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DOMESTIC INVESTMENT, EXPORT, IMPORT AND ECONOMIC GROWTH IN BRAZIL: AN APPLICATION OF VECTOR ERROR CORRECTION MODEL

Sayef Bakari

Department of Economics Sciences, Faculty of Economic Sciences and Management of Tunis, University of Tunis El Manar, (Tunisia)

Email: bakari.sayef@yahoo.fr

Nissar Fakraoui

Department of Economics Sciences, Higher Institute of Companies Administration, University of Gafsa, (Tunisia)

Email: fakraoui.nissar@yahoo.com

Sofien Tiba

Faculty of Economic Sciences and Management of Sfax, University of Sfax, (Tunisia)

Email: sofienetiba@gmail.com

Abstract: This paper aims to investigate the nexus between domestic investment, exports, imports, and economic growth for the Brazilian economy during the period 1970-2017, using the VECM methodology. In the short-run, our empirical results pointed out that import, exports, and domestic investment cause economic growth. Also, economic growth causes exports. Exports, imports, and economic growth cause domestic investment. However, in the long-run, our results revealed that domestic investment and exports have a positive effect on economic growth. Also, imports have a negative effect on economic growth. The results recorded a positive impact of economic growth and imports on domestic investment. Exports have a negative effect on domestic investment. Finally, we record the absence of significant impact of economic growth, exports and domestic investment on imports, and economic growth, domestic investment, and imports on exports. Due to the importance of these aspects to the economic performance of Brazil, the policymakers are invited to orient these issues towards the sustainability facets to guarantee a sustained growth path.

Keywords: Export, Import, Domestic Investment, Economic Growth, VECM, Brazil.

JEL classification: E22, F14, O16, O47, O54.



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1. Introduction

The growth of any economy strongly depends on the stock of domestic capital, which constitutes one of the main driving forces of the investment and value-added creation (See. Tiba et al, 2015, Bakari and Tiba, 2019; Tiba and Frikha, 2019). The neoclassical theory considered the capital accumulation process as the driving force behind a strong economic performance. Indeed, the accumulation of the capital is a fundamental element that will finance the investment, and then the economic growth will improve.

Also, the contribution of the movements of exports and imports are recognized in the creation of wealth and national prosperity due to the positive externalities obtained from the dynamism and interactions with the foreign actors under the globalization era.

Furthermore, the impact of the exports and imports on growth is structured around two hypotheses. The first hypothesis assumes that the export is the driving force of the economy that called the export leads-growth (ELG) hypothesis. Also, it supposes that the causality direction is running from exports to growth. While the alternative hypothesis assumes that the economic growth stimulates the export which is called the Growth lead-exports (GLE). The sense of causality is running from economic growth to exports.

Regarding the linkage between the imports and economic growth, where the imports lead-growth hypothesis assumes that the imports are the driving force of the economy (ILG). Moreover, the sense of causality is running from imports to economic growth. However, the alternative one assumes growth lead-imports (GLI), and the causality direction is from growth to imports. Due to the importance of domestic investment, exports, imports, and growth, the understanding of these controversial nexuses seen as a priority, especially, for the Brazilian case, where it is one of the most emerging economic power and member of the G20.

Our current survey attempts to treat the nexus between domestic investment, exports, imports, and growth for the Brazilian economy over the period 1970-2017. For this purpose, we employ the VECM methodology. To the best of our knowledge, none of the previous studies deal with this question in the Brazilian context through the use of the VECM technique.



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The rest of the paper is structured as follow: Section 2 presents the empirical literature while Section 3 furnishes data and methodology. Empirical findings sited in Section 4. In the last section, we conclude the study and set up some policy suggestions for policymakers.



2. Literature Survey

The following Table presents a set of empirical studies that are collected during our exploitation of this research theme to inspire the realization of our empirical analysis.

Table 1 Studies related to the relationship between domestic investment, export, import and economic growth

