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## **Structural Gravity Model Estimation Using India-ASEAN Trade Data & Counterfactual Analysis**

### **Introduction:**

India's FTA with ASEAN (Association of Southeast Asian Nations) called **AIFTA** (ASEAN-India Free Trade Area) is one such FTA that has captured significant attention. It was signed in 2009 and came into effect in 2010, creating a free trade area among member countries. The paper aims to analyze its effect on India's global trade and its trade flows both with the ASEAN countries, as well as with other non-member trading partners using a structural gravity model. As India and ASEAN countries continue to develop their economies, their relationship has become a crucial component of their foreign policies. Hence with the objective of increasing economic integration with ASEAN, AIFTA was signed. The Structural Gravity model presented by Anderson and Wincoop in 2003 is used in this research to explain the pattern of trade between ASEAN countries and India. The Anderson and Wincoop gravity equation modifies the original McCallum gravity equation by considering trade expenses and transportation costs in addition to the country border impacts. In this model prices are not equalized among the countries. Ten countries currently comprise the ASEAN or Association of Southeast Asian Nations including Brunei, Cambodia, Indonesia, Laos, Malaysia, Myanmar, Philippines, Singapore, Thailand and Vietnam. We have 27 countries in total - 10 ASEAN member countries, one India and 16 other countries. There would be 729 observations as a result of  $27 \times 27$ .

If India ratifies the ASEAN agreement by simulating trade patterns, we would also employ the counterfactual approach to examine the welfare consequences. This could be done by using the 'gegravity' python package to estimate general equilibrium (GE) structural gravity models and simulate counterfactual experiments. The package is based on the well established version of the gravity model described by Yotov et al (2016).

### **Objective:**

The primary objectives of this term paper are:

1. Estimation of a Structural Gravity Model: The first objective is to estimate a structural gravity model using trade data between India and the Association of Southeast Asian Nations (ASEAN) member states. By employing this econometric framework, we aim to identify and quantify the key determinants driving bilateral trade flows between India and ASEAN countries. This includes factors such as economic size (proxied by GDP), distance (geographical and non-geographical), and other relevant variables that may influence trade patterns.
2. Conducting Counterfactual Analysis: The second objective is to conduct counterfactual analysis based on the estimated structural gravity model. Through this analysis, we seek to explore the potential impact of various policy changes or external shocks on India-ASEAN trade dynamics. By simulating different scenarios and assessing their effects on trade flows, we aim to provide insights into the resilience of India-ASEAN trade relations and the potential implications for regional economic integration.

## **Literature Review**

In this section, we encapsulate some of the previous studies carried out which used the gravity model for estimating the effects of trade agreements on trade flows. The usage of the gravity model has been relatively recent and has also been challenged due to the absence of weak theory to support it. Most of the literary work carried out pertains to the NAFTA, Latin American, and EU agreements, because of its prior existence. However, in later years we do have studies to analyze the effects of the ASEAN-FTA. A series of such studies have documented the increasing trade among members, as well as with non-members. These include Hur and Nandasiri (2008), Ekanayake and Mukherjee (2010), and Eicher, Henn and Papageorgiou (2012). There are inconsistent results on the trade creation dummy, with some papers like Dee and Gali (2005), and Soloaga and Winters (2001) reporting it to be negative, while Elliot and Ikemoto (2004) reported it to be positive. Interestingly, while the agreements have proven to be beneficial for China, they were unfruitful for India. Note that the majority of these studies have utilized a double or triple indexed model to differentiate the impact on members and non-members due to FTA. On the other hand, a recent paper by Subhash Jagambe and Elumulai Kannan (2020), analyzing the impact of ASEAN-India FTA on agricultural trade have found positive results for India. Using the more reliable PPML method, they have shown a higher magnitude of the TC effect as compared to the TD effect.

## Data Sources

The amount of exports for each country to several other countries was taken from World Integrated Trade Solutions (WITS) database. We obtained the distance between countries using the CEPII (Center for Prospective Studies & International Information) database. We used the World Bank database to download the GDP (Gross Domestic Product) of each country. Common language, common colony, colony and common border dummies were obtained from the same CEPII database. Bilateral Trade flows as reported by the origin (1000 current USD) were obtained from UN Comtrade. We also obtained bilateral trade flows from the International Monetary Fund (IMF).

