# Financial Econometrics 871 Practical Exam 2022: Question 5

#### Tian Cater<sup>a</sup>

<sup>a</sup> University of Stellenbosch, Western Cape, South Africa

#### 1. Introduction

The purpose of this question is to comment on the following two statements:

- 1. The South African rand (ZAR) has, over the past few years, been one of the most volatile currencies
- 2. The ZAR has generally performed well during periods where G10 currency carry trades have been favourable, and these currency valuations are relatively cheap. Globally, it has been one of the currencies that most benefit during periods when the Dollar is comparatively strong, indicating a risk-on sentiment.

In summary, the findings here are in contrast to the first statement. That is, even though the ZAR has been highly volatile in the past few years, it is not one of the most volatile, having similar variance in the recent years as in the more distant past. With respect to the second statement, the evidence here strongly agrees. That is, that the ZAR is partially insulated from periods of weak performance in the G10 countries. Specifically, during periods where the USD has been weak, the ZARs positive dynamic conditional correlation with the USD has dropped significantly.

#### 2. Comparing Global Currency Volatility to the ZAR

To investigate the first statement, I select 8 countries' currencies: Brazil, EU, India, SA, Turkey, Poland, Zambia, and the UK and compare their respective dollar exchange rates. The reasoning behind the selection is diversity in comparison; by including developing nations with historically low levels of spillovers from the US (Brazil and India), one of the largest economies in Africa (Zambia), two developing nations from Asia (Poland and Turkey), and two of the strongest currencies in the

Email address: 19025831@sun.ac.za (Tian Cater)

world (UK and EU), a clear picture can be painted on how volatile the ZAR is in comparison. Figure 2.1 graphs the scaled logarithmic growth of the respective currencies to the USD since 2005, giving weak hints that the ZAR may not have been more volatile in the last couple of years than the first comment suggests.

### Scaled (demeaned) Log Growth of Respective Currencies to USD since 2

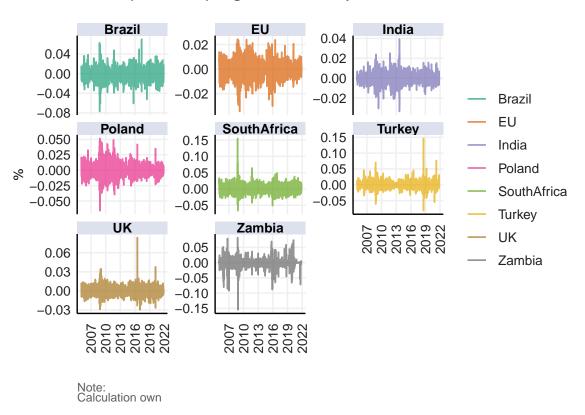


Figure 2.1: Scaled (demeaned) Log Growth of Respective Currencies to USD since 2005.

To get an initial comparison of the volatility of the selected currencies, Figure 2.2 plots the respective sample standard deviation in scaled log growth against the dollar.<sup>2</sup> Upon this initial inspection of Figure 2.2, it is apparent that the ZAR has recently not been the most volatile against the USD.

<sup>&</sup>lt;sup>2</sup>The dates are filtered to consider the period from 2006 onwards to remove extremely volatile periods for international currencies.

#### **Brazil** EU India 0.08 0.04 0.03 0.06 0.03 0.02 0.04 0.02 0.01 0.02 0.01 Brazil 0.00 0.00 0.00 EU **SouthAfrica Poland** Turkey 0.15 0.15 0.06 India 0.10 0.10 0.04 Poland % 0.05 0.05 0.02 SouthAfrica 0.00 0.00 0.00 Turkey UK Zambia UK 0.15 80.0 0.06 Zambia 0.10 0.04 0.05 0.02 0.00 0.00 2007 2010 2013 2016 2016 2019 Note: Calculation own

#### Sample Standard Deviation of Respective Currencies to USD since 2005.

Figure 2.2: Sample Standard Deviation of Respective Currencies to USD since 2005.

