

INFORMATION PROPAGATION BETWEEN EQUITY MARKETS: THE CASE OF BRICT

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Abstract

The overall degree of international equity market connectedness has gradually increased over the past two decades. Reflecting the shift in global economic power, it has been recently suggested to pay more attention to the BRICT (Brazil, Russia, India, China, and Turkey) countries. Building on, and methodologically adding to, an established approach to market connectedness, this study makes an effort to assess the connectedness of these markets, together with developed equity markets. Our methodological contributions consist of the analysis of information propagation, based on Markov chains and information entropy. We find that the BRICT countries' degree of equity market connectedness has increased substantially since the early 2000s, with Brazil and China being the most and least connected equity market, respectively. Furthermore, we find an “informational divide” among BRICT markets insofar as the magnitude of market repercussions of a hypothetical shock within BRICT still depends on the origin of the shock, which is in contrast to western developed markets. We also find that the amount of news produced by the BRICT markets is getting more uniform and approaches the level of developed markets.

Key words: Connectedness of equity markets; spillovers; information propagation; market entropy; BRICT countries

1 Introduction

In the wake of his step in to the head of economic research post at Goldman Sachs in 2001, Jim O'Neill coined the acronym BRIC to refer to the emerging economies (Brazil, Russia, India and China) with high potential to shift the global competitive advantage away from the developed G7 economies towards the developing world in the coming decades (see O'Neill [6], Mpoyi [5]). Even though the “catchy name”¹ BRIC was first formulated on paper by an investment banker mostly as a marketing tool to point out lesser-known markets to investors, the BRIC group has become a political reality and those countries have started cooperating.

BRIC countries do not act as a political alliance, but BRIC foreign ministers have met annually since 2006. The four countries had their first formal summit at the presidential level in 2009, and they meet annually since then. In 2010, South Africa joined the BRIC countries converting the group to BRICS. Together, BRICS represent 43% of the world's population and

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¹Knowledge@Wharton, [4].

17% of trade with the infrastructure needs amounting to \$4.5tn over the next five years². At the end of the 2013 summit, the BRICS countries declared to have agreed to set up a \$100-billion currency crisis fund and a new development bank³. Though the details are not finalized yet, the new development bank is perceived “... as an alternative to the World Bank, although it’s still a long way from meeting that challenge”⁴.

It has been argued that the BRIC does not constitute a homogeneous block, neither with respect to political systems (democratic in India and Brazil, authoritarian in Russia and China)⁵ nor with respect to economic structure (with China and India among the world’s dominant suppliers of manufactured goods and services, respectively, and Brazil and Russia as similarly dominant suppliers of raw materials). Nevertheless, it is not only because of their their potential for a somewhat “symbiotic relationship”⁶ — higher demand for raw materials in India and China boosting the GDP of Russia and Brazil — that it seems reasonable to expect an increasing level of connectedness among BRIC equity markets and also with equity markets in the developed world. For example, to assess the return and volatility spillovers of BRIC countries, Bhar and Nikolova [1] study the level of integration and the dynamic relationship between the BRIC countries and the world, and they find that India showed the highest level of regional and global integration among the BRIC countries, followed by Brazil and Russia and lastly by China.

The present investigation uses the daily stock market data for the BRICs and Turkey, the latter representing the N-11 economies.⁷ We use the MSCI World stock market index to represent the developed markets in the world.

Our approach to measuring equity market connectedness is based on forecast error variance decomposition, see Diebold and Yilmaz [2, 3], and extensions of this approach to assess the propagation of information across markets, see Schmidbauer, Rösch and Uluceviz [8]. The purpose of the present paper is to investigate whether evidence can be found for the following hypotheses:

H1: In view of more regular BRICT member contacts on a political level since 2006 and the prospect of the establishment of a joint development bank, we hypothesize that levels of connectedness among BRICT markets and the developed markets have been increasing since 2006, and so have levels of spillovers from and to those markets.

H2: We also hypothesize that elevated levels of connectedness make the origin of hypothetical shocks, or news, irrelevant for further repercussions of these shocks in the system of BRICT markets — in spite of their heterogeneity.

H3: With BRICT markets becoming more and more attractive to a wider range of investors, these markets become more institutionalized, enabling them to process upcoming information more efficiently; therefore we hypothesize that the level of day-to-day information gain observed in these markets will gradually approach that of developed equity markets.