No	Authors	Countries	Period	Methodology	Results
1	Albiman and Suleiman (2016)	Malaysia	1967-2010	Cointegration Analysis	Y => DI
		-		Vector Autoregression Model	$X \Rightarrow DI$
				Granger Causality Tests	$X \Rightarrow M$
2	Appiah (2018)	Ghana	1960 - 2015	Cointegration Analysis	DI#Y
				Auto-Regressive Distributive Lags	
				Error Correction Model	
3	Bakari (2017a)	Egypt	1965 - 2015	Cointegration Analysis	$DI \Rightarrow Y LR(-)$
				Vector Error Correction Model	$X \Rightarrow Y: LR(-)$
				Wald Tests	$M \Rightarrow Y: LR$
					$M \Rightarrow Y: SR$
4	Bakari (2017b)	Gabon	1980 - 2015	Cointegration Analysis	$X \Rightarrow Y : LR (-)$
				Error Correction Model	$DI \Rightarrow Y: LR (-)$
				Wald Tests	$X \Rightarrow Y: SR$
					$DI \Rightarrow Y: SR$
5	Bakari (2017c)	Sudan	1976 - 2015	Cointegration Analysis	DI # X # M # Y: LR
				Error Correction Model	$Y \Rightarrow DI : SR$
				Wald Tests	$M \Rightarrow X : SR$
					$Y \Rightarrow M : SR$
6	Bakari (2018)	France	1972 - 2016	Cointegration Analysis	$DI \ll Y: LR (-)$
				Vector Error Correction Model	$DI \Rightarrow Y: SR$



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•				Wald Tests	
7	Bakari et al (2018)	Nigeria	1981 - 2015	Cointegration Analysis	DI # X # M # Y: LR
-	_ 332332	- 1-8	-,	Vector Error Correction Model	M <=> Y: SR
				Wald Tests	$M \Rightarrow DI: SR$
8	Bakari and Mabrouki (2017a)	Panama	1980 - 2015	Cointegration Analysis	$X \Rightarrow Y$
Ü	2 411411 4114 1114010 4111 (20174)	- WIIWIIW	1,00 2016	Vector Autoregression Model	$M \Rightarrow Y$
				Granger Causality Tests	112 / 2
9	Bakari and Mabrouki (2017b)	7 South Eastern	2006 - 2016	Fixed Effect Model	$DI \Rightarrow Y$
	201101 010 11100100111 (20170)	Europe Countries	2000 2010	Random Effect Model	$X \Rightarrow Y$
				Hausman Test	M => Y: (-)
10	Bakari and Tiba (2019)	24 Asian economies	2002 - 2017	Fixed Effect Model	$DI \Rightarrow Y$
				Random Effect Model	M # Y
				Hausman Test	X => Y: (-)
11	Bouchoucha and Bakari (2019)	Tunisia	1976 - 2017	Cointegration Analysis	$DI \Rightarrow Y: LR(-)$
	` '			Auto-Regressive Distributive Lags	$M \Rightarrow Y: LR(-)$
				Wald Tests	$X \Rightarrow Y : LR$
					$DI \Rightarrow Y : SR$
					$X \Rightarrow Y: SR$
12	Mbulawa (2017)	Botswana	1985 - 2015	Ordinary Least Squares	$DI \Rightarrow Y$
				Vector Error Correction Model	
13	Tang and Tan (2015)	Malaysia	1991-2010	Cointegration Analysis	$DI \ll Y$
		•		Auto-Regressive Distributive Lags	
				Vector Autoregression Model	
14	Tahir & Azid (2015)	50 Developing	1990 - 2009	Fixed Effect Model	$DI \Rightarrow Y$
		Countries		Random Effect Model	
				Hausman Test	
15	Umar-Gingo et Demireli (2018)	Ghana	1980 - 2015	Cointegration Analysis	$DI \Rightarrow Y: LR(-)$
				Vector Error Correction Model	DI # Y: SR

Note: X means Exports, M means Imports, Y means Economic Growth, LR means Long Run, SR means Short Run, (+) means Positive

Effect and (-) means Negative Effect.



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3. Econometric methodology and data source

In this study, we use the following variables: economic growth (Y), domestic investment, (DI), exports (X) and Imports (M) for the case of Brazil. All the variables are taken from the online World Development Indicators data of the World Bank. Y is the variable referring to Gross Domestic Product (Constant US \$); DI is the variable referring to Gross Fixed Capital Formation (Constant US \$); X is the variable referring to total exports (Constant US \$); M is the variable referring to total imports (Constant US \$).

Following literature on economic growth {Albiman and Suleiman (2016); Bakari (2017d); Kartikasari (2017); Bakari (2018); Bakari and Ali (2018); Fakraoui and Bakari (2019)}, we can consider the long-run and the short-run relationships between domestic investment, economic growth, exports and imports in linear logarithmic form.

$$Log(Y)_t = \alpha_0 + \alpha_1 Log(DI)_t + \alpha_2 Log(X)_t + \alpha_3 Log(M)_t + \epsilon_t \tag{1}$$

The returns to scale are enclosed with domestic investment, exports and imports which are exposed by α_1 , α_2 and α_3 respectively. In addition, ε is error term and t is time index.

