A Comparison of Stock Market Efficiency of the BRIC Countries

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Abstract

This article compares the stock market efficiency of Brazil, Russia, India and China (commonly referred to as BRIC). The profitability of trading rules associated with the Simple Moving Average (SMA), the Relative Strength Index (RSI), the Moving Average Convergence Divergence (MACD) and the Momentum (MOM) are evaluated. It is found that these indicators are most profitable in the Russian stock market. The Brazilian stock market is found to be the most efficient market among the BRIC. An explanation for such a discrepancy is provided.

Keywords: Simple Moving Average, Relative Strength Index, Moving Average Convergence Divergence, Momentum

1. Introduction

Over the past two decades, a good number of empirical studies have been conducted to evaluate the performance of different trading rules. Earlier studies focus on the appealing Variable-length Moving Average (VMA) rule, which states that a long position should be taken if the short-term VMA is above the long-term VMA, and vice versa. The VMA rule has been proven profitable in many studies. For example, Brock et al. [1] show that the VMA trading rule generates excess returns in the US market. Hudson et al. [2] and Mills [3] also find that the rule is profitable in the FT30 index. These early studies, however, focus on developed markets. Over the past decade, there is an increasing number of studies on the performance of technical trading rules in emerging markets. One strand of literature examines the currency market. For example, Martin [7] applies the moving-average rule to 12 emerging currencies and shows that the risk-adjusted return is not significant. Lee et al. [8] demonstrate that the moving average rule and the channel rule are profitable for the Brazilian Real, the Mexican Peso, and the Venezuelan Bolivar. Ahmed et al. [9] find that the moving average rule can beat the buy-and-hold strategy in the daily spot exchange rates of Chile, Mexico, Indonesia, the Philippines, South Korea, and Thailand. Craig et al. [10] show that including emerging market currencies in an investment portfolio substantially increases the Sharpe ratio associated with carry trades. Chong and Ip [11] demonstrate that the momentum rule is profitable in emerging currency markets.

Another strand of literature focuses on emerging stock markets. For example, Ito [12] finds profitable technical rules in the stock markets of Indonesia, Mexico and Taiwan. Parisi and Vasquez [13] show that buy signals generate higher returns than sell signals in the Chilean stock market. Hameed and Ting [14] find evidence of predictability for the Malaysian stock market. Gunasekara and Power [15] conclude that technical trading rules have predictive power in the markets of Bombay, Colombo, Dhaka and Karachi. Kang et al. [16] find significant returns of momentum strategies in the Chinese A-share market.

Recently, there has been an increasing interest in the study of BRIC, which refers to Brazil, Russia, India and China, the four largest and fast growing emerging countries. Given their increasingly important roles in the...
world financial markets, this paper attempts to compare the efficiency of the BRIC stock markets. The profitability of four popular technical indicators, namely, the Simple Moving Average (SMA), the Relative Strength Index (RSI), the Moving Average Convergence Divergence (MACD) and the Momentum (MOM), are reported. It is found that the trading rules associated with these indicators outperform the buy-and-hold strategy in India and Russia, while the stock market of Brazil is shown to be relatively efficient. Our results also show that, in general, the 10-day SMA and MACD rules are most profitable, followed by the 14-day RSI, 10-day SMA and 40-day MOM, while the 250-day SMA is the least profitable. The remainder of the paper is structured as follows: Section 2 describes the data and methodology used. Section 3 reports the results and Section 4 concludes the paper.

2. Data and Methodology

Our data are sourced from DataStream International available at the University Library of the Chinese University of Hong Kong. The data set consists of the daily closing indices of the stock markets of Brazil, Russia, India and China from September 1995 to November 2008. The details are shown in Table 1.

Let $\text{SMA}_N(t)$ be the $N$-day simple moving average and $P(t)$ be the closing value of the stock market index at day $t$. When the index crosses its moving average from below, the market is considered bullish, while it is considered bearish otherwise. Therefore, we define the SMA trading rule at time $t$ as

Buy if $P(t-1) < \text{SMA}(t-1)$ and $P(t) > \text{SMA}(t)$;

Sell if $P(t-1) > \text{SMA}(t-1)$ and $P(t) < \text{SMA}(t)$.

