

# ZAR Risk and Return Spillovers in a Global Model of the Foreign Exchange Network

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*The views expressed are those of the authors and do not necessarily represent those of the South African Reserve Bank or Reserve Bank policy.*

# Motivation

- We develop a network model to capture the dynamic interactions among foreign exchange (FX) returns and realised risk measures among a large panel of 20 developed and emerging market currencies, including the rand (ZAR).
- We demonstrate how this framework can be used to assess the sensitivity of a given currency to shocks from other currencies and to provide narratives contextualising currency movements, focusing on the ZAR.

# Summary of findings

- We show that variations in the risk-return profile of the USDZAR correlate with variations in the risk-return profile of many other currencies, and that this is especially notable with respect to emerging market currencies.
- We interpret this as evidence of the ZAR's role as a bellwether emerging market currency.
- We show that the model is able to highlight risk transmission channels in a timely manner during foreign exchange flash crashes and periods of heightened financial market uncertainty.

# Introduction

- Much policy debate about relative volatility of ZAR and factors influencing ZAR volatility, but relatively little research into nature and drivers of ZAR volatility.
- Likewise, trend depreciation of ZAR has drawn attention to directional risks, yet there is an absence of research into skewness of ZAR (i.e. implied risk that ZAR could move in a particular direction).
- We construct a database of log-returns and higher-order realised moments for ZAR and large number of emerging market (EM) and developed market (DM) currencies.
- We develop and estimate an empirical network model to measure intensity of bilateral FX risk and return spillovers among currencies.

# Dataset

- 12 DM currencies and 8 EM currencies
- Construct returns and a variety of higher-moment risk measures for each currency from Refinitiv intra-day data (and compare our estimates to Bloomberg estimates in our data appendix)
- Following conventions described in Andersen et al. (2003), we consider a trading day to start at 21:05GMT and to end at 21:00GMT.
- Convention is to construct mid-rates at five-minute intervals ( $n = 288$ ), which balances asymptotic considerations against microstructure noise, to compute first four realised moments of the return distribution at daily frequency

# Calculation of daily log-returns

$$r_{it} = \sum_{j=1}^n r_{it,j},$$

where:

- $r_{it,j} = \log(p_{it,j}/p_{it,j-1})$  is  $j$ th period-to-period intraday log-return or for currency  $i$  on day  $t$
- $p_{it,j}$  denote  $j$ th intraday mid-rate

# Calculation of realized variance

- Realised variance:

$$UV_{it} = \sum_{j=1}^n r_{it,j}^2.$$

- Scaled by trading days per year  $N$  to obtain annualised measure:

$$V_{it} = N \cdot UV_{it}.$$

- Realised variance captures dispersion of returns, which is a measure of uncertainty.



# Calculation of higher moments

- Realised skewness measures degree of asymmetry of returns

$$Q_{it} = \sqrt{n} \sum_{j=1}^n r_{it,j}^3 / (UV_{it})^{3/2}.$$

- Realised kurtosis measures mass in tails of return distribution: larger  $K_{it}$  indicate higher probability of extreme FX returns

$$K_{it} = n \sum_{j=1}^n r_{it,j}^4 / (UV_{it})^2.$$

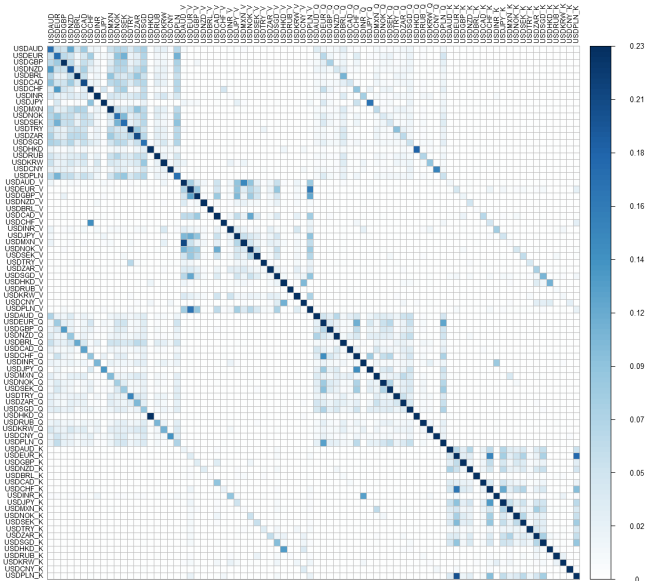
# Network model

- We adopt a modified version of Diebold and Yilmaz (2009, 2014), in which the structure of the network of bilateral spillover effects is estimated via a decomposition of variance of forecast errors obtained from a VAR model.
- By analysing structure of the network on a rolling sample basis, we track evolution of spillovers over time.
- The spillover statistics that we obtain can be used to shed light on how changes in a currency's risk-return profile transmit through global foreign exchange markets and to assess the degree to which such changes reflect global and local factors.

