into equation (1).[17] Results are essentially unchanged from those in Figures 5 and 6 from following this strategy for dealing with the global financial crisis (and are available from the authors).

6. Conclusion

In this paper we investigate the influence of liquidity as it arises from the major developed and major developing economies on commodity and disaggregated prices. The magnitude of the positive effect of positive BRIC liquidity innovations on commodity prices compared to that of positive G3 liquidity innovations on commodity prices is much larger and the disparity grows over time. Positive shocks in BRIC liquidity have much larger effects on energy prices, mineral and metal prices, and raw material prices than do positive shocks in G3 liquidity. It is only on precious metal prices that shocks to G3 M2 have larger effects than shocks to BRIC M2. A positive shock in BRIC M2 is associated with a boom in global industrial production and global tightening in monetary policy as indicated by increases in central bank discount rates. Global industrial production and global interest rates do not respond significantly to innovations in G3 M2.

Results are robust to alternative identification schemes in the structural factor-augmented vector error correction model, to different measurement of global variables, to

[17] Perri and Quadrini (2011) identify the third and fourth quarter of 2008 as being the global financial crises period. To correspond to this analysis we identify July 2008 to December 2008 as being the crisis period. With monthly data a narrower focus is possible. We experiment with September 2008 to November 2008 as being the global financial crises without changing results.

treatment of the global financial crisis, and to variation in lag length. Findings suggest during what Hamilton (2013) refers as a "new industrial age" (1997-2010), characterized by billions of people making the transition from agricultural to industrial activity with increases in real income beyond subsistence levels, increases in liquidity in the major developing countries have much more powerful consequences for global commodity prices than do increase in liquidity in advanced countries.

References

Anzuini, A., Lombardi, M.J., Pagano, P., 2013. The impact of monetary policy shocks on commodity prices. International Journal of Central Banking 9 (3), 119-144.

Bachmeier, L.J., Swanson, N.R., 2005. Predicting Inflation: Does The Quantity Theory Help? Economic Inquiry 43, 570-585.

Barsky R.B., Kilian L., 2004. Oil and the Macroeconomy Since the 1970s. Journal of Economic Perspectives 18, 115-134.

Belke, A., Orth, W., Setzer, R., 2010. Liquidity and the dynamic pattern of asset price adjustment: A global view. Journal of Banking and Finance 34, 1933-1945.

Belke, A., Bordon, I.G., Volz, U., 2013. Effects of Global Liquidity on Commodity and Food Prices. World Development 44, 31-43.

Bernanke, B., 1986. Alternative explanations of the money-income correlation. Carnegie-Rochester series on Public Policy 25, 49-99.

Bernanke, B., Boivin, J., Eliasz, P.S., 2005. Measuring the Effects of Monetary Policy: A Factor-augmented Vector Autoregressive (FAVAR) Approach. Quarterly Journal of Economics 120, 387-422.

Beyer, A., Doornik, J.A., Hendry, D.F., 2000. Constructing Historical Euro-Zone Data. Economic Journal 111, 308-327.

Christiano, L.J., Eichenbaum, M., Evans, C., 1999. Monetary policy shocks: What have we learned and to what end? In: Taylor, J.B., Woodford, M. (Eds.), Handbook of Macroeconomics, Vol. 1A. North-Holland, Amsterdam, pp. 65–148.

D'Agostino, A., Surico, P., 2009. Does global liquidity help to forecast US inflation? Journal of Money, Credit and Banking 41 (2/3), 479-489.

Darius, R., Radde, S., 2010. Can Global Liquidity Forecast Asset Prices? IMF Working Paper 10/196, International Monetary Fund, Washington, D.C.

Dees, S., di Mauro, F., Pesaran, M.H., Smith, L.V., 2007, Exploring the international linkages of the euro area: a global VAR analysis. Journal of Applied Econometrics 22, 1-38.

Dungey, M., Fry-McKibbin, R., Linehan, V., 2013. Chinese Resource Demand and the Natural Resource Supplier. Working paper University of Tasmania. Forthcoming Applied Economics.

Enders, W., 2004. Applied Econometric Time Series, second ed. John Wiley and Sons, New York.

Engle, R.F., Granger, C.W.J., 1987. Cointegration and Error-Correction: Representation, Estimation, and Testing. Econometrica 55, 251-276.

Frankel J.A., 1984. Commodity Prices and Money: Lessons from International Finance. American Journal of Agricultural Economics 66, 560-566.

Frankel J.A., Hardouvelis G.K., 1985. Commodity Prices, Money Surprises, and Fed Credibility. Journal of Money, Credit and Banking 17, 427-438.

Garratt, A., Koop, G., Mise, E., Vahey, S.P., 2009. Real-Time Prediction With UK Monetary Aggregates in the Presence of Model Uncertainty. Journal of Business & Economic Statistics, American Statistical Association 27, 480-491.

Giese, J. V.,Tuxen, C. K., 2007. Global Liquidity, Asset Prices and Monetary Policy: Evidence from Cointegrated VAR Models. Working Paper. University of Oxford, Nuffield College and University of Copenhagen, Department of Economics

Gordon, D.B., Leeper, E.M., 1994. The Dynamic Impacts of Monetary Policy: An Exercise in Tentative Identification. Journal of Political Economy 102, 1228-47.

Hamilton, J.D., 2013. Historical Oil Shocks. In: Parker, R.E., Whaples, R.M., (Eds.), The Routledge Handbook of Major Events in Economic History. New York: Routledge Taylor and Francis Group, pp. 239-265.

Humphreys, D., 2010. The great metals boom: A retrospective. Resources Policy 35, 1-13.

Kilian, L., Hicks, B., 2013. Did Unexpectedly Strong Economic Growth Cause the Oil Price Shock of 2003-2008? Journal of Forecasting 32, 385-394.

Kim, S., 2001. International transmission of US Monetary policy shocks: evidence from VARs. Journal of Monetary Economics 48, 339–372.

Kim, S., Roubini, N., 2000. Exchange rate anomalies in the industrial countries: a solution with a structural VAR approach. Journal of Monetary Economics 45, 561–586.

Koop, G., Pesaran, M.H., Potter, S.M., 1996. Impulse response analysis in nonlinear multivariate models. Journal of Econometrics 74, 119–147.

Perri, F., Quadrini, V., 2011. International Recessions. National Bureau of Economic Research, NBER Working Paper 17201.

Pesaran, M. H., Y. Shin, 1998. Generalized Impulse Response Analysis in Linear Multivariate Models. Economics Letters 58, 17–29.

Radetzki, M., 2006. The anatomy of three commodity booms. Resources Policy 31, 56–64.

Ratti, R.A., Vespignani, J., 2013. Why are crude oil prices high when global activity is weak? Economics Letters 121, 133–136.

Roberts, I., Rush, A., 2010. Sources of Chinese demand for resource commodities. Research Discussion Paper 2010-8, Reserve Bank of Australia.

Sims, C.A., 1980. Macroeconomics and Reality. Econometrica 48, 1-48.

Sims, C.A., Zha, T., 1995. Does monetary policy generate recessions?: Using less aggregate price data to identify monetary policy. Working paper, Yale University, CT.