The 10, 50 and 250-day SMA are examined.

The RSI index was developed by Wilder [20]. An $N$-day RSI is the ratio of the sum of all positive changes of prices to the sum of absolute changes of prices in $N$ days, multiplied by 100:

$$\text{RSI}_N(t) = \frac{\sum_{i=0}^{N-1} (P(t-i) - P(t-i-1)) |\{P(t-i) > P(t-i-1)\}|}{\sum_{i=0}^{N-1} |P(t-i) - P(t-i-1)|} \times 100$$

where $|P(t-i) - P(t-i-1)|$ is the absolute daily change of the index, and $1 \{P(t-i) > P(t-i-1)\}$ is an indicator function which equals one if $P(t-i) > P(t-i-1)$, and equals zero otherwise. The RSI ranges from 0 to 100. If the index is trending down, the RSI will approach zero. If the index is trending up, the RSI will approach 100. The middle of this range is often regarded as the cut-off point of bullish and bearish markets. The RSI trading rule at time $t$ is defined as follows:

Buy if $\text{RSI}(t-1) < 50$ and $\text{RSI}(t) > 50$;

Sell if $\text{RSI}(t-1) > 50$ and $\text{RSI}(t) < 50$.

The rule states that a buy signal is triggered when the RSI crosses 50 from below, while a sell signal is generated when it crosses 50 from above. In this paper, the commonly used 9-day and 14-day RSI will be examined.

The moving average convergence divergence (MACD) proposed by Appel [21] is the difference between the long-term exponential moving average (EMA) and the short-term EMA. The N-day exponential moving average at time $t$ is defined as

$$\text{EMA}_N(t) = aP(t) + (1-a)\text{EMA}_{N}(t-1),$$

where $P(t)$ is the index value at time $t$. The initial EMA is defined as the closing value of the index at time 1. For any integer $M > N > 0$, the MACD is defined as

$$\text{MACD}(t) = \text{EMA}_M(t) - \text{EMA}_N(t).$$

In addition, we define the signal line as the EMA of the MACD. The MACD rule is then defined as follows:

Buy if $\text{MACD}(t-1) < \text{signal-line}(t-1)$ and $\text{MACD}(t) > \text{signal-line}(t)$;

Sell if $\text{MACD}(t-1) > \text{signal-line}(t-1)$ and $\text{MACD}(t) < \text{signal-line}(t)$.

Thus, a buy signal is generated when the MACD crosses its signal line from below, while a sell signal is generated otherwise. Following Murphy [22], we use the 12 and 26-day EMAs as our short and long-run EMAs, and the 9 and 14-day EMA of the MACD as our signal lines.

The momentum indicator (Chande and Kroll [23]) is defined as the difference between the closing prices of two trading days. Specifically, the N-day momentum is constructed by subtracting the closing price at time $t - N$ from the closing price at time $t$.

Table 1. The stock market indices of BRIC and their sample periods.

| Countries   | Index                               | Founded/Reopening year | From       | To         |
|-------------|-------------------------------------|------------------------|------------|------------|
| Brazil      | Brazil Bovespa                       | 1895                   | 1/9/1995   | 1/12/2008  |
| India       | India BSE (Sensex) 30 Sensitive      | 1875                   | 1/9/1995   | 1/12/2008  |
| Russia      | Russia RTS Index                     | 1995                   | 1/9/1995   | 1/12/2008  |
| China       | Shenzhen SE Composite SUB            | 1990                   | 1/9/1995   | 24/11/2008 |
| China       | Shanghai SE A Share                  | 1990                   | 1/9/1995   | 24/11/2008 |
respectively in the
S(j) and B(j) are the selling and buying price of the index
where

\[ C \quad o \quad r \quad i \quad g \quad h \quad i \quad t \quad c \quad o \quad p \quad i \quad r \quad t \quad g \quad e \quad n \]

