

The impact of exchange rate volatility on BRICS Trade

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Acknowledgement

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Introduction

- The breakdown of the Bretton-Woods agreement signalled the dawn of a new era for the global economy in which many of the worlds major trading nations would embrace a regime of floating exchange rate determination (McKenzie & Melbourne, 1999)
- Thus the currencies are traded on the basis of the forces of demand and supply
- Consequences are that risk averse traders would cut back on out put in fear of exchange rate volatility
- Exchange rate volatility poses a potential to reduce trade flows
- **Consequences of exchange rate volatility on trade** have long been at the centre of the debate on the optimality of alternative **exchange rate**(see Côté, A., 1994. *Exchange rate volatility and trade: A survey* (No. 1994-5). Bank of Canada.)

- The **exchange rate volatility** has a significant negative **impact due to exchange rate** uncertainty (see *Chowdhury AR. Does exchange rate volatility depress trade flows? Evidence from error-correction models. The review of economics and statistics. 1993 Nov 1:700-6.*)
- Khan AJ, Azim P, Syed SH(2014) use GARCH based **exchange rate volatilities** measure the **volatility impact** on trade (see *The impact of exchange rate volatility on trade: a panel study on Pakistan's trading partners. The lahore journal of economics. 2014 Jul 1;19(1):31.*)

Aim of the Study & Methodology

- The aim of this study is to assess the impact of exchange rate volatility on BRICS trade flows.
- Following the work of Khan AJ, Azim P, Syed SH(2014) we make use of GARCH Model to test for volatility in the BRICS Currencies
- We make use of a non-linear quintile regression to detect the impact on trade flows
- We incorporate ARDL Model to substantiate Quintile Regression Results and also to capture the long run and short run effects