Sims, C.A., Zha, T., 2006. Does Monetary Policy Generate Recessions?, Macroeconomic Dynamics 10, 231-272.

Stock, J.H., Watson, M.W., 2002. Forecasting using principal components from a large number of predictors. Journal of the American Statistical Association 97, 1167–1179.

Sousa, J., Zaghini, A., 2007. Global Monetary Policy Shocks in the G5: A SVAR Approach. Journal of International Financial Markets, Institutions and Money 17, 403-419.

Swanson, N.R., 1998. Money and output viewed through a rolling window. Journal of Monetary Economics 41, 455-474.

Table 1: Correlation between country-specific and global factors.

Country/Global	Global/Country IR	Global/Country IP	Global/Country CPI
Euro area	0.84	0.61	0.99
US	0.89	0.62	0.99
China	0.62	0.93	0.95
Japan	0.51	0.31	-0.75
UK	0.89	-0.62	0.97
India	0.32	0.92	0.95
Russia	0.68	0.99	0.99
Brazil	0.38	0.97	0.99

Notes: Correlations between country-specific and global factor for the short-term interest rate (IR), for industrial production (IP), and for the inflation rate (CPI) are reported in the columns in Table 1. The global factors are given by first principal components for the global interest rate (*GIR*), global industrial production rate (*GIP*), and global inflation rate (*GCPI*).

Table 2: Granger causality tests 1999:1-2012:12 (log-level).

Null Hypothesis: x does not Granger cause y

Granger test/Lags	1	3	6	12
Commodity price does not Granger cause BRIC M2	0.15	0.35	0.71	1.47
BRIC M2 does not Granger cause commodity prices	3.08***	8.05***	4.71***	3.15***
Commodity price does not Granger cause G3 M2	0.22	0.26	1.53	1.16
G3 M2 does not Granger c causeommodity prices	6.37***	4.52***	2.39***	2.29***

Notes: Variables are in logs. *** Indicates rejection of the null hypothesis at 1% level of significance.

Table 3: Granger causality tests 1999:1-2012:12 (log-first difference).

Null Hypothesis: x does not Granger cause y

Granger test/Lags	1	3	6	12
Commodity price does not Granger cause BRIC M2	0.20	1.76	2.22***	1.51
BRIC M2 does not Granger cause commodity prices	7.04***	4.77***	2.32***	2.27***
Commodity prices does not Granger cause G3 M2	0.24	2.28***	1.34	0.96
G3 M2 does not Granger cause commodity prices	1.17	1.49	1.28	1.41

Notes: Variables are in log-first differences. *** Indicates rejection of the null hypothesis at 1% level of significance.

Table 4: Test for unit roots 1999:1-2012:12.

variables					
Level	ADF	KPSS	First difference	ADF	KPSS
$\log(G3\ M2_t)$	-0.13	1.61***	$\Delta\log(G3\ M2_t)$	-12.4***	0.09
$\log(BRIC\ M2_t)$	2.20	1.62***	$\Delta\log(BRIC\ M2_t)$	2.6*	0.73
$\log(\mathrm{GCPI}_t)$	0.47	1.62***	$\Delta\log(\mathrm{GCPI}_t)$	-8.31***	0.13
$\log(\mathrm{GIP}_t)$	-1.48	1.48***	$\Delta\log(\mathrm{GIP}_t)$	-5.93***	0.08
$\log(Commodities_t)$	-1.52	1.53***	$\Delta\log(Commodities_t)$	-8.86***	0.04
$\log(Energy_t)$	-2.20	1.52***	$\Delta\log(Energy_t)$	-9.70***	0.07
$\log(Agricultural_t)$	-0.66	1.52***	$\Delta\log(Agricultural_t)$	-8.16***	0.11
$\log(Food_t)$	-0.77	1.53***	$\Delta\log(Food_t)$	-8.26***	0.10
$\log(Min.\ and\ metal_t)$	-1.31	1.39***	$\Delta\log(Min.\ and\ metal_t)$	-8.13***	0.07
$\log(Precious\ met._t)$	0.50	1.58***	$\Delta\log(Precious\ metal._t)$	-11.8***	0.21
$\log(Raw\ material._t)$	-1.00	1.45***	$\Delta\log(Raw\ material._t)$	-5.98***	0.07

Notes: The null hypothesis for the ADF test is the variable has a unit root and the null hypothesis for the KPSS test is the variable is stationary. The first difference of the series is indicated by Δ.The lag selection criteria for the ADF is based on Schwarz information Criteria (SIC) and for the KPSS is the Newey-West Bandwidth. ***, **, * indicates rejection of the null hypothesis at 1%, 5% and 10%, levels of significance.

Table 5: VAR Johansen cointegration test summary:

5.1 Cointegration test: logs of commodity prices and money (G3 and BRIC)

Endogenous variables: $\log(Commodity\ prices_t), \log(G3\ M2_t), \log(BRIC\ M2_t)$

Exogenous variables: $\log(Global\ industrial\ production_t), Global\ interest\ rate$

Data Trend:	None	Linear
Test Type	Intercept	Intercept
Trace	1	0
Max-Eig	1	0

5.2 Cointegration test: logs of global inflation (*GCPI*), money (G3 M2 and BRIC M2) and global output (*GIP*).

Endogenous variables: $\log(Commodity\ prices_t), \log(Global\ Money_t)$

Exogenous variables: $\log(Global\ industrial\ production_t), Global\ interest\ rate$

Data Trend:	None	Linear
Test Type	Intercept	Intercept
Trace	2	1
Max-Eig	2	1

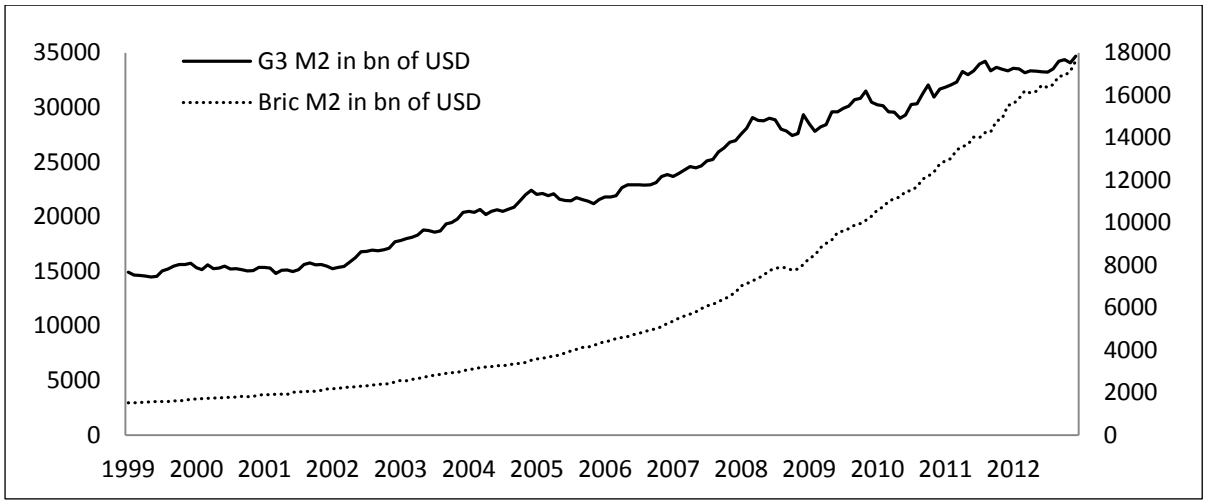
Notes: *Critical values based on MacKinnon-Haug-Michelis (1999). **Selected (0.05 level*) Number of Cointegrating Relations by Model.