We use recent developments in time series econometrics to analyze the short-term and longrun causal relationships between domestic investment, exports, imports, and economic growth. This approach will be done in three steps: unit root tests, Johansen cointegration tests, Granger causality tests as part of an error-correction vector model.

3.1.Unit root tests

Unit root tests identify the presence of a unit root in a series. A time series is stationary if it has neither trend nor seasonality. The enhanced Dickey-Fuller test is used for this purpose. In this test, we try to verify the null hypothesis against the alternative hypothesis.

3.2. Cointegration Analysis

The study of cointegration makes it possible to test the existence of a stable long-term relationship between two non-stationary variables, including delayed variables and exogenous variables. The analysis of cointegration makes it possible to identify the true relationship



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between two variables, by looking for the existence of a cointegrating vector, and by eliminating its effect if necessary. Two series x and y are cointegrated if the following two conditions are satisfied: they are assigned a stochastic trend of the same order of integration and a linear combination of these series leads to a series of the order of integration more lower. Finally, the Johansen cointegration test uses two statistics: the trace statistic and the maximum Eigenvalue. The asymptotic distributions of these statistics are nonstandard.

3.3. Vector Error Correction Model

The presence of a cointegration relationship between two variables leads to the existence of a causal relation between them in at least one direction. This causal relationship can be analyzed through the Granger causality test (Wald test), which relies on the Vector Error Correction Model. According to Granger's representation theorem, any cointegrated system implies the existence of an error-correcting mechanism that prevents variables from deviating too far from their long-run equilibrium. In addition, one of the important roles of Vector Error Correction Model is exterminating the causal relationship between all variables in the long run and in the short run.

To inspect the existence of a short-term relationship between variables, we will use the Wald Test. If the variable has a probability less than 5% in this case we can say that dependent variable cause independent variable in the short term. However, if the variable has a probability greater than 5% in this case, there is no causal relationship between variables in the short run.

To inspect whether there is a long-term equilibrium relationship in the estimation of vector error correction model (VECM), we use the method of Gauss-Newton linear regression. The coefficient of the Error Correction Term (ECT) must be negative and has a probability of less than 5%.

4. Empirical Results

Stationarity of each variable was checked through ADF with constant, constant and trend, without constant and without a trend. The below Table.2 examined the stationarity of each variable through ADF. The estimated results indicate that Y, DI, X, and M are stationary at

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first difference. The examined results confirm that Johansen Test can be applied with I (1) order.

Table.2 Test for Unit Test ADF

Variables	ADF Test							
	Constant	Constant and Trend	None					
Log (Y)	(4.405565)***	(3.712343)**	(5.823315)					
_	[4.195397]***	[4.631443]***	[3.243802]***					
Log (DI)	(2.717631)*	(3.267304)*	(2.177443)					
_	[4.602554]***	[4.668510]***	[4.433843]***					
Log (M)	(1.033659)	(2.158673)	(2.944798)					
	[4.903941]***	[4.835945]***	[4.475212]***					
Log (X)	(2.346542)	(1.611668)	(5.812467)					
	[6.468603]***	[7.144465]***	[1.588614]					

***; ** and * denote significances at 1%; 5% and 10% levels respectively

Table.3 indicates the results of AIC, SC, and HQ with order test for lag selection based on VAR Model. According to the results of SC and HQ, lag zero is suitable (which is impossible), but AIC indicates that lag Seven is suitable for the model. We used AIC information criterion with indicate that lag seven is suitable for our model.

Table.3 VAR Lag Order Selection

	VAR Lag Order Selection Criteria									
Lag LogL LR FPE AIC SC										
0	259.0410	NA*	3.40e-11*	-12.75205	-12.58316*	-12.69098*				
1	267.0008	13.92967	5.11e-11	-12.35004	-11.50560	-12.04472				
2	278.7118	18.15208	6.51e-11	-12.13559	-10.61560	-11.58601				
3	290.6627	16.13363	8.51e-11	-11.93313	-9.737589	-11.13929				
4	302.6809	13.82099	1.19e-10	-11.73405	-8.862950	-10.69595				
5	318.4573	14.98760	1.52e-10	-11.72287	-8.176219	-10.44051				
6	347.4756	21.76372	1.18e-10	-12.37378	-8.151582	-10.84717				
7	394.0054	25.59140	5.06e-11	-13.90027*	-9.002521	-12.12940				

^{*} indicates lag order selected by the criterion

LR: sequential modified LR test statistic (each test at 5% level)