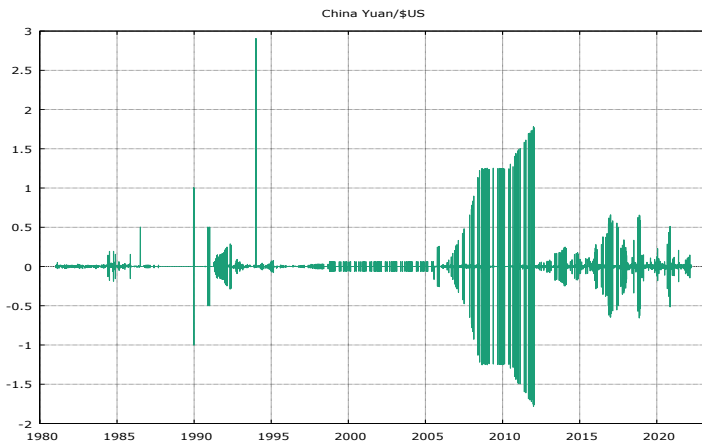
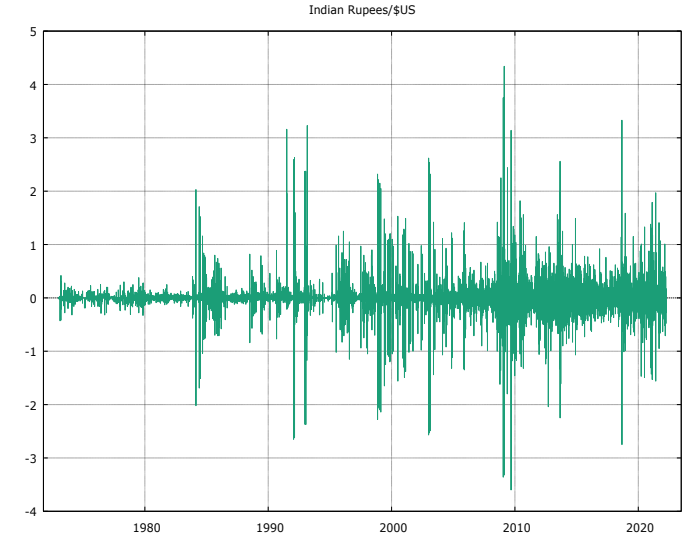
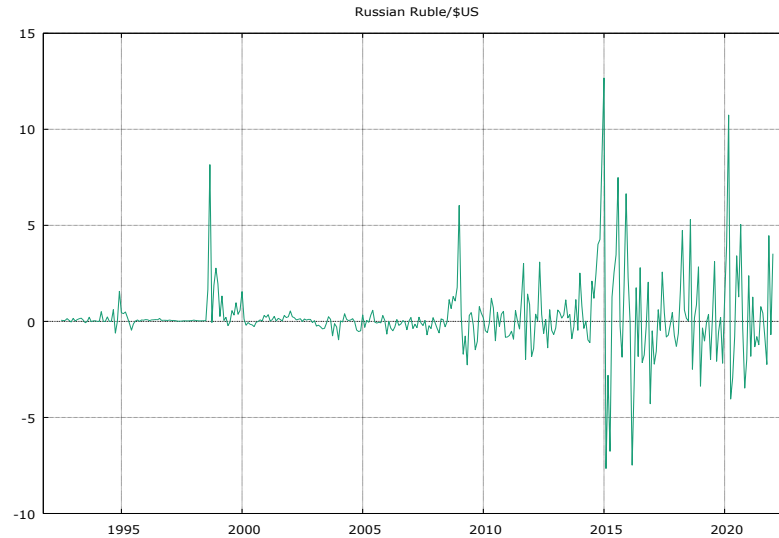
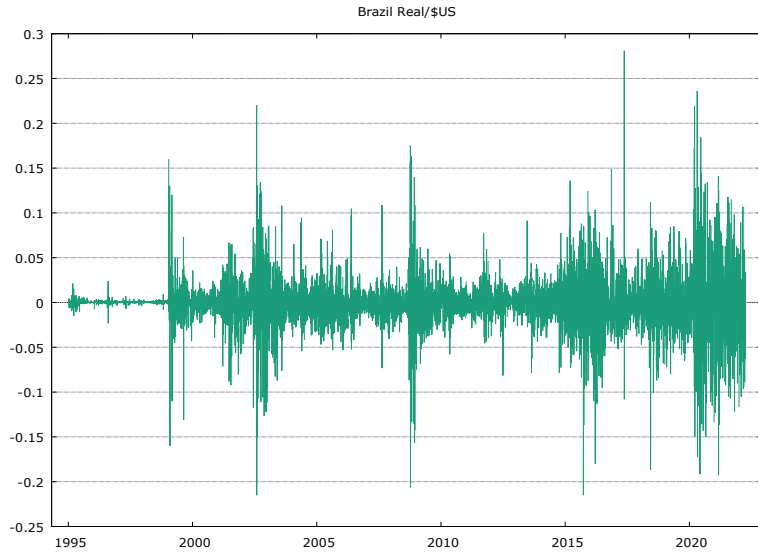
DATA SOURCES

- **Federal Reserve Economic Data daily exchange rates**
- **UNCTAD**
- **WORLD BANK**
- **IMF**

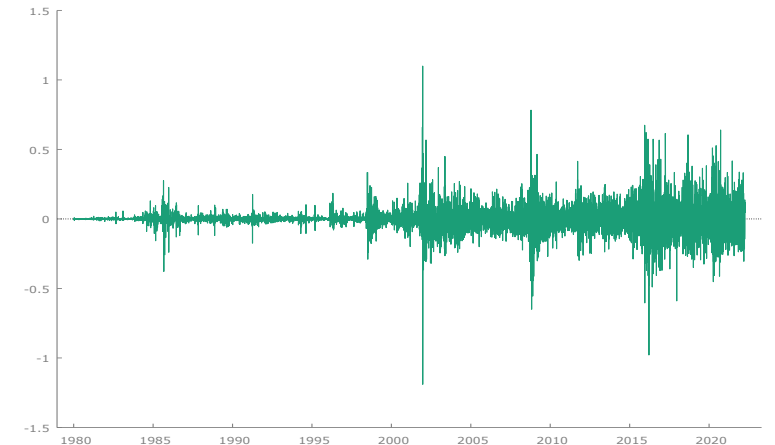
ECONOMETRIC MODELS USED IN ANALYSIS

- GARCH is a statistical model that can be used to **analyze a number of different types of financial data**, for instance, macroeconomic data. Financial institutions typically use this model to estimate the volatility of returns for stocks, bonds, and market indices.
- We make use of GARCH (1,1) Model
- Quantile regression is **an extension of Standard linear regression, which estimates the conditional median of the outcome variable and can be used when assumptions of linear regression do not meet.**