Table 6: LR test for over-identified restrictions.

Equation	$\mathrm{LR}\sim\chi^2(q-k)$	Chi-square value	Probability
8	2	7.102	0.0290
9	2	7.273	0.0263
10	3	14.717	0.0021
11	1	6.199	0.0128
12	0	0.811	0.3678

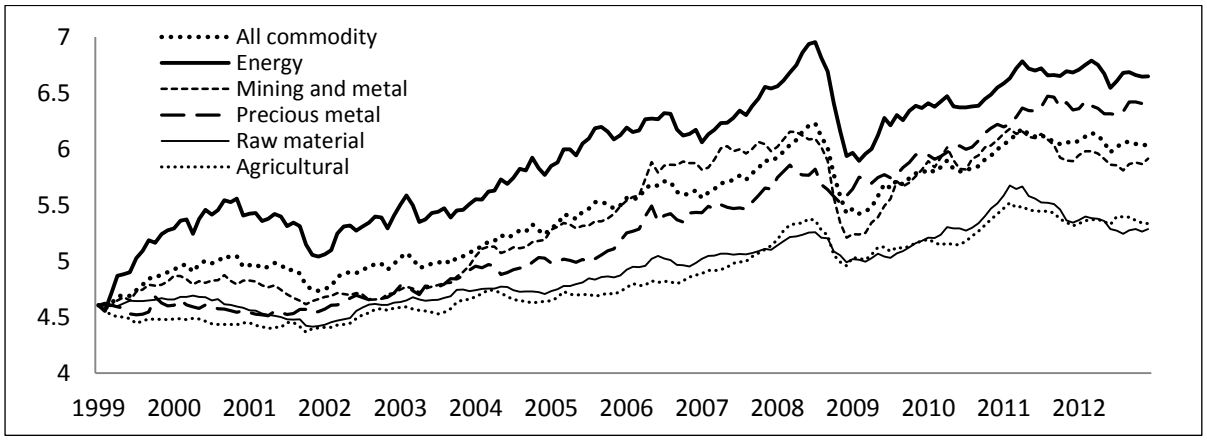
*Null hypothesis: restrictions are valid.
**q is the number of restrictions.

Figure 1: BRIC M2 and G3 M2 in billions of US dollars: 1999:01-2012:12.



Notes: The BRIC countries are Brazil, Russia, India and China. G3 economies are the US, EU and Japan. Data are monthly over 1999:01-2012:12 in billions of US dollars. The scale of the right hand side of Figure 1 is for M2 for the BRIC countries and the scale of the left hand side of Figure 1 is for M2 for the G3.

Figure 2: Log of US dollar commodity price indices: 1999:01-2012:12.



Notes: The US dollar commodity price component indices are for energy commodities, agriculture commodities, mineral and metal commodities, precious metal commodities, and raw materials commodities.

Figure 3a: Principal components estimation of global interest rate: 1999:01 to 2012:12.

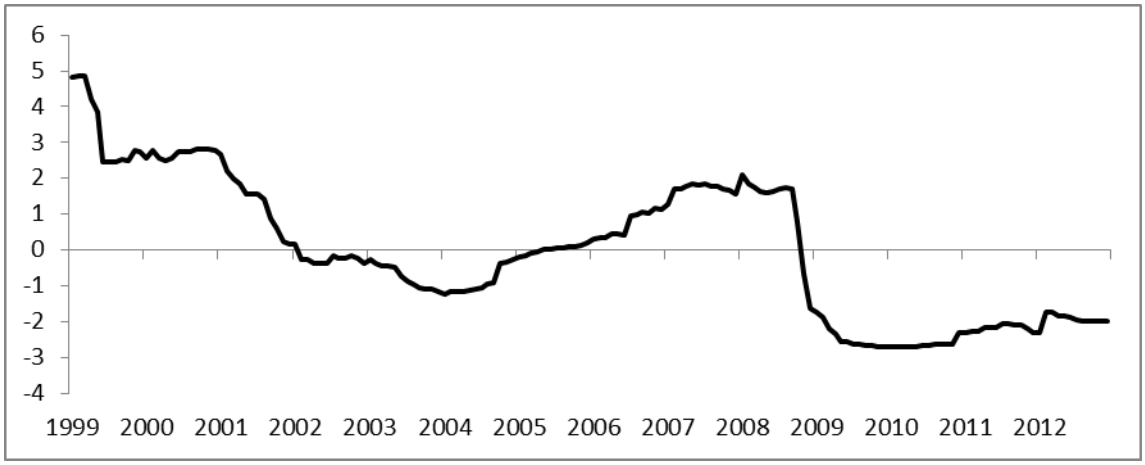


Figure 3b: Principal components estimation of global inflation: 1999:01 to 2012:12.