## Model Description

The OneSectorGE model in the *gegravity* package replicates the structural model of Yotov et al. (2016). That model is based on the earlier demand-side, constant elasticity of substitution (CES)-Armington structural gravity model of Anderson and van Wincoop (2003). As shown by Arkolakis (2012), the structural gravity model can be derived from a wide range of different trade models such as the canonical supply-side, Ricardian version of Eaton and Kortum (2002). Thus, this particular version of the model can be considered reflective of a much more general class of trade models. For the sake of parsimony, I keep the theoretical discussion short as the ‘*gegravity*’ package is merely an implementation of an existing model and makes no theoretical contributions of its own. For more details and discussion of the model, I refer the reader to Yotov et al. (2016) and Anderson et al. (2018). The model system takes the following form:

$$\begin{aligned} X_{ij} &= \frac{Y_i E_j}{Y} \left( \frac{\tau_{ij}}{\Pi_i P_j} \right)^{1-\sigma}, \\ \Pi_i^{1-\sigma} &= \sum_j \left( \frac{\tau_{ij}}{P_j} \right)^{1-\sigma} \frac{E_j}{Y}, \\ P_j^{1-\sigma} &= \sum_i \left( \frac{\tau_{ij}}{\Pi_i} \right)^{1-\sigma} \frac{Y_i}{Y}, \\ p_i &= \left( \frac{Y_i}{Y} \right)^{\frac{1}{1-\sigma}} \frac{1}{\gamma_i \Pi_i}, \end{aligned}$$

First equation is a typical gravity equation, which relates bilateral trade ( $X_{ij}$ ) between exporter 'i' and importer 'j' to exporter output ( $Y_i$ ), importer expenditures ( $E_j$ ), global output ( $Y$ ), bilateral trade costs ( $\tau_{ij}$ ), the elasticity of substitution ( $\sigma$ ) and outward and inward multilateral resistances ( $\Pi_i$  and  $P_j$  respectively). The multilateral resistance (MR) terms are defined by second and third equations. These terms can be thought of as aggregate trade cost or price indices for the exporter and importer. Fourth equation defines factory gate prices ( $p_i$ ), which are determined by output, OMRs, and the CES preference parameter ( $\gamma_i$ ).

$$E_i = \phi_i Y_i = \phi_i p_i Q_i.$$

Finally the last equation determines expenditures and provides a market clearing condition. Expenditures are determined as a fixed ratio ( $\phi_i$ ) of the value of domestic production. Domestic production is defined by the product of output quantity ( $Q_i$ ) and factory gate prices. In this version of the model, output quantity is fixed/exogenous and all changes in the value of output are captured through the price term.

## Methodology

We used the Poisson Pseudo-Maximum Likelihood Estimator for estimation of the gravity model. It is the same as applying a specific kind of nonlinear least squares to the initial equation. Poisson Probability Mass Function given by :  $(\lambda k! e^{-\lambda}) / (k!)$  For applied policy researchers utilizing gravity models, the Poisson estimator provides a number of extra useful characteristics. First off, it makes sense when there are fixed effects, which may be inserted as dummy variables just like in a straightforward OLS model. Second, observations for which the observed trade values 0 are naturally included in the Poisson estimator. Third, the interpretation of the Poisson model's coefficients is simple and adheres to the same pattern as OLS. The Poisson model's coefficients may be easily understood and are interpreted in exactly the same way as they would be under OLS. The coefficients of any independent variables entered in logarithms can still be regarded as simple elasticities even when the dependent variable for the Poisson regression is stated as exports in levels rather than in logarithms. As with OLS, the independent variable coefficients entered in levels are read as semi-elasticities. Poisson has the further benefit of readily supporting counterfactual simulations while adhering to key empirical limitations.