Volatile BRICS currencies data series



Source: FRED Economic Data, 2022



Model 1: BRAZIL : GARCH, using observations 1995-01-03:2022-04-01 (T = 7109)

Dependent variable: Brazillian Real/\$US Daily Rates
Standard errors based on Hessian

	Coefficient	Std. Error	z	p-value	
const	-0.0053624	0.00540902	-0.9914	0.3215	
	6				
Real/\$US_1	1.00718	0.00212579	473.8	<0.0001	***
C	0.0308929	0.00120931	25.55	<0.0001	***
Resid(-1)^2	0.328960	0.0269205	12.22	<0.0001	***
GARCH(-1)	0.296056	0.0221624	13.36	<0.0001	***

Null hypothesis: no ARCH effect is present

Test statistic: LM = 6192.95 with p-value = P(Chi-square(5) > 6192.95) = 0.0001

Model 2:RUSSIA:GARCH(1,1) [Bollerslev] (Normal)*

Dependent variable: Ruble/\$US daily rates

Reduced Sample: 1992:07 -- 2022:01 (T = 355), VCV method: Hessian

Conditional mean equation

	coefficient	std. error	z	p-value	
const	0.131041	0.0310372	4.222	2.42e-05	***
exch_rate_1	0.996466	0.00145706	683.9	0.0000	***

Conditional variance equation

	coefficient	std. error	z	p-value	
C	0.0249229	0.00616399	4.043	5.27e-05	***
Resid(-1)^2	1.00378	0.141258	7.106	1.19e-012	***
GARCH(-1)	0.444317	0.0366779	12.11	8.90e-034	***

Null hypothesis: no ARCH effect is present
Test statistic: LM = 330.827
with p-value = $P(\text{Chi-square}(12) > 330.827) = 1.54401e-63$

Model 3:INDIA: GARCH, using observations 2000-01-03:2022-04-01

(T = 5805) – Reduced Sample

Dependent variable: Rupee/\$US daily rates

Standard errors based on Hessian

	Coefficient	Std. Error	z	p-value	
const	-0.00361559	0.0154862	-0.2335	0.8154	
Rupee/\$US (-1)	1.00013	0.000285783	3500.	<0.0001	***
C	0.00330862	0.000299280	11.06	<0.0001	***
Resid(-1)^2	0.0880998	0.00695548	12.67	<0.0001	***
GARCH(-1)	0.884822	0.00765742	115.6	<0.0001	***

Null hypothesis: no ARCH effect is present
 Test statistic: LM = 5776.69 with p-value = 0.0001

Model 4: CHINA: GARCH(1,1) [Bollerslev] (Normal)*

Dependent variable: Yuan/\$US

Sample: 1981-01-05 -- 2022-04-01 (T = 10760), VCV method: Hessian

	coefficient	std. error	z	p-value
const	0.00305289	0.00119207	2.561	0.0104 **
exch_1	0.999695	0.000171319	5835	0.0000 ***

Conditional variance equation

	coefficient	std. error	z	p-value
C	7.28272e-05	2.00477e-06	36.33	6.06e-289 ***
Resid(-1)^2	0.0332466	0.00130827	25.41	1.83e-142 ***
GARCH(-1)	0.960860	0.000835552	1150	0.0000 ***