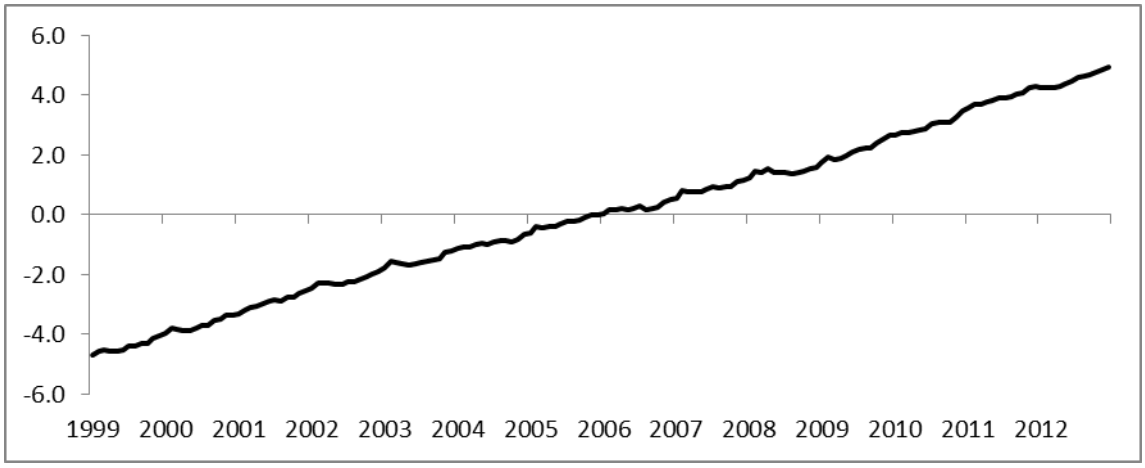
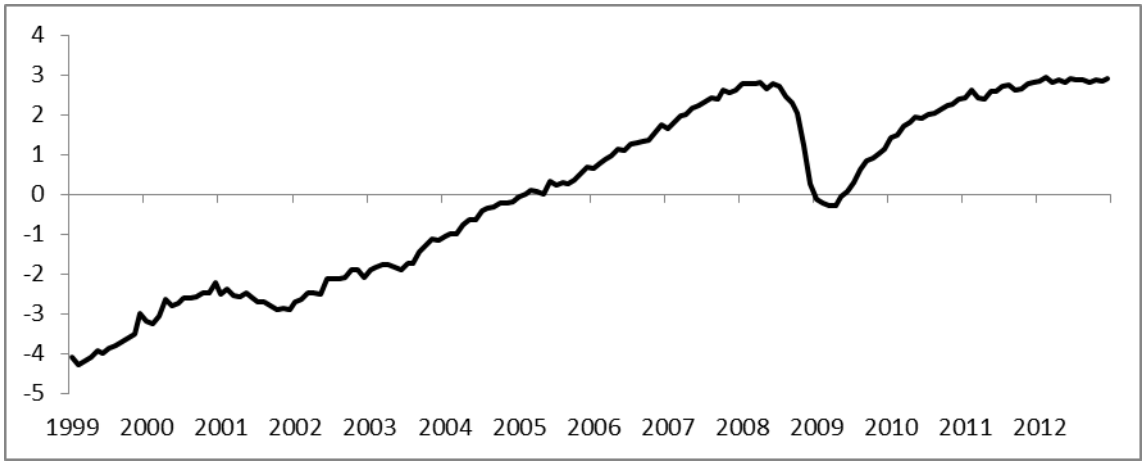
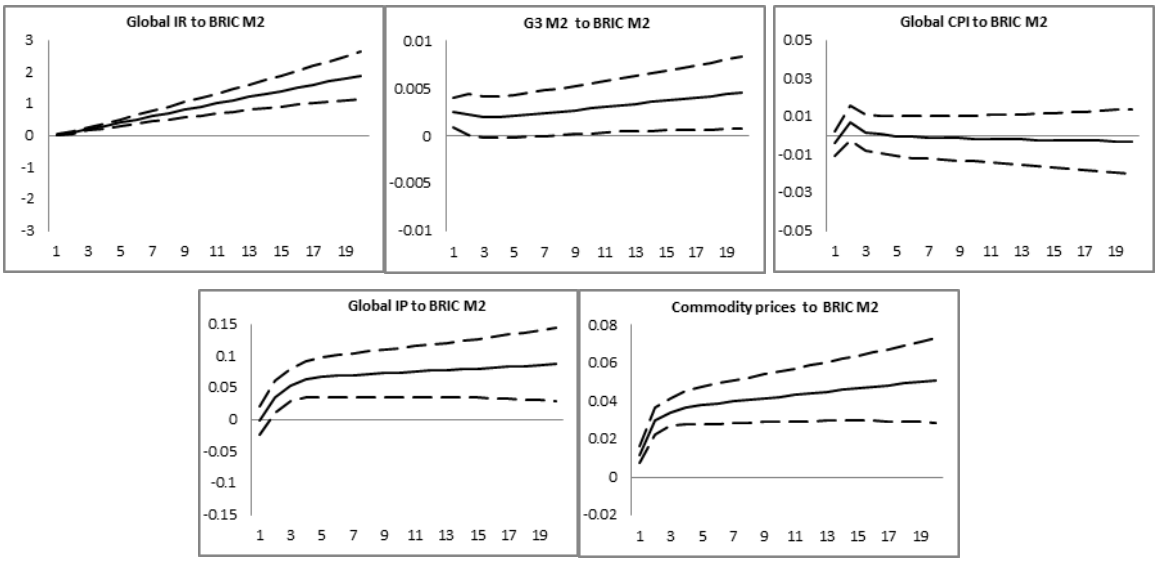


Figure 3c: Principal components estimation of global real output: 1999:01 to 2012:12.



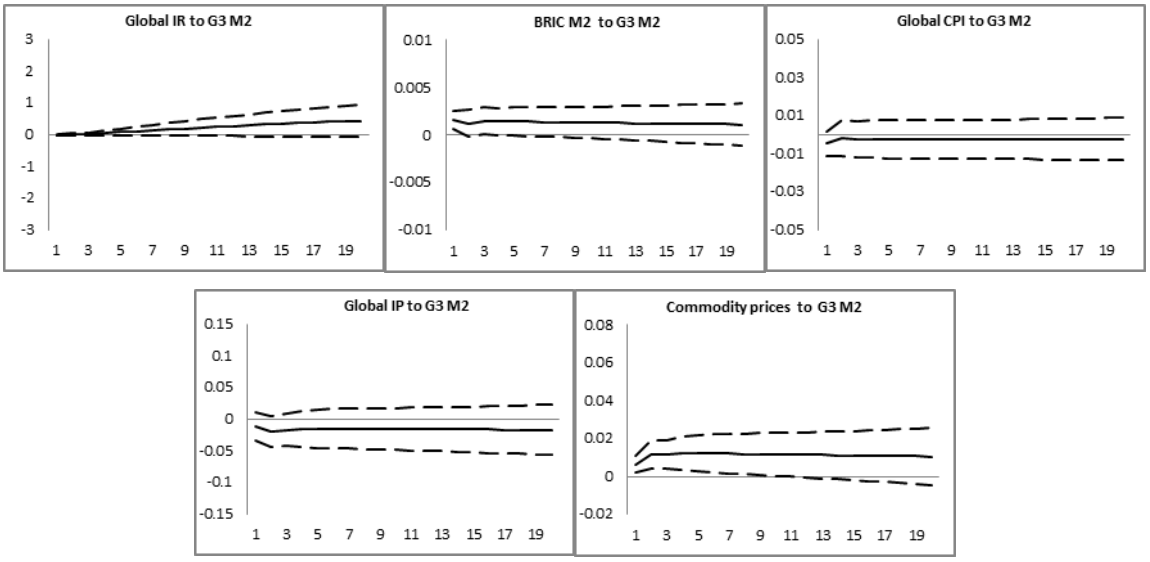
Notes: The principal components of the BRIC and G3 economies' short-term interest rate, industrial production, and inflation are taken to represent global interest rate, global industrial production, and global inflation, respectively.

Figure 4a: One standard deviation generalised cumulative response functions of global variables to shocks in BRIC M2.