Section 2 casts a glance at the data used in the present investigation. Section 3 briefly outlines the theory behind assessing connectedness and information propagation. Empirical findings are presented in Section 4; Section 5 concludes.

²Guardian.co.uk, 2013-03-28.

³Bloomberg, 2013-03-28.

⁴Foreignpolicy.com, 2013-04-09.

⁵Nouriel Roubini, see livemint.com, 2013-04-25.

⁶Economist, 2008-04-21.

⁷In 2005, Goldman Sachs identified the “Next 11” (N-11) after BRIC, which — owing to their economic and political conditions — could greatly impact the global economy. The N-11 countries include Bangladesh, Egypt, Indonesia, Iran, Korea, Mexico, Nigeria, Pakistan, Philippines, Turkey and Vietnam; see Goldmansachs.com, accessed on 2013-05-10.

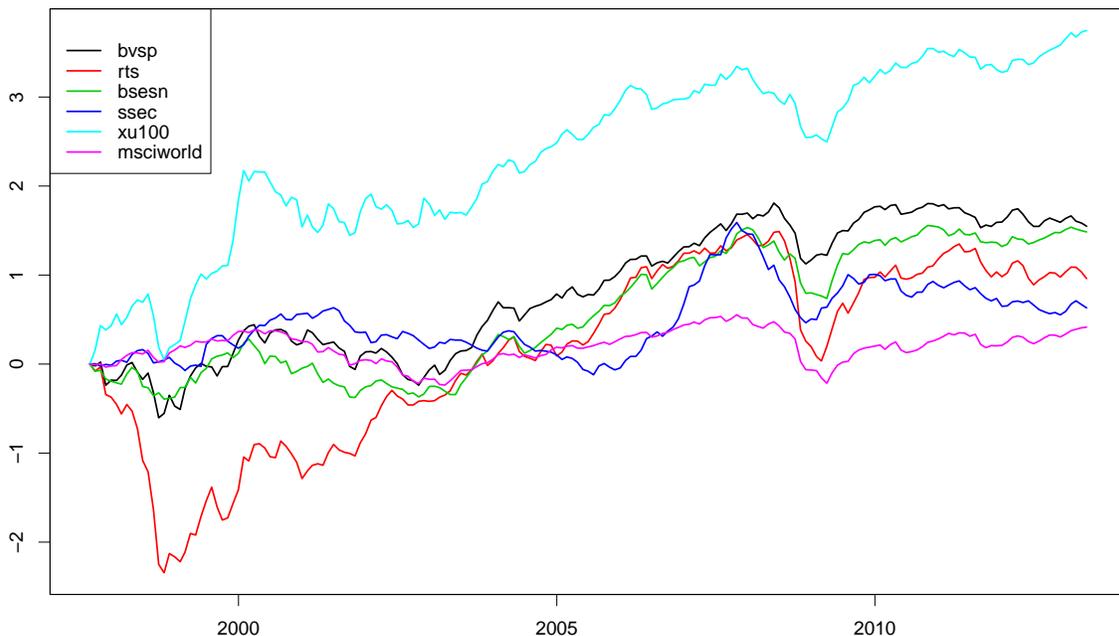


Figure 1: Stock indices, log price series of monthly averages

2 Data

Daily quotations of stock indices from five countries (“BRICT”) are used, namely: bvsp (São, Brazil), rts (Moscow, Russia), bsesn (Mumbai, India), ssec (Shanghai, China), and xu100 (Istanbul, Turkey). The msci world stock market index is included as an “anchor” to reflect spillovers from and to developed markets. The data set extends from July 1997 to April 2013 (4160 observations). Figure 1 gives an impression of the development of average monthly levels, where a log scale on the ordinate axis was found useful. The level series are normed such that “August 1997 \equiv 1”

3 Measuring spillovers and shock repercussions

This study is based on forecast error variance decompositions (fevds), obtained from successively fitting vector autoregressive models to windows of return data. This methodology can be briefly outlined as follows:

Using a moving window of 100 days,

1. fit a standard VAR (vector autoregressive) model to the series.
2. Establish a 5-day-ahead forecast.
3. Decompose the error variance of the forecast for each component with respect to shocks from the same or other markets on day t (the last day in the window). To circumvent the undesirable dependence on the ordering of markets, a generalized fevd is used, proposed by Pesaran and Shin [7].
4. Following Diebold and Yilmaz [2], for each market, arrange the fevd in the so-called

spillover matrix:

| | | from (day t) | | | | | |
|-------------------|-----------|-----------------|-----|-------|------|-------|-----------|
| | | bvsp | rts | bsebn | ssec | xu100 | msciworld |
| to (day $t + 5$) | bvsp | □ | ■ | ■ | ■ | ■ | ■ |
| | rts | ■ | □ | ■ | ■ | ■ | ■ |
| | bsebn | ■ | ■ | □ | ■ | ■ | ■ |
| | ssec | ■ | ■ | ■ | □ | ■ | ■ |
| | xu100 | ■ | ■ | ■ | ■ | □ | ■ |
| | msciworld | ■ | ■ | ■ | ■ | ■ | □ |

(1)

Each row thus sums up to 1 (or 100%) and provides a breakdown of the forecast error variance of the corresponding stock index return with respect to shock origins in terms of percentages. Each entry in the spillover table is called a directional spillover. Schematically, the spillover index is then obtained as

$$\frac{\sum \blacksquare}{\sum \blacksquare + \sum \square}. \quad (2)$$

The network structure of the spillover matrix with respect to the propagation of shocks lends itself to a broader perspective, using concepts borrowed from population and Markov chain theory, as well as from information theory. Again, we give only a very short outline of the concepts used below; for a more comprehensive presentation, see Schmidbauer, Rösch and Uluceviz [8]:

1. Let \mathbf{M}_t denote the spillover matrix for day t . A hypothetical shock (“news”, “information”) of unit size to market i on day t can be denoted as $n_0 = (0, \dots, 0, 1, 0, \dots, 0)'$, where 1 is in the i -th component of n_0 . We assume that the propagation of this shock across the markets within day t will take place in short time intervals of unspecified length according to

$$\mathbf{n}_{s+1} = \mathbf{M}_t \cdot \mathbf{n}_s, \quad s = 0, 1, 2, \dots \quad (3)$$

(where step $s = 0$ initializes the recursion). The index s in equation (3) therefore denotes a hypothetical step in information flow. Moreover, assuming that information flow across markets can proceed instantly, with spillover conditions persisting throughout day t , it makes sense to investigate steady-state properties (as $s \rightarrow \infty$) of the model defined by equation (3), which can be discussed in terms of the eigenvalue structure of the matrix \mathbf{M}_t . The corresponding left eigenvector \mathbf{v}_t , satisfying

$$\mathbf{v}'_t = \mathbf{v}'_t \cdot \mathbf{M}_t, \quad (4)$$

permits a discussion of “propagation values” of markets: The propagation value of a market renders the value of a shock which comes from the market as seed for future shock impact variability. A higher propagation value of a market on day t means that news originating from this market will have stronger repercussions across the system of markets considered.

2. Apart from measuring the strength of market repercussions of a shock, propagation values can also be interpreted as stationary distribution of a Markov chain defined on the basis of a spillover matrix. As given, a spillover matrix is not a suitable transition matrix, because its *rows* (the *to* part), and not its *columns* (the *from* part) sum up to 1, to the effect that such a Markov chain would be running backward in time. Time can be reversed by applying the transformation

$$\mathbf{P}_t = \mathbf{V}_t^{-1} \cdot \mathbf{M}'_t \cdot \mathbf{V}_t \quad (5)$$

(see Tuljapurkar [9]), where the diagonal matrix \mathbf{V}_t contains the left eigenvector \mathbf{v}_t (corresponding to eigenvalue 1) of \mathbf{M}_t , and after re-scaling:

$$\pi'_s = \frac{\mathbf{n}'_s \cdot \mathbf{V}_t}{\mathbf{n}'_0 \cdot \mathbf{v}_t},$$

the Markov chain equation

$$\pi'_{s+1} = \pi'_s \cdot \mathbf{P}_t, \quad s = 0, 1, 2, \dots, \quad (6)$$

emerges. This Markov chain can be interpreted as follows: On day t , the initial location of a shock in the system is given by π_0 (a unit vector). The shock moves through the system according to equation (6). It can be shown that the stationary distribution of shock location is given by the vector of propagation values, which in this sense represents the “information equilibrium” or “news balance” among markets on that day.