FPE: Final prediction error

AIC: Akaike information criterion **SC:** Schwarz information criterion

⁽⁾ denotes stationarity in level

^[] denotes stationarity in first difference

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HQ: Hannan-Quinn information criterion

The results of the Johansen co-integration test for series Y, DI, X, and M are reported in Table 3. The likelihood ratio tests show that the null hypothesis of absence of co-integrating relation (R=0= can be rejected at 5% level, but the null hypothesis of the existence of at most one co-integrating relation (R<1) cannot be rejected at the 5% level. This implies that in our case there are three co-integrating equation at the 5% level. Thus, we conclude that Y, DI, X, and M are co-integrated. That is, there is a long-run relationship between economic growth, domestic investment, exports and imports for Brazil.

Table.4 Johansen co-integration test

Unrestricted Cointegration Rank Test (Trace)									
Hypothesized No. of CE(s) Eigenvalue Trace Statistic Critical Value 0.05 Prob.*									
None *	0.942665	180.3875	47.85613	0.0000					
At most 1 *	0.636214	68.89280	29.79707	0.0000					
At most 2 *	0.520610	29.45640	15.49471	0.0002					
At most 3	0.019852	0.782020	3.841466	0.3765					

Trace test indicates 3 cointegrating eqn(s) at the 0.05 level

Since there are 3 cointegrating equations at the 5% level, Vector Error Correction Model can be returned. The estimation of the vector error correction model (VECM) has two steps. The first is to estimate the equation of long-run equilibrium by applying Gauss-Newton linear regression. The second step is to determine the causal links between the different variables in the short term by applying the WALD test.

The equations of the vector error correction model are presented as follows:

$$\begin{split} D(DLOG(Y)) &= C(1)^*(\ DLOG(Y(-1))\ -\ 0.30743433369^*DLOG(X(-1))\ +\ 0.0921173554092^*DLOG(M(-1))\ -\ 0.479377945542^*DLOG(DI(-1))\ -\ 0.00195092631235\)\ +\ C(2)^*D(DLOG(Y(-1)))\ +\ C(3)^*D(DLOG(Y(-2)))\ +\ C(4)^*D(DLOG(Y(-3)))\ +\ C(5)^*D(DLOG(Y(-4)))\ +\ C(6)^*D(DLOG(Y(-5)))\ +\ C(7)^*D(DLOG(Y(-6)))\ +\ C(8)^*D(DLOG(Y(-7)))\ +\ C(9)^*D(DLOG(X(-1)))\ +\ C(10)^*D(DLOG(X(-2)))\ +\ C(11)^*D(DLOG(X(-3)))\ +\ C(12)^*D(DLOG(X(-4)))\ +\ C(13)^*D(DLOG(X(-5)))\ +\ C(14)^*D(DLOG(X(-6)))\ +\ C(15)^*D(DLOG(X(-7)))\ +\ C(16)^*D(DLOG(M(-1)))\ +\ C(17)^*D(DLOG(M(-2)))\ +\ C(18)^*D(DLOG(M(-3)))\ +\ C(19)^*D(DLOG(M(-4)))\ +\ C(20)^*D(DLOG(M(-5)))\ +\ C(21)^*D(DLOG(M(-6)))\ +\ C(22)^*D(DLOG(M(-7)))\ +\ C(23)^*D(DLOG(DI(-1))) \end{split}$$