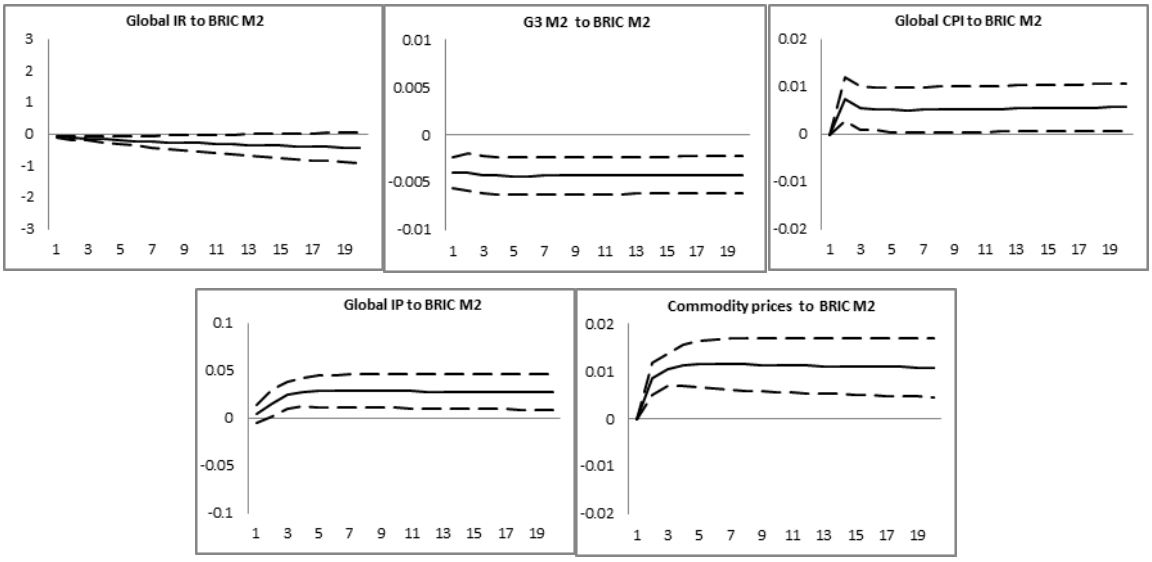
Notes: The one standard deviation generalised cumulative impulse response functions of the global interest rate (Global IR), G3 M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in BRIC M2, based on the SFAVEC model in equations (6)-(7). The global variables are based on principal components.

Figure 5a. One standard deviation generalised cumulative impulse response functions of global variables to shocks in G3 M2.



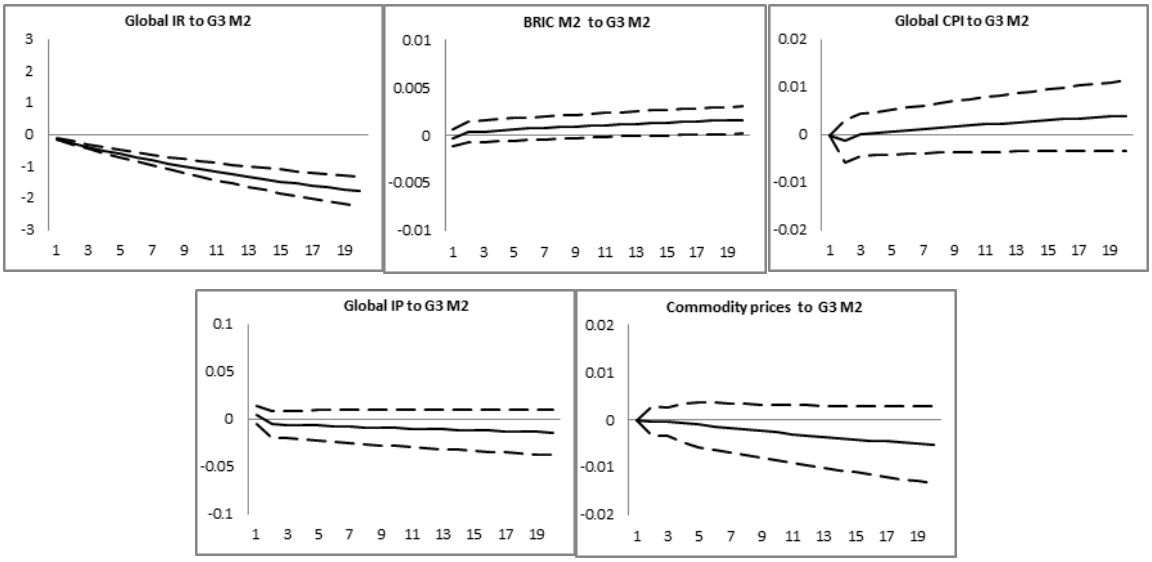
Notes: The one standard deviation generalised cumulative impulseresponse functions of the global interest rate (Global IR), BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in G3 M2, based on the SFAVEC model in equations (6)-(7). The global variables are based on principal components.

Figure 4b: One standard deviation cumulative impulse response functions of global variables to shocks in BRIC M2: Commodity price contemporaneously exogenous.



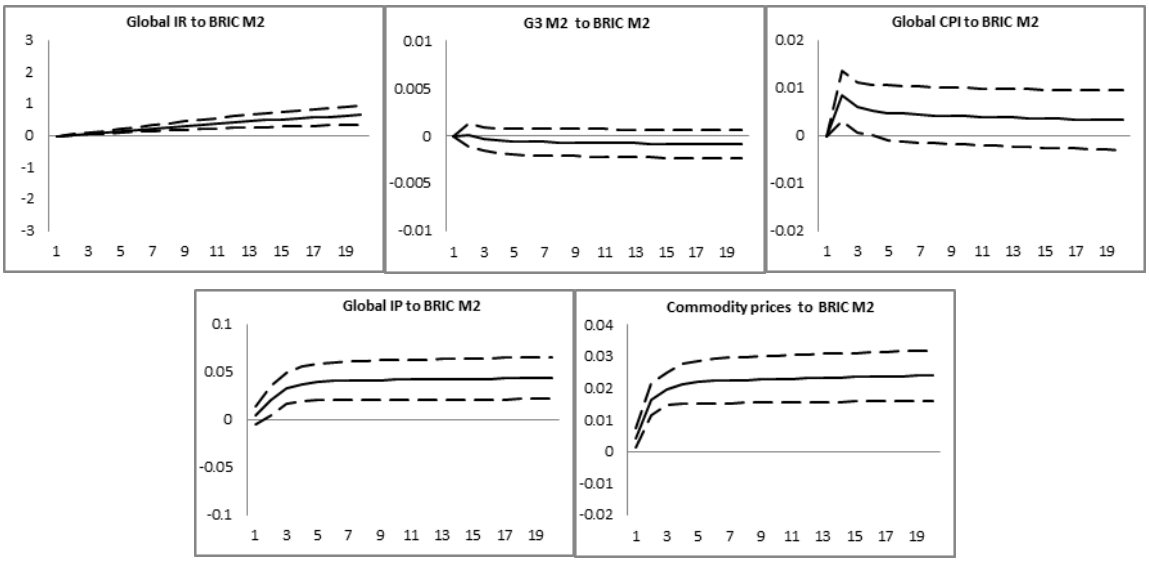
Notes: The one standard deviation cumulative impulse response functions of the global interest rate (Global IR), BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in BRIC M2, based on the SFAVEC model in equations (6)-(8). The global variables are based on principal components.

Figure 5b. One standard deviation cumulative impulse response functions of global variables to shocks in G3 M2: Commodity price contemporaneously exogenous.