3. How much information is produced by the system of markets from day to day? The question of “information gain” from today’s to tomorrow’s news balance among markets can be answered by applying the concept of Kullback-Leibler divergence (also known as Kullback-Leibler information criterion, KLIC), which measures the entropy of day t with respect to day $t-1$, of the propagation values belonging to day t and day $t+1$. For details, see Schmidbauer, Rösch and Uluceviz [8].

4 Empirical findings

An assessment of market connectedness can now be obtained by applying the methodology outlined in Section 3.

4.1 Overall and directional spillovers

Overall spillovers (that is, the daily time series of spillover index values) is displayed in Figure 2, together with a smoothed version. The latter hit the lowest mark in the period considered in December 1999 and was almost equally low again in March 2003. Its subsequent increase gained momentum in early 2005, increasing steadily to about 50% in September 2007. It currently (early May 2013) stands at about 27%.

For the purpose of comparison, it may be helpful to consider spillover levels in a system of four developed equity markets, namely dji (New York, USA), fchi (Paris, France), gdaxi (Frankfurt, Germany) and n225 (Tokyo, Japan), for the same period of time. The (similarly smoothed) spillover series reached its lowest level of 36% in February 2000, and its highest level of more than 60% in November 2008.

Directional spillovers, as shown in Figures 3 (spillovers from a given market to others, that is: off-diagonal column sums of matrix (1)) and 4 (spillovers to a given market from others, that is: off-diagonal row sums of matrix (1)) provide more insight into the way spillovers have increased. China remains secluded in the sense that very little spillover is coming from China; it is a net receiver in the sense that spillovers going to China outweigh those coming from China, and the gap between incoming and outgoing spillovers has even been widening from 2005 to 2012. The other net receivers are India and Turkey; Brazil is a net sender. Russia has been a net sender since about mid-2010. As was to be expected, msci-world is also a net sender. — In summary, it can be said that hypothesis $H1$ is confirmed.