Table 2
3. Results
rentheses are the numbers of transactions for the corre-
by the aforementioned trading rules. The figures in pa-

| Performance of these trading rules using the annualized rate
| Annual Rate of Return
| \[ = \left(1 + r_1\right)\left(1 + r_2\right)\ldots \left(1 + r_m\right)^{250/T} - 1 \],
| where
| \[ 1 + r_j = S(j)/B(j); \]
| S(j) and B(j) are the selling and buying price of the index
| respectively in the \( j \text{th} \) transaction, \( m \) is the number of
| transactions and \( T \) is the number of trading days in our
| sample.

3. Results

Table 2 reports the annualized rate of return generated by the aforementioned trading rules. The figures in par-

| Index Name | SMA\(_{10}\) | SMA\(_{50}\) | SMA\(_{250}\) | MOM\(_{10}\) | MOM\(_{50}\) | MOM\(_{40}\) | RSI\(_{9}\) | RSI\(_{14}\) | MACD\(_{12,26,9}\) | MACD\(_{12,26,14}\) | B & H |
|------------|------------|------------|------------|------------|------------|------------|---------|---------|---------------|---------------|-------|
| BRAZIL BOVESPA | 1.33% | 15.62% | 9.33% | -4.18% | \textbf{16.27%} | -4.18% | 10.99% | -1.30% | 3.52% | 16.02% | (1) |
| (550) | (228) | (79) | (466) | (200) | (464) | (363) | (309) | (247) | (1) |
| INDIA BSE (SENSEX) 30 SENSITIVE | 11.92% | \textbf{16.92%} | 8.52% | 8.49% | 1.69% | \textbf{8.49%} | \textbf{8.22%} | 15.71% | 11.39% | 7.25% | (1) |
| (469) | (150) | (88) | (405) | (175) | (403) | (309) | (255) | (233) | (1) |
| RUSSIA RTS INDEX | \textbf{60.58%} | \textbf{40.35%} | 8.52% | \textbf{49.63%} | \textbf{26.75%} | \textbf{49.63%} | \textbf{48.88%} | \textbf{31.73%} | \textbf{30.21%} | 14.26% | (1) |
| (502) | (164) | (63) | (383) | (164) | (381) | (296) | (259) | (227) | (1) |
| SHENZHEN SE COMPOSITIVE SUB | 8.45% | \textbf{29.37%} | 5.58% | 3.19% | 8.41% | 4.46% | \textbf{24.04%} | \textbf{20.86%} | \textbf{17.12%} | 13.52% | (1) |
| (502) | (153) | (47) | (462) | (135) | (452) | (345) | (271) | (237) | (1) |
| SHANGHAI SE A SHARE | -5.71% | \textbf{19.59%} | -5.25% | 2.36% | -4.32% | 3.92% | \textbf{8.69%} | \textbf{19.87%} | \textbf{23.05%} | 7.28% | (1) |
| (503) | (138) | (77) | (458) | (175) | (446) | (308) | (277) | (221) | (1) |

generate the same return, we prefer the rule with the least
number of transactions. The highest return for each mar-
ket ranges from 16.23% (Brazil) to 60.58% (Russia). Amongst these nine trading rules, the SMA\(_{10}\), SMA\(_{50}\) and
the two MACD rules are most profitable. The rules of
RSI\(_{14}\) and MACD yield excess returns in at least four
markets, whereas the rule of SMA\(_{250}\) is the least profit-
able. Among these trading rules, the SMA\(_{250}\) Rule for the
case of Russia and the MOM\(_{40}\) Rule for the case of India
cannot generate excess returns.