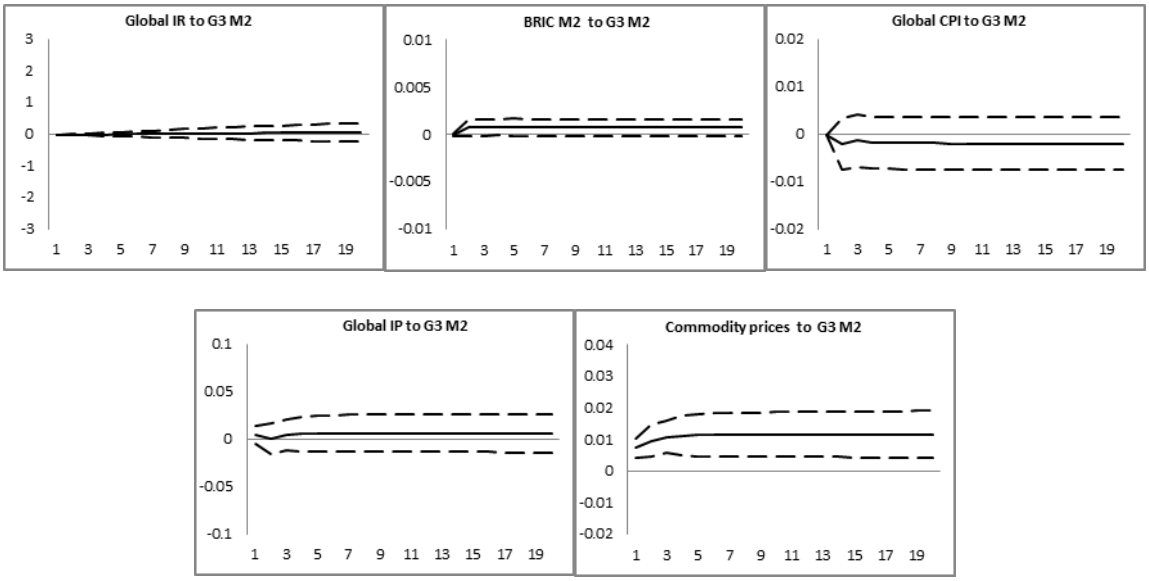
Notes: The one standard deviation cumulative impulse response functions of the global interest rate (Global IR), BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in G3 M2, based on the SFAVEC model in equations (6)-(8). The global variables are based on principal components.

Figure 4c: One standard deviation cumulative impulse response functions of global variables to shocks in BRIC M2: Commodity price contemporaneously endogenous.



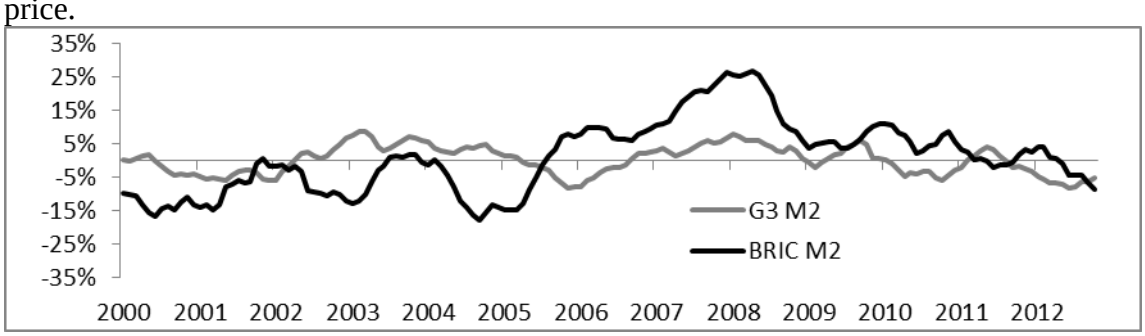
Notes: The one standard deviation cumulative impulse response functions of the global interest rate (Global IR), BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in BRIC M2, based on the SFAVEC model in equations (6), (7) and (12). The global variables are based on principal components.

Figure 5c. One standard deviation cumulative impulse response functions of global variables to shocks in G3 M2: Commodity price contemporaneously endogenous.



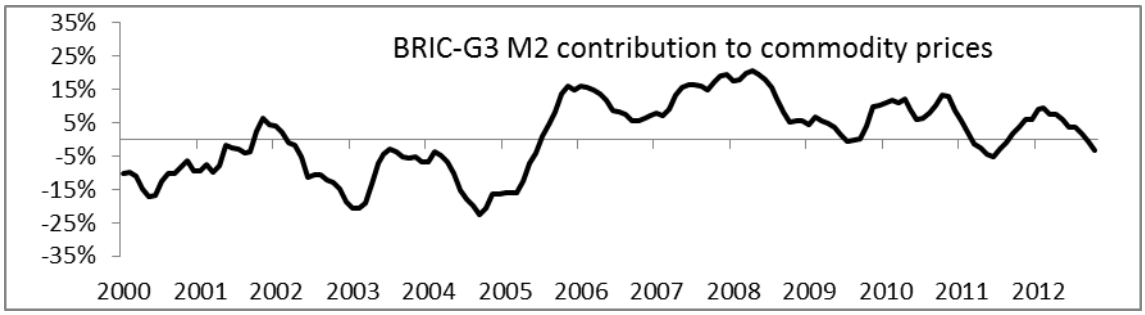
Notes: The one standard deviation cumulative impulse response functions of the global interest rate (Global IR), BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in G3 M2, based on the SFAVEC model in equations (6), (7) and (12). The global variables are based on principal components.

Figure 6a: Cumulative effect of structural shocks to BRIC M2 and to G3 M2 on commodity price.