^{*} denotes rejection of the hypothesis at the 0.05 level

^{**}MacKinnon-Haug-Michelis (1999) p-values



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+ C(24)*D(DLOG(DI(-2))) + C(25)*D(DLOG(DI(-3))) + C(26)*D(DLOG(DI(-4))) + C(27)*D(DLOG(DI(-4))) + C(
  5))) + C(28)*D(DLOG(DI(-6))) + C(29)*D(DLOG(DI(-7))) + C(30)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        (2)
D(DLOG(X)) = C(31)*(DLOG(Y(-1)) - 0.30743433369*DLOG(X(-1)) + 0.0921173554092*DLOG(M(-1)) - 0.092117354092*DLOG(M(-1)) - 0.092117454092*DLOG(M(-1)) - 0.092117454000
0.479377945542*DLOG(DI(-1)) - 0.00195092631235 \ ) + C(32)*D(DLOG(Y(-1))) + C(33)*D(DLOG(Y(-2))) \\
  + C(34)*D(DLOG(Y(-3))) + C(35)*D(DLOG(Y(-4))) + C(36)*D(DLOG(Y(-5))) + C(37)*D(DLOG(Y(-6))) + C(37)*D(DLOG(Y(-6)
  C(38)*D(DLOG(Y(-7))) + C(39)*D(DLOG(X(-1))) + C(40)*D(DLOG(X(-2))) + C(41)*D(DLOG(X(-3))) + C(41)*D(DLOG(X(-3)))
  C(42)*D(DLOG(X(-4))) + C(43)*D(DLOG(X(-5))) + C(44)*D(DLOG(X(-6))) + C(45)*D(DLOG(X(-7))) + C(45)*D(DLOG(X(-7)))
  C(46)*D(DLOG(M(-1))) + C(47)*D(DLOG(M(-2))) + C(48)*D(DLOG(M(-3))) + C(49)*D(DLOG(M(-4))) + C(49)*D(DLOG(M(-4)))
  C(50)*D(DLOG(M(-5))) + C(51)*D(DLOG(M(-6))) + C(52)*D(DLOG(M(-7))) + C(53)*D(DLOG(DI(-1)))
  + C(54)*D(DLOG(DI(-2))) + C(55)*D(DLOG(DI(-3))) + C(56)*D(DLOG(DI(-4))) + C(57)*D(DLOG(DI(-3))) + C(
(5)) + (58)*(DLOG(DI(-6))) + (59)*(DLOG(DI(-7))) + (60)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                 (3)
  D(DLOG(M)) = C(61)*(DLOG(Y(-1)) - 0.30743433369*DLOG(X(-1)) + 0.0921173554092*DLOG(M(-1)) - 0.092117354092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.092117354092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.0921173554092*DLOG(M(-1)) - 0.092117354092*DLOG(M(-1)) - 0.092117454092*DLOG(M(-1)) - 0.0921174540000*DLOG(M(-1)) - 0.092117454000*DLOG(M(-1)) - 0.092117454000*DLOG(M(-1)) - 0.092
  0.479377945542*DLOG(DI(-1)) - 0.00195092631235 + C(62)*D(DLOG(Y(-1))) + C(63)*D(DLOG(Y(-2)))
  +C(64)*D(DLOG(Y(-3)))+C(65)*D(DLOG(Y(-4)))+C(66)*D(DLOG(Y(-5)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(Y(-6)))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(Y(-6))+C(67)*D(DLOG(
  C(68)*D(DLOG(Y(-7))) + C(69)*D(DLOG(X(-1))) + C(70)*D(DLOG(X(-2))) + C(71)*D(DLOG(X(-3))) + C(71)*D(DLOG(X(-3)))
  C(72)*D(DLOG(X(-4))) + C(73)*D(DLOG(X(-5))) + C(74)*D(DLOG(X(-6))) + C(75)*D(DLOG(X(-7))) + C(75)*D(DLOG(X(-7)))
  C(76)*D(DLOG(M(-1))) + C(77)*D(DLOG(M(-2))) + C(78)*D(DLOG(M(-3))) + C(79)*D(DLOG(M(-4))) + C(79)*D(DLOG(M(-4)))
  C(80)*D(DLOG(M(-5))) + C(81)*D(DLOG(M(-6))) + C(82)*D(DLOG(M(-7))) + C(83)*D(DLOG(DI(-1)))
    + C(84)*D(DLOG(DI(-2))) + C(85)*D(DLOG(DI(-3))) + C(86)*D(DLOG(DI(-4))) + C(87)*D(DLOG(DI(-4))) + C(
5))) + C(88)*D(DLOG(DI(-6))) + C(89)*D(DLOG(DI(-7))) + C(90)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                         (4)
D(DLOG(DI)) = C(91)*(DLOG(Y(-1)) - 0.30743433369*DLOG(X(-1)) + 0.0921173554092*DLOG(M(-1))
  -0.479377945542*DLOG(DI(-1)) -0.00195092631235 \ ) + C(92)*D(DLOG(Y(-1))) + C(93)*D(DLOG(Y(-1))) + C(93)*D(DLOG(
  2))) + C(94)*D(DLOG(Y(-3))) + C(95)*D(DLOG(Y(-4))) + C(96)*D(DLOG(Y(-5))) + C(97)*D(DLOG(Y(-5))) + C(97)*D(DLOG(Y(-5)) + C(97)*D(DLOG(Y(-5)) + C(97)*D(DLOG(Y(-5))) + C(97)*D(DLOG(Y(-5))) + C(97)*D(DLOG(Y(
                                                                                                                                                C(98)*D(DLOG(Y(-7))) + C(99)*D(DLOG(X(-1)))
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                          C(100)*D(DLOG(X(-2)))
  C(101)*D(DLOG(X(-3))) + C(102)*D(DLOG(X(-4))) + C(103)*D(DLOG(X(-5))) + C(104)*D(DLOG(X(-5))) + C(10