Null hypothesis: no ARCH effect is present

Test statistic: LM = 10730.8

with p-value = $P(\text{Chi-square}(5) > 10730.8) = 0.0001$

Model 5: SOUTH AFRICA: GARCH, using observations 2000-01-03:2022-04-01

(T = 5805): Reduced Sample

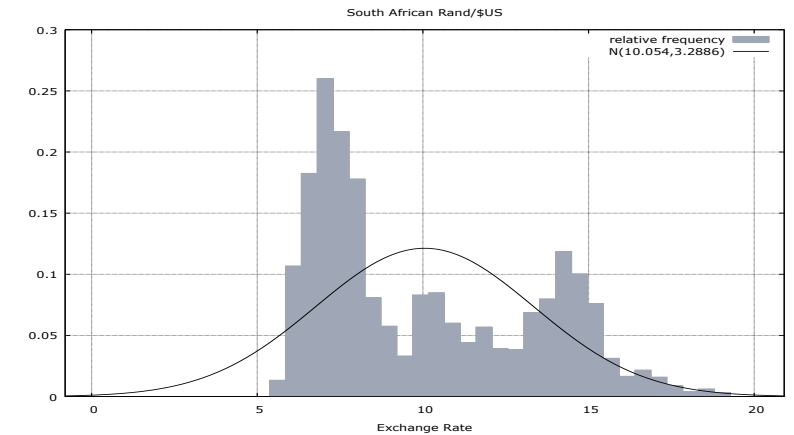
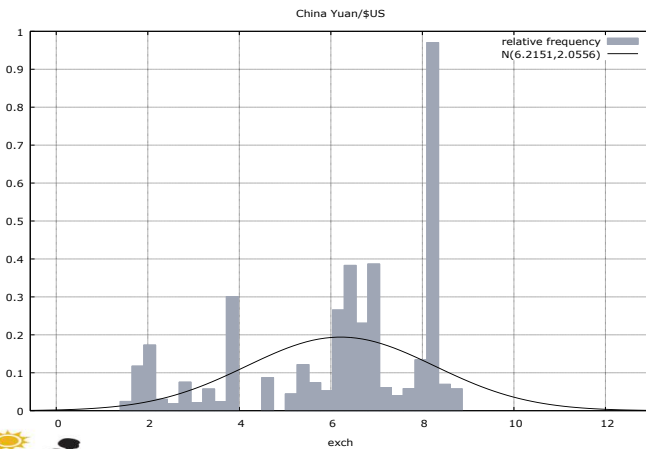
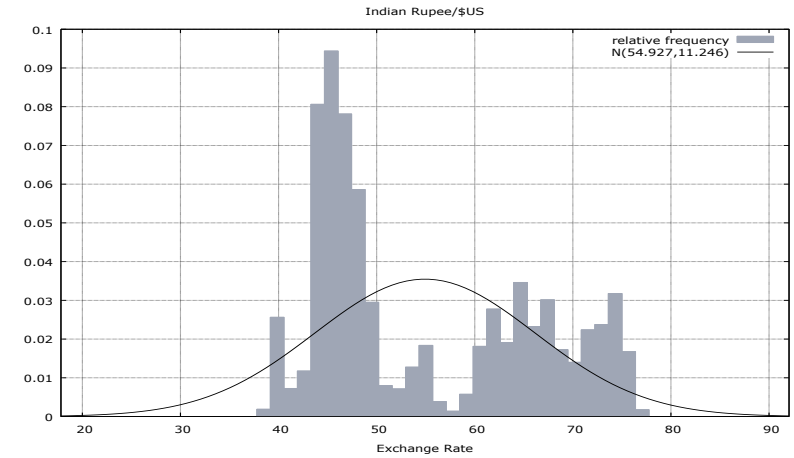
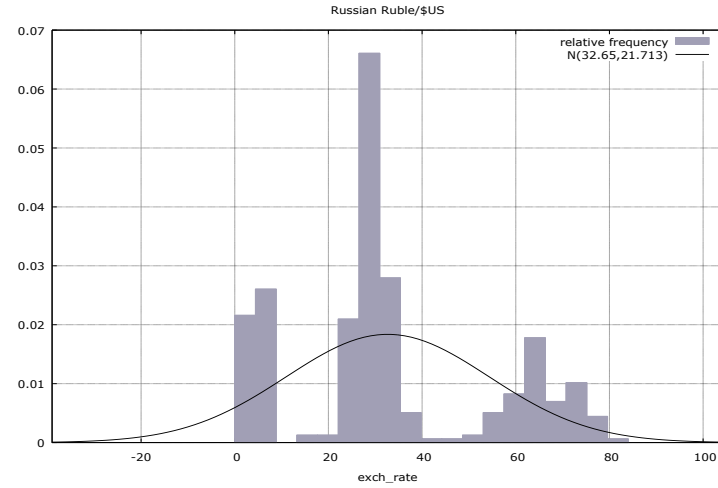
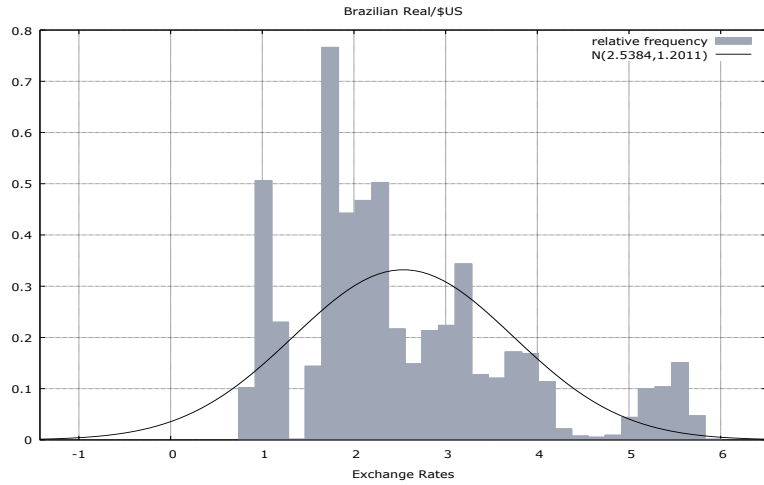
Dependent variable: Rand/\$US