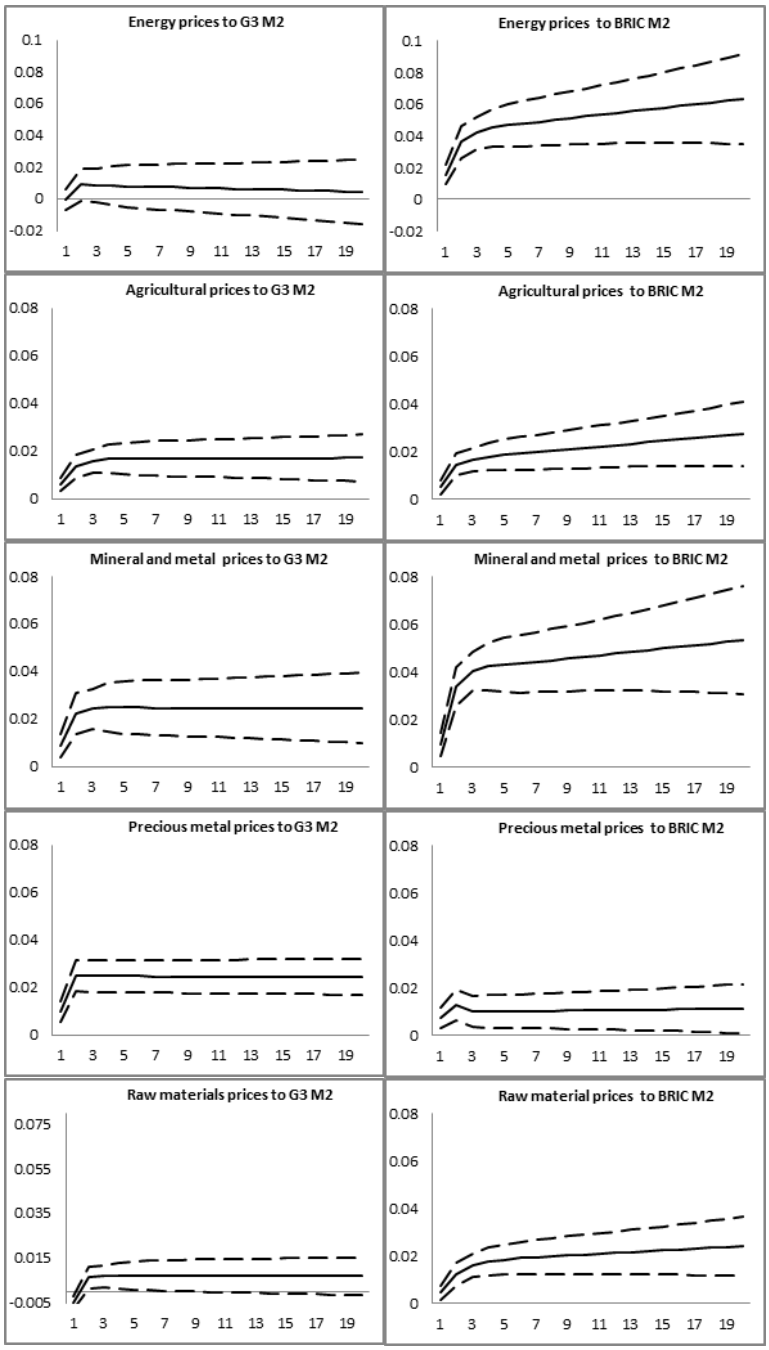
Notes: The cumulative contributions to commodity price of the structural shocks to G3 M2 and to BRIC M2 are obtained from estimating the SFAVEC model in equations (6)-(8). The cumulative contributions of structural shocks to commodity price are the moving average of the last 12 months expressed at an annualized rate.

Figure 6b: Difference in cumulative effect on commodity price of structural shocks to BRIC M2 and G3 M2.



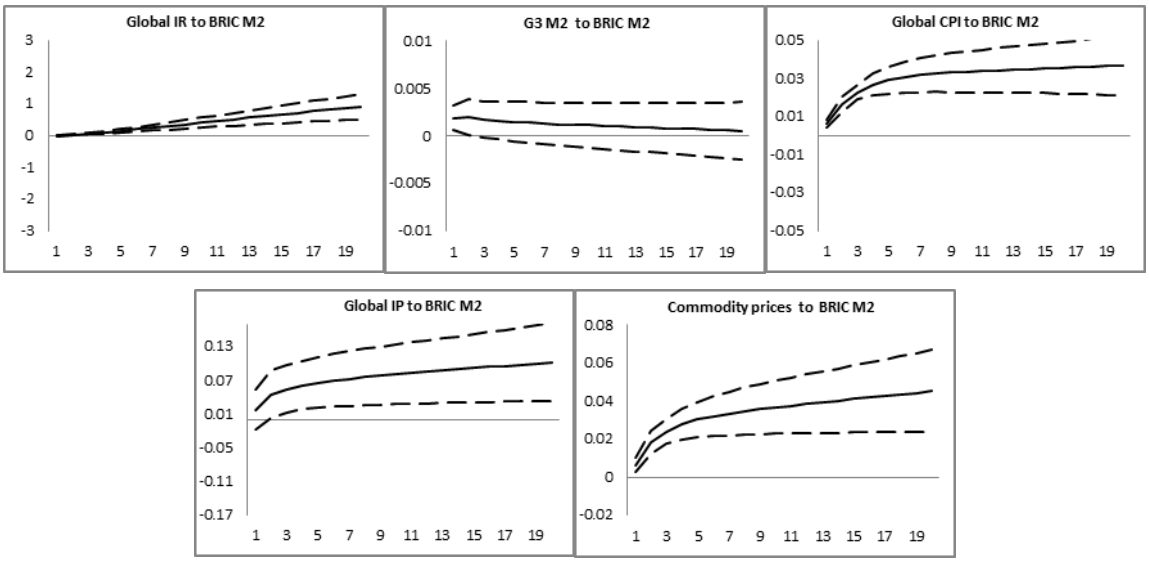
Notes: A positive (negative) value for difference in cumulative effect of structural shocks to BRIC M2 and G3M2 on commodity price, BRIC-G3 M2, indicates larger (smaller) effect of BRICM2 on commodity price than that of G3M2.

Figure 7: One standard deviation generalised cumulative impulse response functions of disaggregated commodity prices to shocks in G3 M2 and BRIC M2.



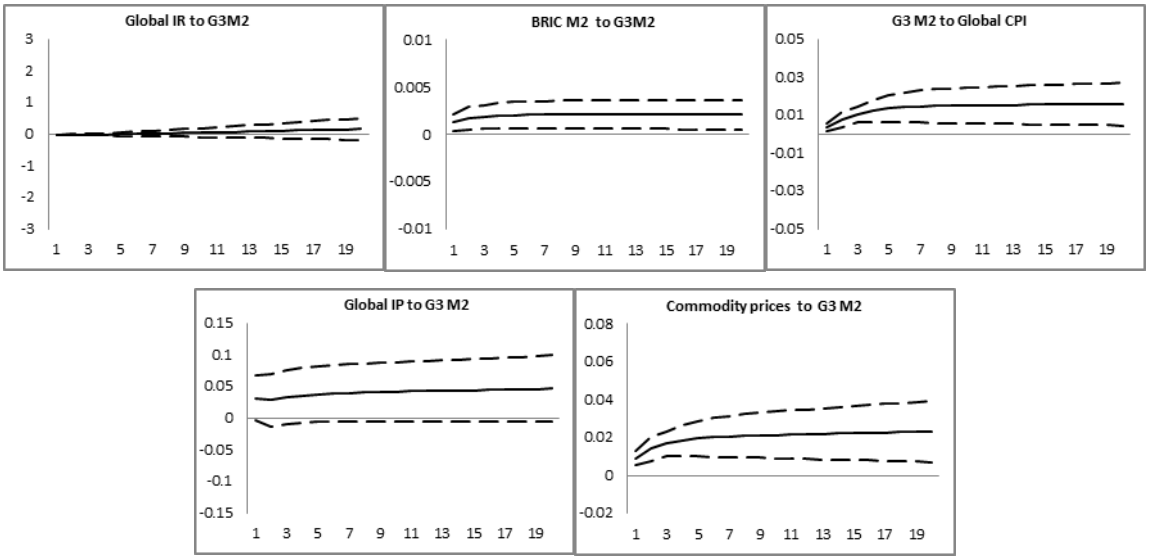
Notes: The one standard deviation generalised cumulative impulse response functions of the commodity price component indices to structural innovations in G3 M2 and in BRIC M2 are shown in first column and in the second column, respectively, based on the SFAVEC model in equations (6) and (7). The commodity price component indices considered are energy commodities, agriculture commodities, mineral and metal commodities, precious metal commodities, and raw materials commodities.