                                                              + C(105)*D(DLOG(X(-7))) + C(106)*D(DLOG(M(-1))) + C(107)*D(DLOG(M(-2))) +
  C(108)*D(DLOG(M(-3))) + C(109)*D(DLOG(M(-4))) + C(110)*D(DLOG(M(-5))) + C(111)*D(DLOG(M(-5))) + C(111)*D(DLOG(M(-5))) + C(110)*D(DLOG(M(-5))) + C(11
   6))) + C(112)*D(DLOG(M(-7))) + C(113)*D(DLOG(DI(-1))) + C(114)*D(DLOG(DI(-2))) 
  C(115)*D(DLOG(DI(-3)))
                                                                                                                                                                                                                                                                                                                                                                                                                                                        C(116)*D(DLOG(DI(-4)))
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                   C(117)*D(DLOG(DI(-5)))
                                                                                                                                                                                                                                                                                                                                                                               +
  C(118)*D(DLOG(DI(-6))) + C(119)*D(DLOG(DI(-7))) + C(120)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                       (5)
```

The equation of long-run equilibrium is presented as follows:

$$Log(Y) = 0.001951 + 0.479378Log(DI) + 0.307434Log(X) - 0.092117Log(M)$$
 (6)

Table 5: Estimation of VECM (Results of causality in Long run and short run)



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	Log (Y)	Log (M)	Log (X)	Log (DI)
Log (Y)	-	(0.1840)	(0.0834)*	(0.0022)***
Log (M)	(0.0509)**	-	(0.5953)	(0.0124)**
Log (X)	(0.0018)***	(0.2769)	-	(0.0668)*
Log (DI)	(0.0160)**	(0.1217)	(0.4255)	-
ECT	[-7.553584]***	[-13.24822]	[-3.168606]	[-11.55755*]

^{***; **} and * indicate significance at 1%, 5% and 10%, respectively

The estimation of the vector error correction model shows the following results:

In the long run:

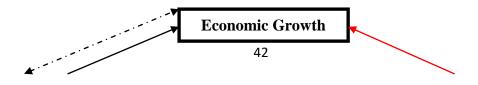
- ✓ Domestic investment and exports have a positive effect on economic growth.
- ✓ Imports have a negative effect on economic growth.
- ✓ Economic growth and imports have a positive effect on domestic investment.
- ✓ Exports have a negative effect on domestic investment.
- ✓ Economic growth, exports, and domestic investment don't have any effect on imports.
- ✓ Economic growth, domestic investment, and imports don't have any effect on exports.

In the short run

- ✓ Imports, exports, and domestic investment cause economic growth.
- ✓ Economic growths cause exports.
- ✓ Exports, imports, and economic growth cause domestic investment.

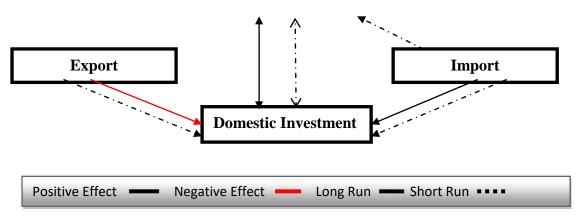
We have summarized these results in Graph 1 below:

Graph.1: The links between the four variables



⁽⁾ denotes the value of the probability of the variables in the short term

^[] denotes the significance of long-term co-integration equations



To verify the credibility and the robustness of our error vector correction model, we apply a set of diagnostic tests. These are the heteroscedasticity tests and the residual autocorrelation test. The results of various diagnostic tests are given in the lower segment of Table 6. The diagnostic tests show that the estimation results are acceptable and that the model meets the MCO application conditions. Indeed, the probabilities of heteroscedasticity tests and the residual autocorrelation test are greater than 5%.

Table.6 Diagnostics tests