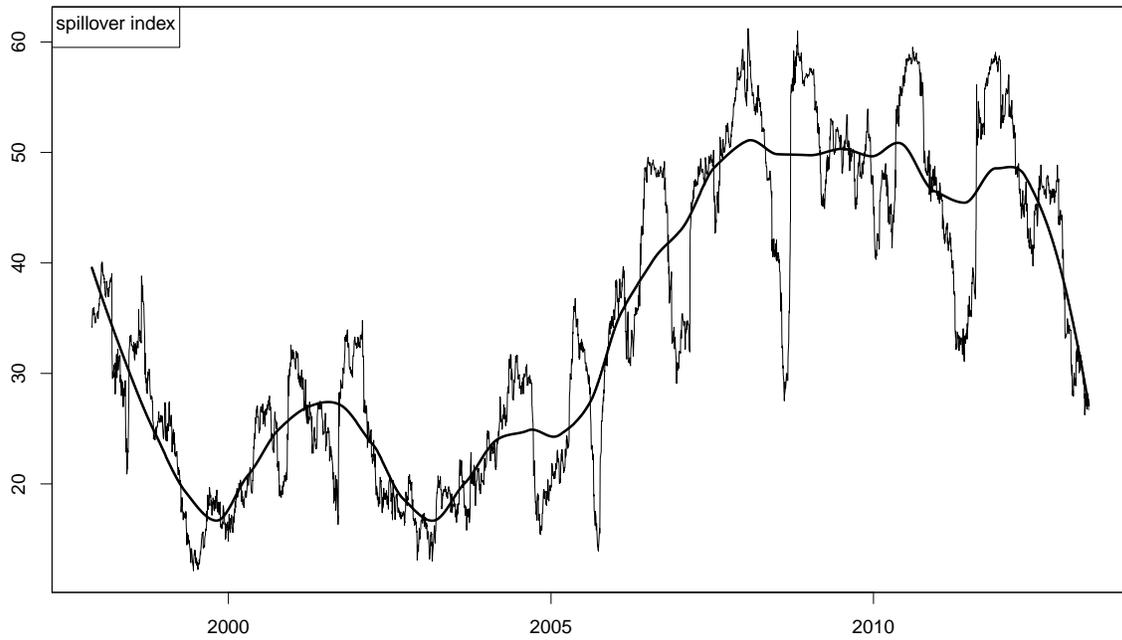


Figure 2: Spillovers index, the case of BRICT

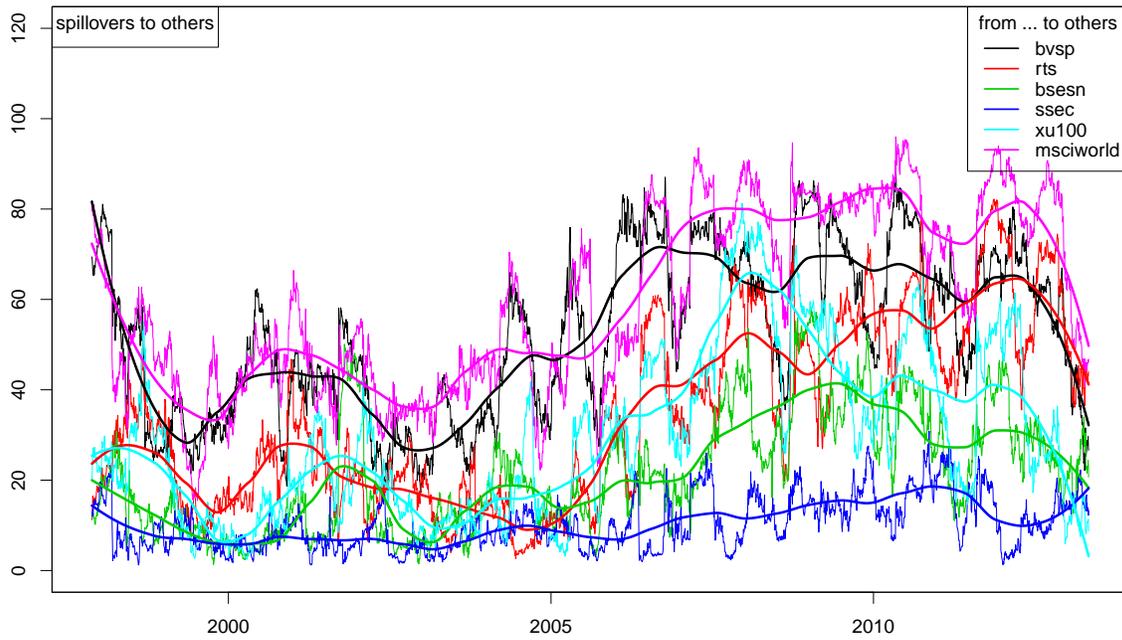


Figure 3: Spillovers from a market to others, the case of BRICT

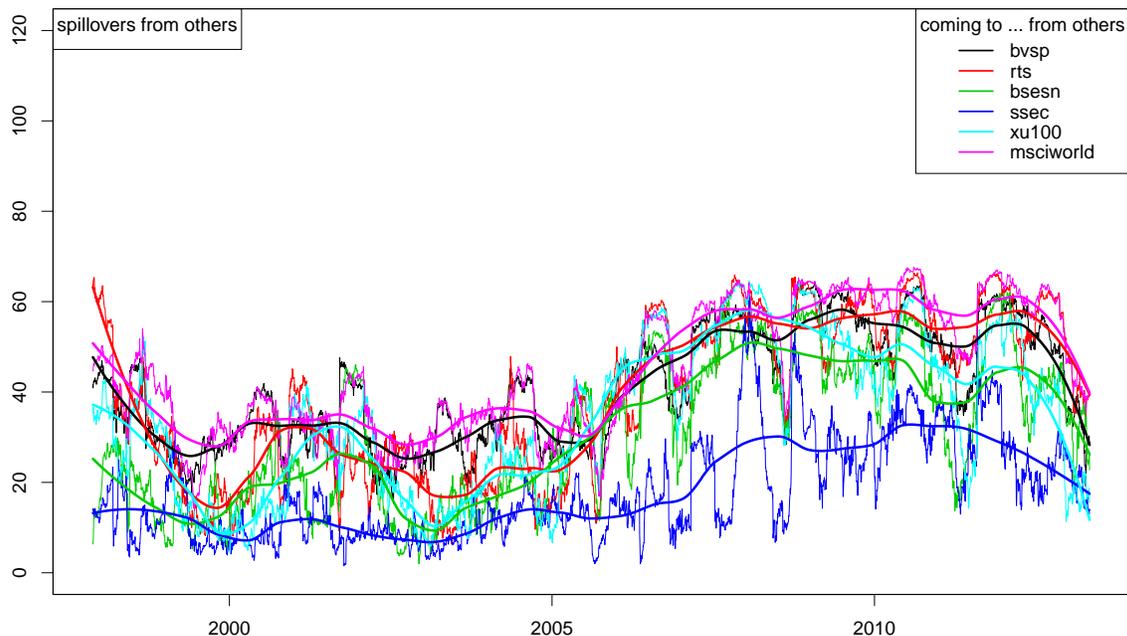


Figure 4: Spillovers to a market from others, the case of BRIC

4.2 Propagation values

Propagation values quantify the strength of repercussions in the system of markets when a hypothetical shock originates from one of the markets. Figure 6 shows that propagation values of dji, fchi, and gdaxi have been in a narrow corridor since about 2001 when considering again the system of four markets (together with n225): market repercussions will be more or less equally strong, irrespective of the origin of the shock (the US, French, or German equity market).

A similar phenomenon has not occurred among the BRIC countries, as is obvious from Figure 5: throughout the period considered, bvsp has the highest propagation value, recently approached by rts, and surpassing Turkey in spite of the latter’s soaring stock market. The situation, however, appears to be far from being as settled as in the case of the four developed markets. Our hypothesis *H2* is thus not confirmed: The origin of a shock, when considering the BRIC economies, does matter for the further propagation of information, with different weight attached to markets at different points in time, and no sign of convergence in sight as yet. This observation can be described as an “informational divide” between the BRIC countries. In contrast, Figure 6 shows that no informational divide exists between France and Germany.

4.3 Day-by-day information production

As shown in Figure 7, the time series of KLIC values (the relative market entropy on a daily basis) has evolved such that (i) KLIC spikes have gradually become shorter, (ii) the density of points for a given KLIC level has decreased. Indeed, the BRIC KLIC series has become more similar to the KLIC series characterizing developed markets (Figure 8), providing evidence in favor of our third hypothesis *H3*.

5 Summary and conclusions

The present study investigates the amount of connectedness of equity markets in BRIC countries with daily closing quotes of their most important stock index values as empirical basis, and