Standard errors based on Hessian

	Coefficient	Std. Error	z	p-value
const	0.00318815	0.00366897	0.8690	0.3849
Rand/\$US	0.999786	0.000424908	2353.	<0.0001 ***
C	4.55399e-05	1.12830e-05	4.036	<0.0001 ***
Resid(-1)^2	0.0739302	0.00583322	12.67	<0.0001 ***
GARCH (-1)	0.925495	0.00555113	166.7	<0.0001 ***

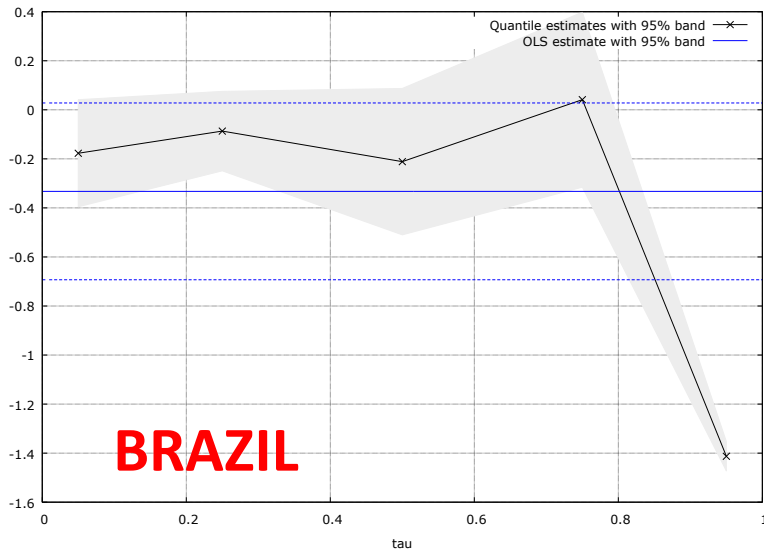
Null hypothesis: no ARCH effect is present
 Test statistic: LM = 5759.06 with p-value = P(Chi-square(5) > 5759.06) = 0.0001