Log	(Y)			Log	(X)				
Godfrey Seria	l Correlation LM Test:		Breusch-Godfrey Serial Correlation LM Test:						
0.928774	Prob. F(2,7)	0.4388	F-statistic	1.215117	Prob. F(2,7)	0.3524			
8.178831	Prob. Chi-Square(2)	0.0167	Obs*R-squared	10.05056	Prob. Chi-Square(2)	0.0066			
asticity Test:	Breusch-Pagan-Godfrey		Heterosked	asticity Test:	Breusch-Pagan-Godfrey				
0.254462	Prob. F(36,2)	0.9714	F-statistic	1.647106	Prob. F(36,2)	0.4496			
32.01114	Prob. Chi-Square(36)	0.6588	Obs*R-squared	37.72748	Prob. Chi-Square(36)	0.3902			
2.167196	Prob. Chi-Square(36)	1.0000	Scaled explained SS	1.602586	Prob. Chi-Square(36)	1.0000			
teroskedastic	ity Test: Harvey	Heteroskedasticity Test: Harvey							
0.470240	Prob. F(36,2)	0.8660	F-statistic	1.702904	Prob. F(36,2)	0.4389			
34.87926	Prob. Chi-Square(36)	0.5218	Obs*R-squared	37.76786	Prob. Chi-Square(36)	0.3884			
16.81432	Prob. Chi-Square(36)	0.9973	Scaled explained SS	19.06252	Prob. Chi-Square(36)	0.9908			
teroskedastic	ity Test: Glejser		Не	teroskedastic	ity Test: Glejser				
0.314654	Prob. F(36,2)	0.9464	F-statistic	1.587372	Prob. F(36,2)	0.4616			
33.14746	Prob. Chi-Square(36)	0.6050	Obs*R-squared	37.68122	Prob. Chi-Square(36)	0.3922			
7.872062	Prob. Chi-Square(36)	1.0000	Scaled explained SS	7.370156	Prob. Chi-Square(36)	1.0000			
teroskedastic	rity Test: ARCH		Не	eteroskedastic	ity Test: ARCH				
0.900587	Prob. F(1,36)	0.3490	F-statistic	1.208840	Prob. F(1,36)	0.2789			
0.927419	Prob. Chi-Square(1)	0.3355	Obs*R-squared	1.234544	Prob. Chi-Square(1)	0.2665			
	3.14746 7.872062 0.908774 8.178831 asticity Test: 0.254462 32.01114 2.167196 teroskedastic 0.470240 34.87926 16.81432 teroskedastic	Godfrey Serial Correlation LM Test: 0.928774 Prob. F(2,7) 8.178831 Prob. Chi-Square(2) asticity Test: Breusch-Pagan-Godfrey 0.254462 Prob. F(36,2) 32.01114 Prob. Chi-Square(36) 2.167196 Prob. Chi-Square(36) teroskedasticity Test: Harvey 0.470240 Prob. F(36,2) 34.87926 Prob. Chi-Square(36) teroskedasticity Test: Glejser 0.314654 Prob. F(36,2) 33.14746 Prob. Chi-Square(36) 7.872062 Prob. Chi-Square(36) teroskedasticity Test: ARCH 0.900587 Prob. F(1,36)	0.928774 Prob. F(2,7) 0.4388 8.178831 Prob. Chi-Square(2) 0.0167 asticity Test: Breusch-Pagan-Godfrey 0.254462 Prob. F(36,2) 0.9714 32.01114 Prob. Chi-Square(36) 0.6588 2.167196 Prob. Chi-Square(36) 1.0000 teroskedasticity Test: Harvey 0.470240 Prob. F(36,2) 0.8660 34.87926 Prob. Chi-Square(36) 0.5218 16.81432 Prob. Chi-Square(36) 0.9973 teroskedasticity Test: Glejser 0.314654 Prob. F(36,2) 0.9464 33.14746 Prob. Chi-Square(36) 0.6050 7.872062 Prob. Chi-Square(36) 1.0000 teroskedasticity Test: ARCH 0.900587 Prob. F(1,36) 0.3490	Godfrey Serial Correlation LM Test: Breusch-Omegan (Content of the Content of the Cont	Godfrey Serial Correlation LM Test: Breusch-Godfrey Serial 0.928774 Prob. F(2,7) 0.4388 F-statistic 1.215117 8.178831 Prob. Chi-Square(2) 0.0167 Obs*R-squared 10.05056 asticity Test: Breusch-Pagan-Godfrey Heteroskedasticity Test: 0.254462 Prob. F(36,2) 0.9714 F-statistic 1.647106 32.01114 Prob. Chi-Square(36) 0.6588 Obs*R-squared 37.72748 2.167196 Prob. Chi-Square(36) 1.0000 Scaled explained SS 1.602586 teroskedasticity Test: Harvey Heteroskedastic 1.702904 34.87926 Prob. F(36,2) 0.8660 F-statistic 1.702904 34.87926 Prob. Chi-Square(36) 0.5218 Obs*R-squared 37.76786 16.81432 Prob. Chi-Square(36) 0.9973 Scaled explained SS 19.06252 teroskedasticity Test: Glejser Heteroskedastic 1.587372 33.14746 Prob. Chi-Square(36) 0.6050 Obs*R-squared 37.68122 7.872062 Prob. Chi-Square(36) 1.0000 <t< td=""><td> Breusch-Godfrey Serial Correlation LM Test: Breusch-Godfrey Serial Correlation LM Test: </td></t<>	Breusch-Godfrey Serial Correlation LM Test: Breusch-Godfrey Serial Correlation LM Test:			