#### 3. DCC Multivariate GARCH Model

To better understand the volatility of these currencies, I take a deeper dive into the volatility of these currencies and fit a multivariate GARCH model, but by also including the Bloomberg Dollar Spot Index (BBDXY) as a variable; it tracks the performance of a basket of 10 leading global currencies versus the U.S. Dollar. It has a dynamically updated composition and represents a diverse set of important currencies from trade and liquidity perspectives.

I start by conducting the MARCH test, which indicates that all the multivariate portmanteau tests reject the null of no conditional heteroskedasticity, motivating my use of a MVGARCH model.<sup>3</sup>

I decide to use a DCC MVGARCH Model since DCC models offer a simple and more parsimonious means of doing MV-vol modelling. In particular, it relaxes the constraint of a fixed correlation structure (assumed by the CCC model), to allow for estimates of time-varying correlation.

 $<sup>^3{\</sup>rm The}$  relevant test statistics can be seen in my README.

The DCC GARCH estimated volatility (sigma) for each currency is depicted in Figure 3.1. These volatility estimates are slightly different than the simple sample SD graphed in Figure 2.2 above, in that the ZARs volatility has increased relative to the other currencies, however, Brazil and Turkey still showcases more volatility, even in the recent few years.

## 0.06 **BBDXY** Brazil 0.04 India Poland SouthAfrica 0.02 Turkey 0.00 2008 2010 2012 2014 2016 2018 2020 Note: Calculation own

### DCC GARCH: Estimated Volatility (Sigma) for Each Currency

Figure 3.1: DCC GARCH: Estimated Volatility (Sigma) for Each Currency

After fitting the DCC GARCH, I plot the dynamic conditional correlation with respect to the ZAR in Figure 3.2 below, where, even though the graph is not very clear, it is apparent that the ZAR is the least correlated with the Indian Rupee. To get a clearer picture of the impact of the USD on the ZAR, I remove some clutter and only plot the dynamic conditional correlations between the ZAR and the USD in Figure 3.3.

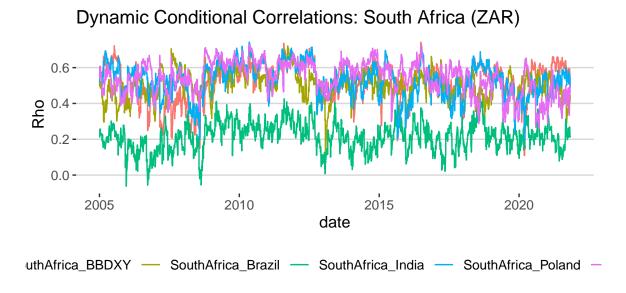


Figure 3.2: Dynamic Conditional Correlations: South Africa (ZAR)

In analysing Figure 3.3, it becomes clear that The ZAR has generally performed well during periods where G10 currency carry trades have been favourable, and these currency valuations are relatively cheap. That is, in periods where the USD has performed poorly against the other G10 currencies (reflected by BBDXY), for example following the GFC from 2008 to 2012 and the COVID-19 pandemic from 2019 onwards, the ZAR has had the lowest conditional correlation with the BBDXY. Therefore, indicating that the ZAR is partially insulated against the backdrop of advanced economies' currency downturns, which indicates a risk-on sentiment

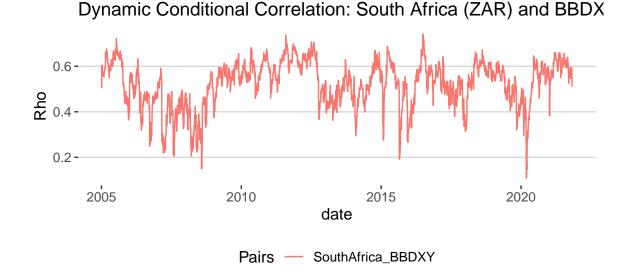


Figure 3.3: Dynamic Conditional Correlation: South Africa (ZAR) and BBDXY

## References

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