Figure 8: One standard deviation generalised cumulative impulse response functions of global variables to shocks to BRIC M2 (Global variables weighted by nominal GDP).



Notes: The one standard deviation generalised cumulative impulse response functions of the global interest rate (Global IR), G3 M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in BRIC M2, based on the SVEC model in equations (6) and (7) with global variables constructed by summing national variables weighted by relative nominal GDP.

Figure 9: One standard deviation generalised cumulative impulse response functions of global variables to shocks to G3 M2 (Global variables weighted by nominal GDP).



Notes: The one standard deviation generalised cumulative impulse response functions of the global interest rate (Global IR), BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in G3 M2, based on the SVEC model in equations (6) and (7) with global variables constructed by summing national variables weighted by relative nominal GDP.

Appendix A: The FAVAR model.

This Appendix briefly describes the use of principal component estimation in conjunction with a VAR model. Let X_t denote a $N\ x\ 1$ vector of time series variables, Y_t a vector $M\ x\ 1$ observable time series variables that constitute a subset of X_t and F_t a $k\ x\ 1$ vector of unobserved factors that capture most of the information contained in X_t. Bernanke et al. (2005) describe the joint dynamic of (F_t, Y_t) as follows:

$$\begin{bmatrix} F_t \\ Y_t \end{bmatrix} = \Phi * (L) \begin{bmatrix} F_{t-1} \\ Y_{t-1} \end{bmatrix} + v_t \leftrightarrow \Phi(L) \begin{bmatrix} F_t \\ Y_t \end{bmatrix} = v_t, \qquad (13)$$

where $\Phi(\mathrm{L}) = \mathrm{I} - \Phi * (\mathrm{L})\mathrm{L} = \mathrm{I} - \Phi_1 L - \cdots - \Phi_d L^d$ is a lag polynomial of finite order d in the lag operator L, $\Phi_j\ (j = 1, \ldots, d)$ is the coefficient matrix and v_t is an error term with zero mean and covariance matrix Q. Equation (13) is VAR model, which includes both observable and unobservable variables. We can assume that the relation between the 'informal' time series X_t, the observed variables Y_t and the factor F_t can be summarised in the following representation of a dynamic factor model:

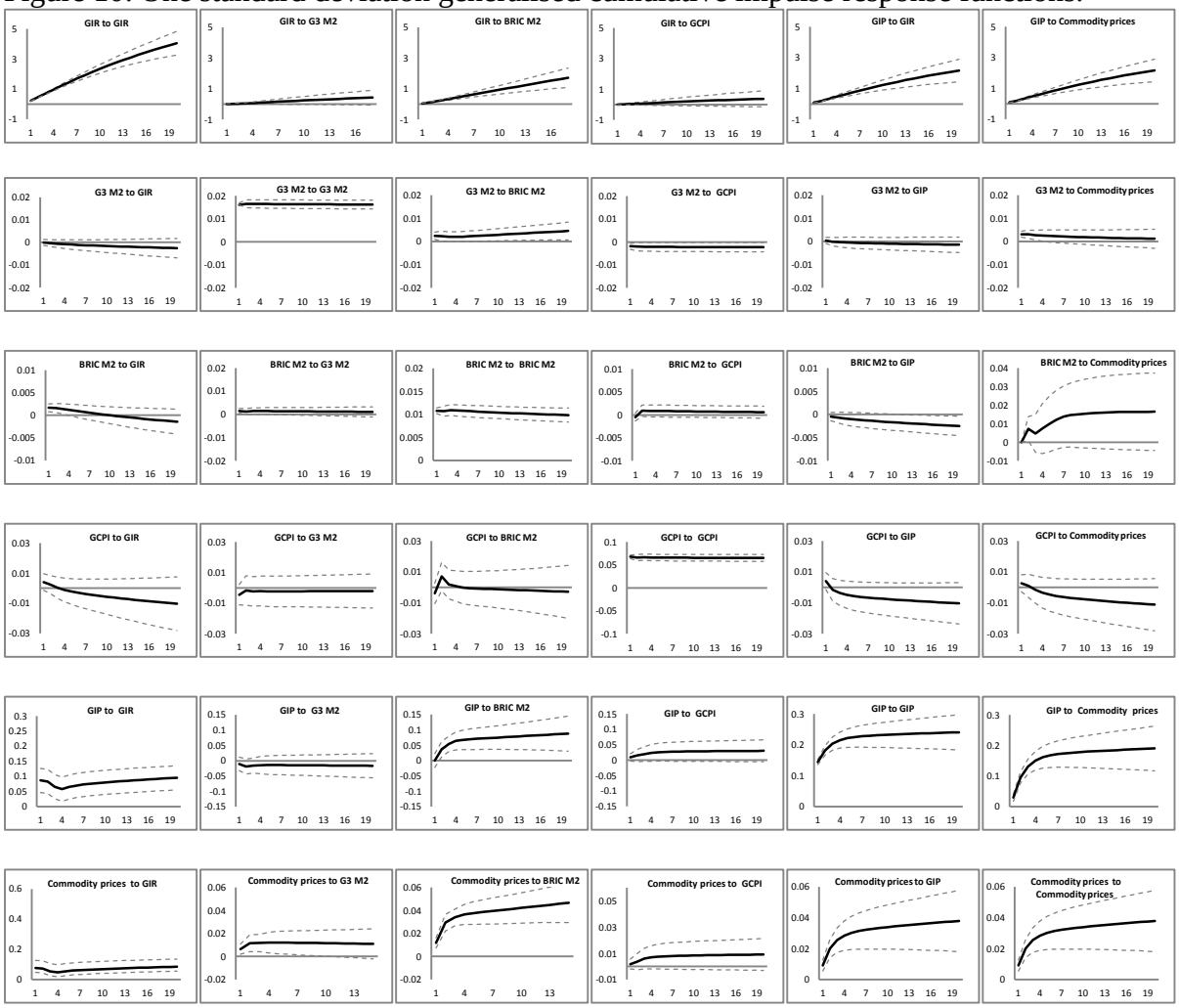
$$X_t = \Lambda^f F_t + \Lambda^y Y_t + e_t \qquad (14)$$

Where Λ^f is a N x k matrix of factor loadings, Λ^y is N x M and e_t is the vector of N x 1 error cross-sectionally and correlated and with mean zero. Following Stock and Watson (2002) X_t does not depend on the lagged values of F_t. Since we assume that $\mathrm{M} + \mathrm{k} \ll \mathrm{N}$, the amount of information that can be handled in the FAVAR increases significantly in comparison to standard VAR model.

Appendix B: One standard deviation generalised cumulative impulse response functions.

The one standard deviation generalised cumulative impulse response functions of the variables in the SFAVEC model in equations (6)-(8) to shocks to all the variables are reported in Table 10.

Figure 10: One standard deviation generalised cumulative impulse response functions.



Notes: The one standard deviation generalised cumulative impulse response functions of the global interest rate (Global IR), G3 M2, BRIC M2, global CPI (Global CPI), global industrial production (Global IP), global commodity price (Commodity prices) to structural innovations in all the endogenous variables in the SFAVEC model in equations (6)-(8). The global variables are based on principal components.