LEPTOKURTIC FREQUENCY DISTRIBUTION

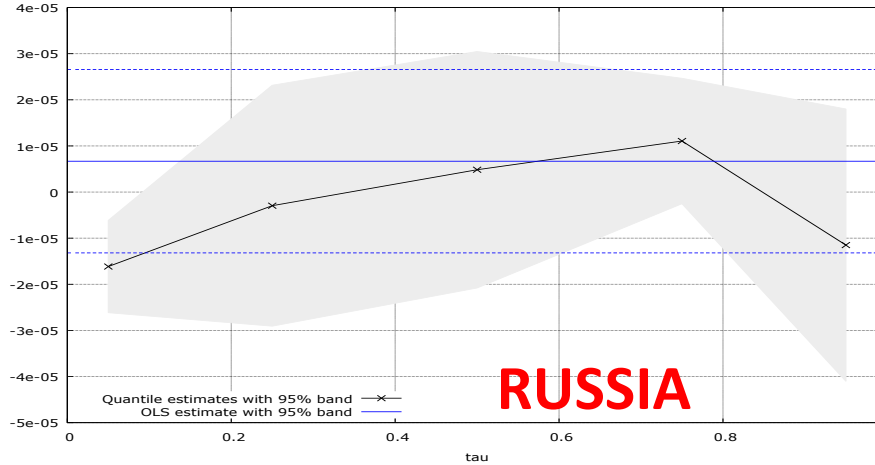


Exchange Rate Volatility on BRICS Exports – Quintile Regression Results

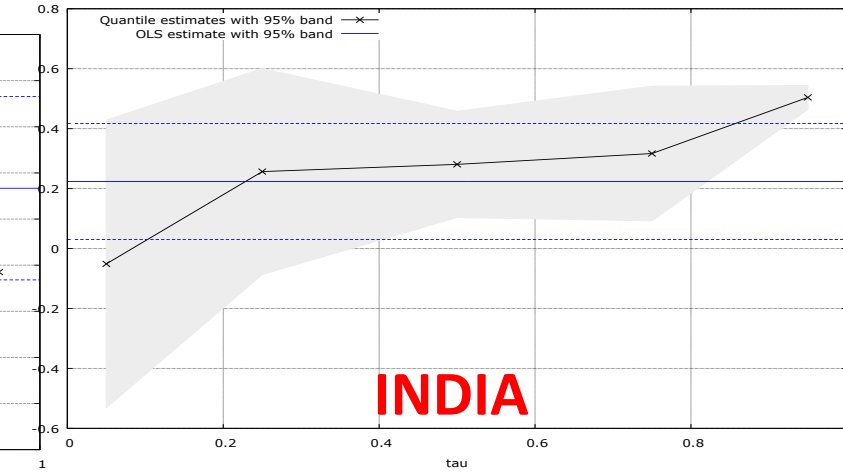
Coefficient on $d_I_Exch_Vol$ on Brazil Exports



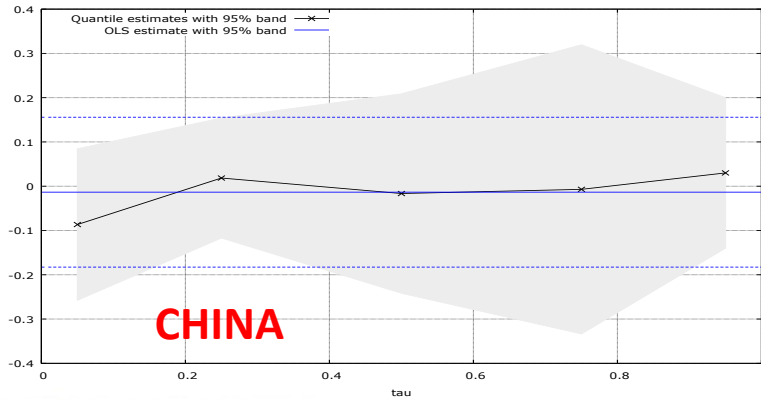
Coefficient on d_Exch_Vol on Russia's Exports