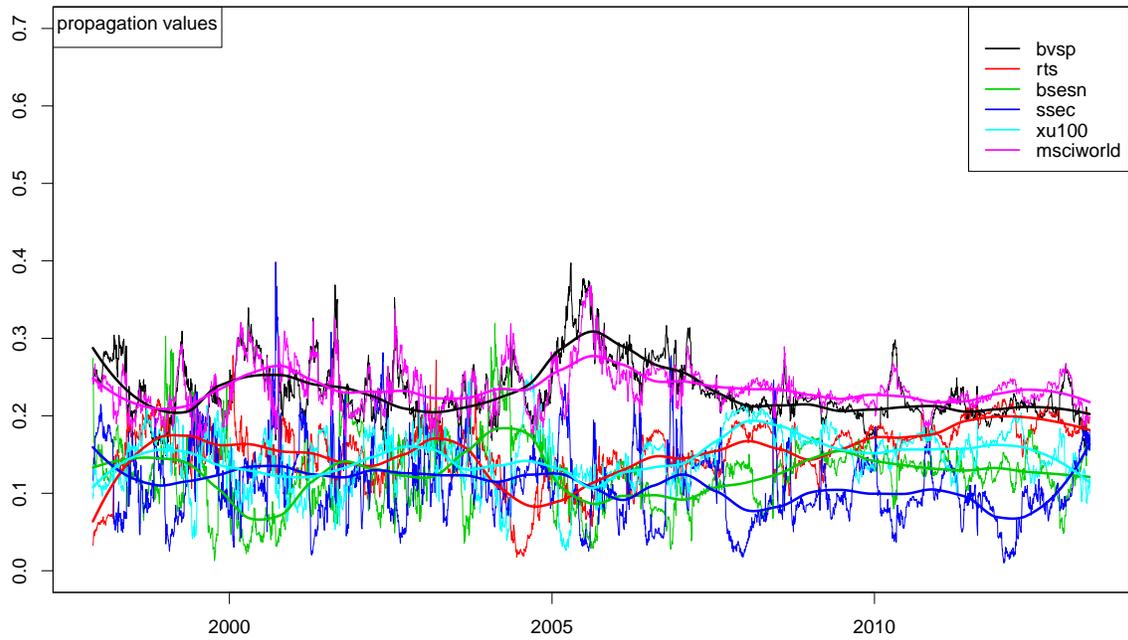


Figure 5: Propagation values, the case of BRICT

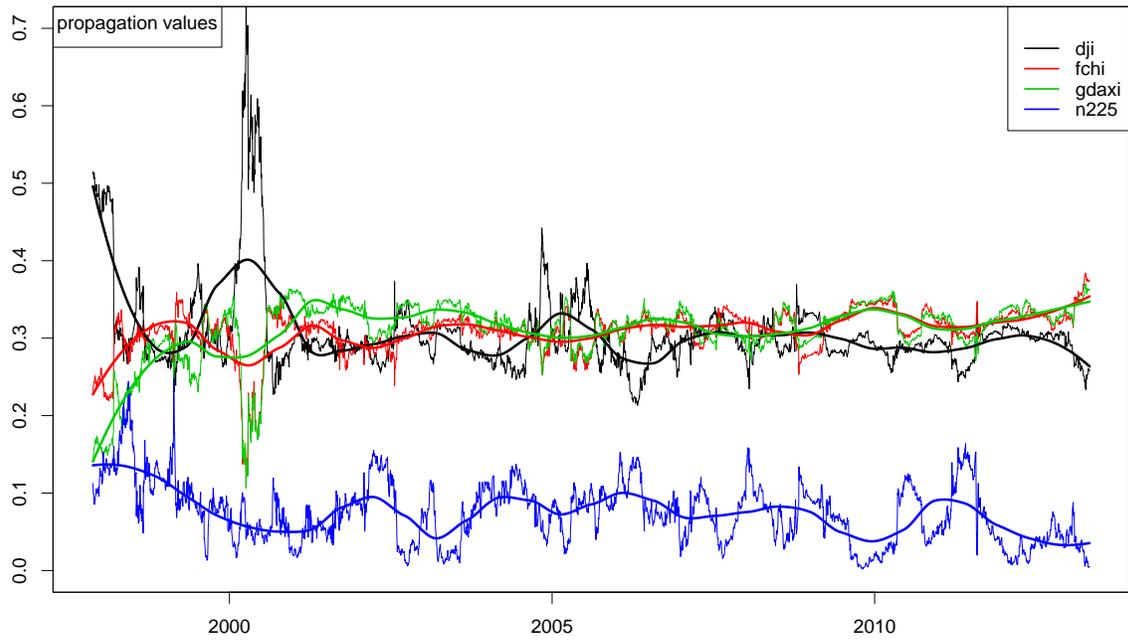


Figure 6: Propagation values, the case of dji, fchi, gdaxi, n225

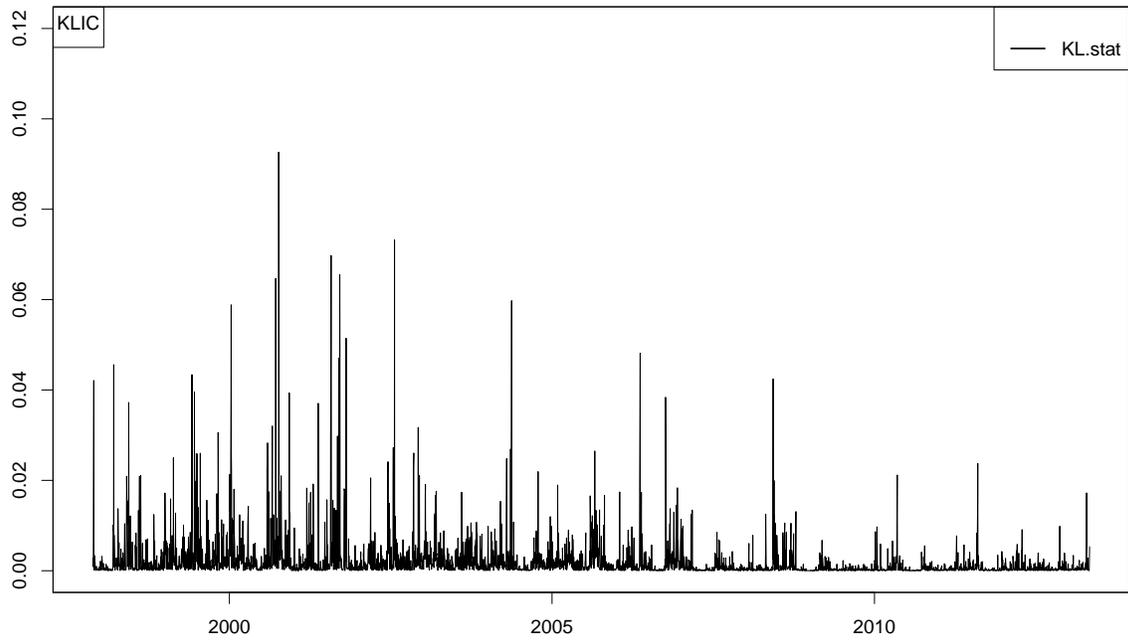


Figure 7: KL distance, daily stationary distributions, the case of BRICT

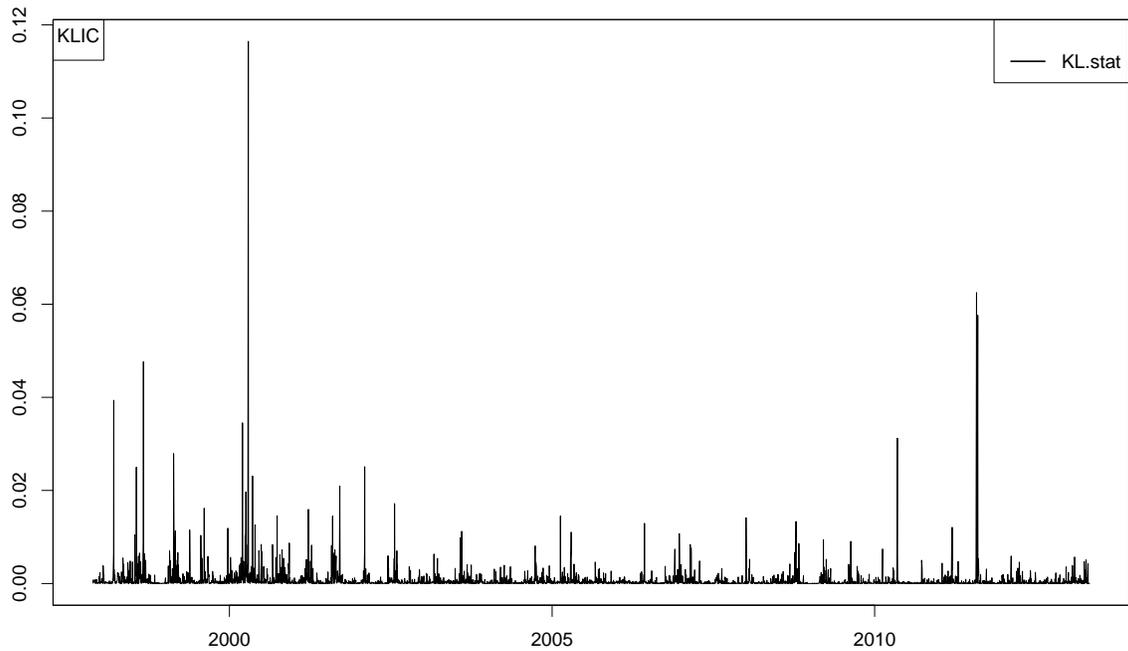


Figure 8: KL distance, daily stationary distributions, the case of dji, fchi, gdaxi, n225

a summary index to represent stock markets of developed markets. Building on an established methodology to assess return spillovers, we proceed to investigate the spread of information (hypothetical shocks to a market on a given day) through the network of markets.

We find that BRICT markets have become more connected over the past 15 years, with a significant rise beginning in 2005. Throughout the period considered, Brazil has the highest and China the lowest connectedness. We find evidence for an “informational divide” in the sense that the origin of a shock (a sudden drop or rise in a stock index) matters — in contrast to western developed equity markets. The level of information produced from day to day in the system of BRICT markets is approaching that of developed markets.

It may well be that convergence, to the effect of a reduced “informational divide” among BRICT markets, will be observed in the future. The present paper provides the methodology to track the future interplay of markets.

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