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	Log	(M)		Log	(DI)				
Breusch-C	Godfrey Serial	l Correlation LM Test:		Breusch-Godfrey Serial Correlation LM Test:					
F-statistic	0.663073	Prob. F(2,7)	0.5449	F-statistic	0.870781	Prob. F(2,7)	0.4595		
Obs*R-squared	6.211721	Prob. Chi-Square(2)	0.0448	Obs*R-squared	7.769885	Prob. Chi-Square(2)	0.0205		
Heterosked	asticity Test:	Breusch-Pagan-Godfrey		Heterosked	asticity Test:	Breusch-Pagan-Godfrey			
F-statistic	1.303012	Prob. F(36,2)	0.5284	F-statistic	0.237973	Prob. F(36,2)	0.9771		
Obs*R-squared	37.40518	Prob. Chi-Square(36)	0.4044	Obs*R-squared	31.61854	Prob. Chi-Square(36)	0.6770		
Scaled explained SS	2.883968	Prob. Chi-Square(36)	1.0000	Scaled explained SS	2.027724	Prob. Chi-Square(36)	1.0000		
Не	teroskedastici	ity Test: Harvey		Heteroskedasticity Test: Harvey					
F-statistic	1.569472	Prob. F(36,2)	0.4653	F-statistic	0.592470	Prob. F(36,2)	0.8008		
Obs*R-squared	37.66669	Prob. Chi-Square(36)	0.3929	Obs*R-squared	35.65651	Prob. Chi-Square(36)	0.4848		
Scaled explained SS	28.84370	Prob. Chi-Square(36)	0.7958	Scaled explained SS	34.48312	Prob. Chi-Square(36)	0.5408		
Не	teroskedastici	ity Test: Glejser		Не	teroskedastic	ity Test: Glejser			
F-statistic	1.871071	Prob. F(36,2)	0.4094	F-statistic	0.288570	Prob. F(36,2)	0.9580		
Obs*R-squared	37.87541	Prob. Chi-Square(36)	0.3838	Obs*R-squared	32.70385	Prob. Chi-Square(36)	0.6262		
Scaled explained SS	10.50250	Prob. Chi-Square(36)	1.0000	Scaled explained SS	8.099435	Prob. Chi-Square(36)	1.0000		
Не	teroskedastic	ity Test: ARCH	Не	eteroskedastic	ity Test: ARCH				
F-statistic	0.027791	Prob. F(1,36)	0.8685	F-statistic	0.504870	Prob. F(1,36)	0.4819		
Obs*R-squared	0.029312	Prob. Chi-Square(1)	0.8641	Obs*R-squared	0.525548	Prob. Chi-Square(1)	0.4685		

5. Conclusion

We attempt to treat the nexus between domestic investment, exports, imports, and economic growth for the emerging power of South America the Brazilian case over the period 1970-2017. In our framework, we use the VECM technique.

The short-run analysis pointed out that imports, exports, and domestic investment cause economic growth. The findings outline that economic growth cause exports. Finally, the highlights reveal that exports, imports, and economic growth cause domestic investment.

In the long-run, our results pointed out a positive influence of the domestic investment and exports on economic growth. However, the imports exert a negative impact on economic growth. Also, the short-run findings recorded that economic growth and imports have positive effect on domestic investment. Moreover, the results revealed a negative impact of exports on domestic investment in the Brazilian economy. However, economic growth, exports, and



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domestic investment have no significant impact on imports. Also, economic growth, domestic investment, and imports have no significant impact on exports.

From this outlook, our results pointed out the importance of exports, importance, and domestic investment to provide the phenomenal economic performance of the emerging economy such as Brazil as one of the G20 member States. Hence, the importance of the contribution of exports, imports, and domestic investment in economic growth following the neoclassical model. Besides, Brazil is invited to orient their exports, importance, and domestic investment towards sustainable goods and practices under the Sustainable Development Goals (SDGs) background of the United Nations.

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