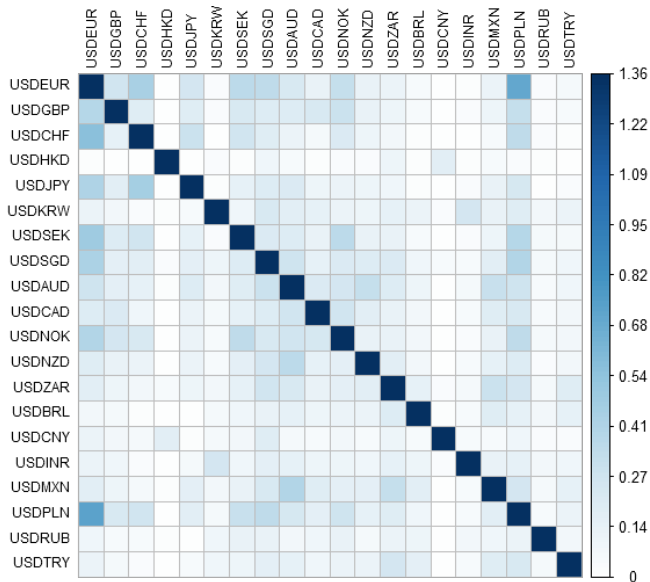
# Network model

- Diebold and Yilmaz (2009) show that a decomposition of forecast error variance from a VAR model can be interpreted as a weighted directed network
- Approach generalised by Diebold and Yilmaz (2014) to achieve invariance to the order of the variables in the VAR model and by Greenwood-Nimmo et al. (2021) to facilitate block aggregation of the estimated network
- We apply the aggregation-robust framework of Greenwood-Nimmo et al. (2021) because it allows us to switch freely between different aggregation routines in order to study different aspects of FX connectedness

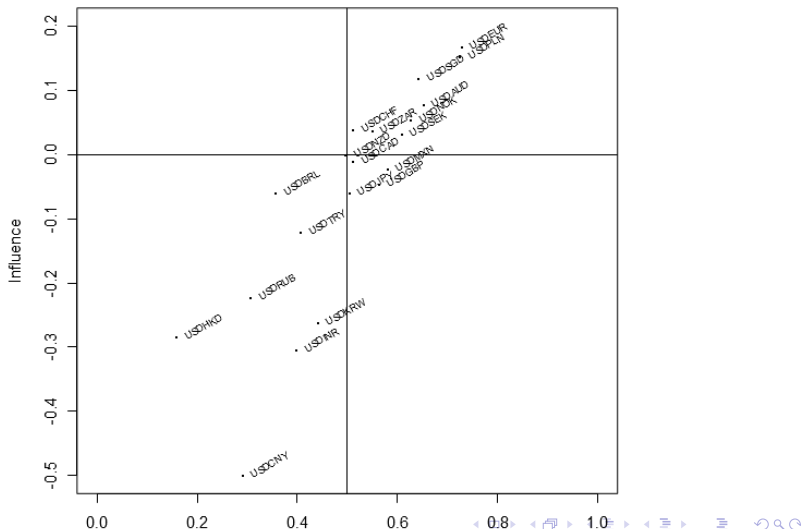
## Disaggregate Connectedness over the Full Sample



# Connectedness among Currencies over the Full Sample



## Influence and Dependence of Currencies over Full Sample



# Rolling-Sample Network Statistics



(a) Disaggregate spillover index

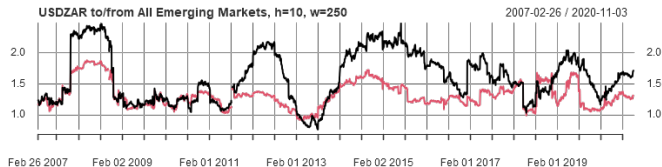


(b) Currency spillover index

# Connectedness of the ZAR with DM/EM Currencies



(a) ZAR vs. All DM Currencies



(b) ZAR vs. All EM Currencies



# Using the FX Network for Narrative Analysis

- Our model can be used to contextualise abrupt movements in exchange rates
- We consider three classes of event: (i) idiosyncratic domestic events; (ii) idiosyncratic foreign events; and (iii) global events.