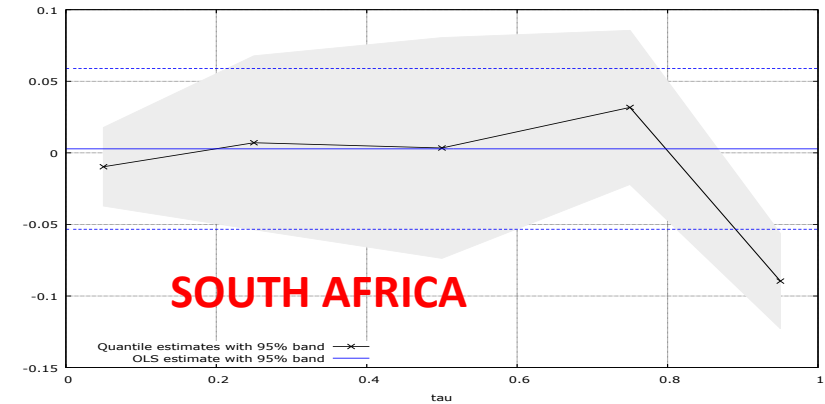
Coefficient on $d_I_Exch_Vol$ on India's Exports



Coefficient on $d_I_Exch_Vol$ on China Exports

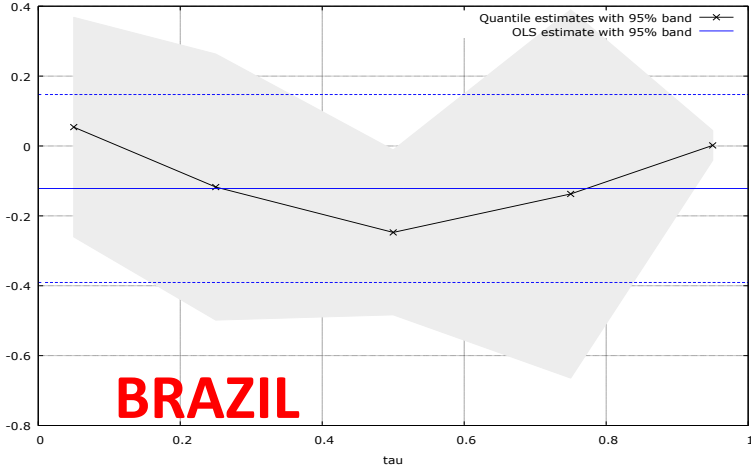


Coefficient on $d_I_Exch_Vol$ on SA Exports

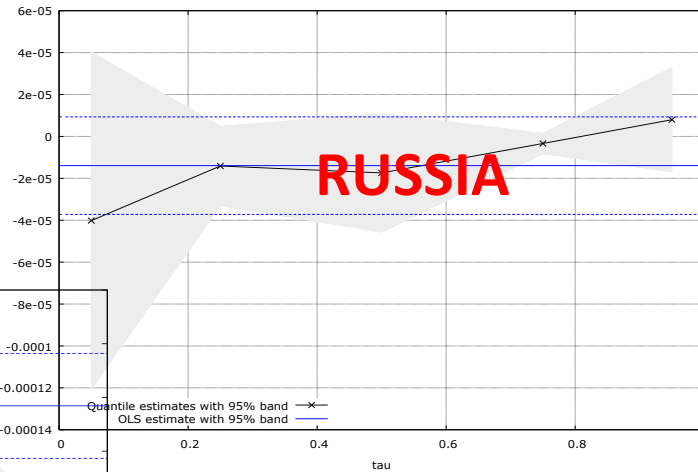


Exchange Rate Volatility on BRICS Imports – Quintile Regression Results

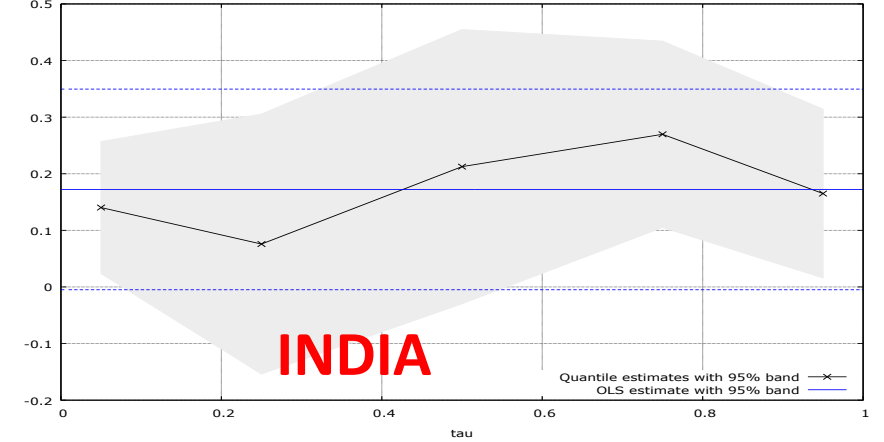
Coefficient on d_I Exch_Vol on Brazil Imports



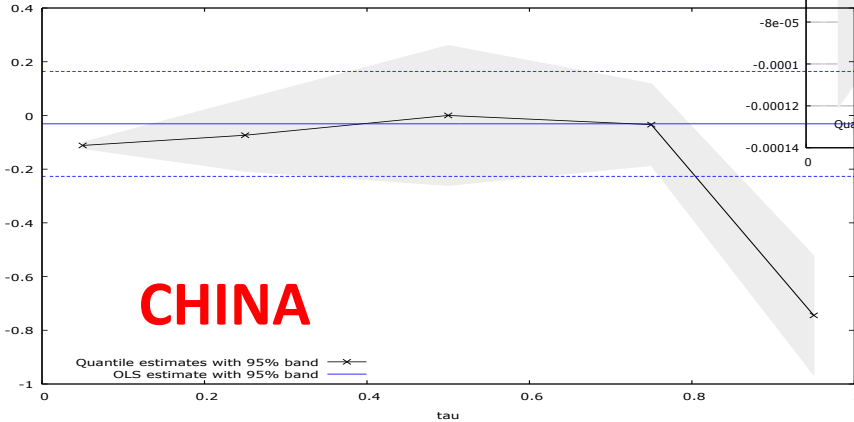
Coefficient on d Exch_Vol on Russia Imports



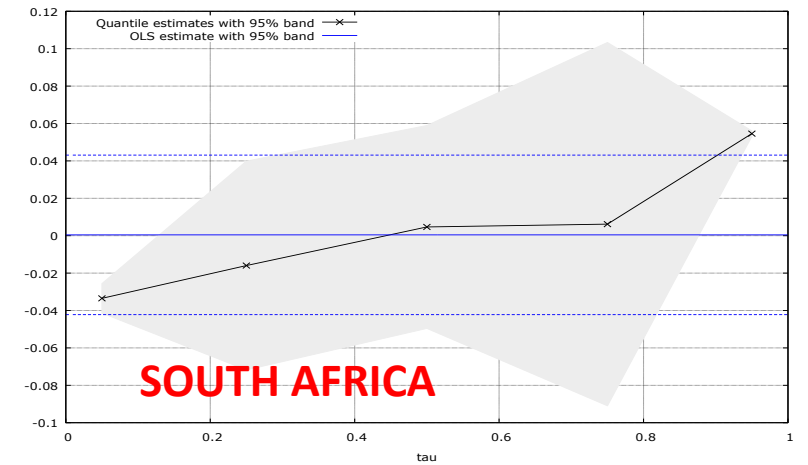
Coefficient on d_I Exch_Vol on India's Imports



Coefficient on d_I Exch_Vol



Coefficient on d_I Exch_Vol on SA Imports



Conclusion

- Despite the best efforts of economists, a basic paradox as to the impact of exchange rate volatility on trade flows remains unresolved at both the theoretical and empirical level (McKenzie & Melbourne, 1999).
- Results are inconclusive due to the heterogeneity of responses to shocks in BRICS economies and this requires further exploration
- Nonetheless, the results suggest that exchange rate volatility has undesirable consequences on trade

Thank You