Our trading rules beat the buy-and-hold strategy in the
India BSE (Sensex) 30 Sensitive and the Russia RTS
Index. The trading-rule profits are attributable to the se-
rial correlation in stock returns. In general, this serial
relation is higher for emerging markets than for de-
veloped markets (Harvey [24]). Note that the Brazilian
market is quite efficient. Most trading rules cannot beat
the buy-and-hold benchmark, except for the MOM\(_{40}\)
rule. The results for China are mixed. The SMA\(_{50}\), RSI\(_{14}\)
and MACD rules beat the buy-and-hold strategy, while
other rules do not.

4. Conclusion

This paper examines the stock market efficiency of Brazil,
Russia, India and China. The profitability of trading rules
associated with the SMA, RSI, MACD and MOM are
evaluated. Our results show discrepancies in trading-rule
performance among the BRIC markets. A reason for this
lies in the history of the stock exchanges of these coun-
tries. Note from Table 1 that the founded/reopening
years of the stock exchanges of these markets vary. Our
trading rules perform quite well in markets with a rela-
tively short history (Russia), but do not work in markets with a long history (Brazil). This provides indirect evidence for the general observation that stock markets are getting more efficient over time (Li et al. [25]). A limitation of our study is that we do not consider transaction costs here. Transaction costs are generally higher in emerging markets as compared to the developed markets. The transaction cost mainly consists of the bid-ask spread and the stamp duty. For developed markets, the cost is about 0.1% to 0.5% for a round-trip transaction. For emerging markets, even the cost is two or three times higher, the overall annual cost will still be lower than 5% if there are ten transactions per years. Our results are robust to the presence of transaction costs. For example, for the RUSSIA RTS INDEX, the annualized returns of most trading rules are over 30%. As the number of transactions in a year is general less than 20, the annualized transaction cost in this market is about 10%. Even this 10% cost is taken into account; it is still profitable to apply our trading rules to the Russian market. A further development of our study would be to allow for transaction costs, or to include more emerging markets from different continents, such as South Africa, United Arab Emirates, Hungary and Malaysia in our analysis.

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Rise of BRICS Economy and its Impact on Global Stock Markets

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Abstract

The world is changing and becoming increasingly multipolar due to the emergence of China, India, Russia, Brazil and South Africa forming so called BRICS. And features, pointing out the most important issues related to demand, supply and industry structure.

- Environmental analysis at a Macro and Micro level, analyse the global demand trends and factors affecting it as well as a market segmentation and geographical distribution of the demand.
- Supply chain analysis, role of BRIC countries as industry suppliers, main competitors and influence of technological evolution on the.

While the stock return series for the BRICS countries have different start dates but same stop dates owing to data availability, the oil shocks data are monthly series starting from February 1975 to July 2018. The start dates for the daily stock returns are presented in the ninth column of Table 1a; with South Africa having the earliest start date.

For robustness purposes, our analysis of stock volatility reaction to oil shocks for each of the BRICS countries is carried out under three different measures. The first uses the aggregated oil shocks for the four different oil shocks (OSS, EAS, OCDS and OIDS).

The second and third measures respectively involves disaggregating the each of the four oil price proxies into negative and positive shocks, using dummy variables. In this paper, we investigated the profitability of technical analysis as applied to the stock markets of the BRICS member nations. In addition, we searched for evidence that technical analysis and fundamental analysis can complement each other in these markets. To implement this research, we created a comprehensive portfolio containing the assets traded in the markets of each BRICS member.

Scholars have tested the efficiency of the tools of technical analysis frequently, for example, in the studies of Allen and Taylor (1990), Jegadeesh (2000), and Kuang et al. (2008). However, none of these studies proposed a comparison of the results for groups of similar countries, so they failed to answer whether TA is profitable for emerging markets as a whole. The stock market value of BRICS is at a three-year low. Top global companies from the BRICS countries suffered erosion in market values, leading to sizeable slippage in their respective global rankings. In this context, the BRICS community stands at an important point of challenge.

While the last two decades have galloped these countries to a position of prominence in global economics, a new set of constraints began to arise in the.

5. OVERVIEW

Ranks of BRICS countriesâ€™ stock exchanges on the basis of stock market capitalization in 2012. Rank Exchange. Country.