# Idiosyncratic Domestic Events

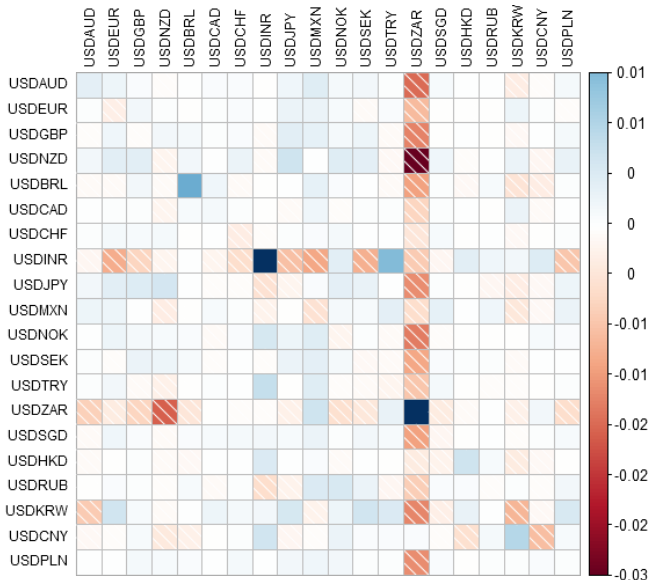
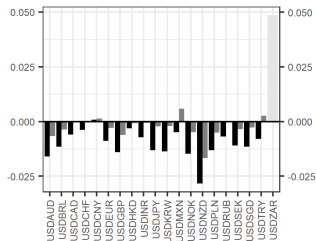
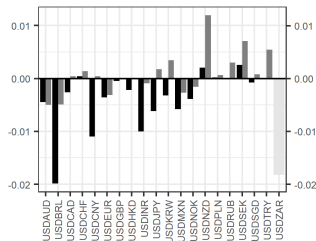


Figure: Change in Spillovers when Nene was Dismissed

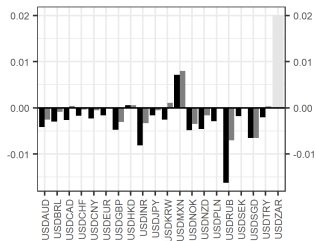
# Domestic Event Days



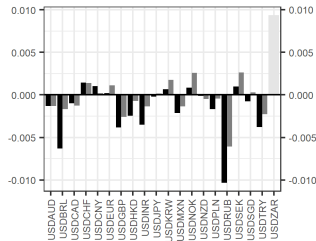
(a) Dismissal of Nene



(b) Appointment of Gordhan

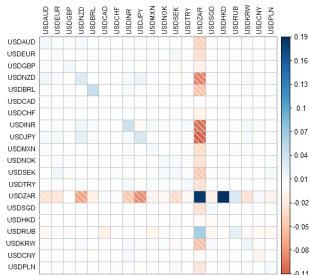


(c) Questioning of Gordhan

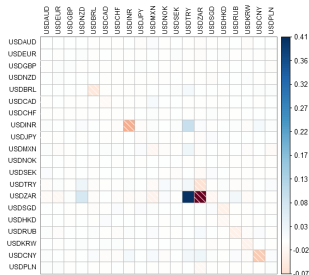


(d) Dismissal of Gordhan

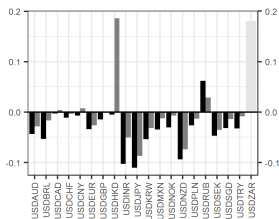
# Foreign Idiosyncratic Events



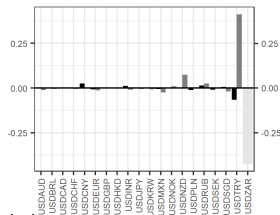
(a) Chinese Stock Market Slump



(b) Turkish Lira Slump

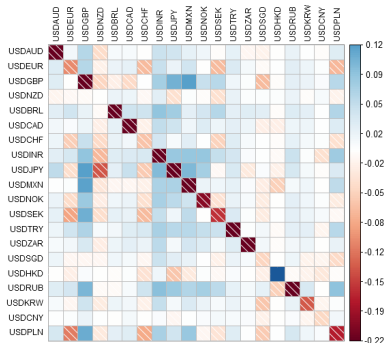


(c) Chinese Stock Market Slump

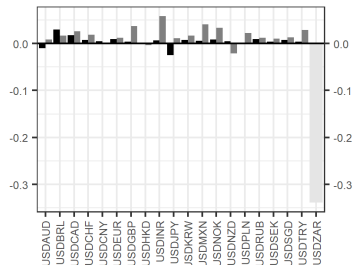


(d) Turkish Lira Slump

# Global Events



(a) Systemwide Connectedness



(b) ZAR Connectedness

Figure: Change in Spillovers to/from ZAR over Brexit Referendum

# Summary of findings

- Our model highlights risk transmission channels in a timely manner during FX flash crashes and periods of heightened financial market uncertainty.
- We show that variations in risk-return profile of USDZAR correlate with variations in the risk-return profile of many other currencies, and that this is especially true of EM currencies in our sample.
- We interpret this as evidence of ZAR's role as a bellwether EM currency, at least in times when South African idiosyncratic risk is not elevated.

# Extra slides

# Data

- We isolate innovations in higher-order moments as residual from a first order autoregression to account for greater persistence in realised moments than FX log-returns:

$$\hat{v}_{it} = V_{it} - \hat{a}_i + \hat{b}_i V_{it-1}, \quad (1)$$

- An extensive data appendix is available on request