## Results

The results obtained after estimating the structural gravity model are as follows:

<b>Dep. Variable:</b>	Exports	<b>No. Observations:</b>	695
<b>Model:</b>	GLM	<b>Df Residuals:</b>	635
<b>Model Family:</b>	Poisson	<b>Df Model:</b>	59
<b>Link Function:</b>	Log	<b>Scale:</b>	1.0000
<b>Method:</b>	IRLS	<b>Log-Likelihood:</b>	-5.2427e+08
<b>Date:</b>	Wed, 17 Apr 2024	<b>Deviance:</b>	1.0485e+09
<b>Time:</b>	18:26:50	<b>Pearson chi2:</b>	1.23e+09
<b>No. Iterations:</b>	20	<b>Pseudo R-squ. (CS):</b>	1.000
<b>Covariance Type:</b>	HC1		

The coefficient estimates for independent variables are as follows:

	coef	std err	z	P> z	[0.025	0.975]
<b>contig_x</b>	0.5991	0.118	5.064	0.000	0.367	0.831
<b>comlang_off_x</b>	-0.0749	0.199	-0.376	0.707	-0.466	0.316
<b>Indist_x</b>	-0.5215	0.047	-11.136	0.000	-0.613	-0.430
<b>international</b>	17.5608	0.575	30.562	0.000	16.435	18.687
<b>asean</b>	0.0768	0.167	0.459	0.646	-0.251	0.404
<b>comlang_ethno_x</b>	0.1779	0.149	1.191	0.234	-0.115	0.471
<b>col45_x</b>	-0.1839	0.210	-0.875	0.381	-0.596	0.228
<b>aifta</b>	0.3205	0.219	1.463	0.144	-0.109	0.750

The coefficient for `contig_x` is positive, indicating that borders play an important role in increasing trade flows amongst the countries. Two countries close to the border share a cultural history which facilitates increased trade flows.

The coefficients for `comlang_off_x` & `comlang_ethno_x` are negative & positive respectively.

The coefficient for `Indist_x` is negative, incorporating the fact that as distance between two trading partners increases, trade cost also increases. As a result there is a decrease in trade flows between the countries.

The coefficient for 'asean' is positive but insignificant, which implies that there is no significant impact of the ASEAN agreement on trade flows.

But the coefficient for 'aifta' is positive and significant which implies that the trade flows between India and ASEAN countries due to the agreement signed in 2009-10 have increased significantly.

The baseline Inward Multilateral Resistance (IMR) and Outward Multilateral Resistance (OMR) are obtained to be:

	baseline $\tilde{omr}$	baseline imr
country		
ARE	0.039518	1.014032
AUS	0.041071	1.053883
BRN	0.036760	0.943275
CAN	0.039195	1.005746
CHL	0.043439	1.114665
CHN	0.036813	0.944630
COL	0.042806	1.098405
DEU	0.039020	1.001264
FRA	0.039501	1.013605
IDN	0.037113	0.952317
IND	0.035960	0.922736
JPN	0.039772	1.020566
KHM	0.034343	0.881249
KOR	0.038378	0.984783
LAO	0.034457	0.884182
LKA	0.038885	0.997797
MMR	0.035315	0.906195
MYS	0.035002	0.898159
NZL	0.041831	1.073404
PAK	0.037606	0.964968
PHL	0.037522	0.962829
RUS	0.039856	1.022711
SGP	0.036016	0.924188
THA	0.034316	0.880563
USA	0.038971	1.000000
VNM	0.034719	0.890900
ZAF	0.041734	1.070908

After implementing the counterfactual that Indian joins the ASEAN nations, the results obtained are:

	Factory Gate Price	IMR	OMR	Real GDP	Foreign Exports	Foreign Imports
ARE	-0.039	1.014	0.040	-0.062	-0.037	-0.037
AUS	-0.028	1.054	0.041	-0.039	-0.037	-0.037
BRN	0.026	0.943	0.037	0.069	0.031	0.031
CAN	-0.018	1.006	0.039	-0.020	-0.037	-0.037
CHL	-0.021	1.115	0.043	-0.026	-0.037	-0.037
CHN	-0.047	0.945	0.037	-0.076	-0.037	-0.037
COL	-0.020	1.098	0.043	-0.024	-0.037	-0.037
DEU	-0.022	1.001	0.039	-0.028	-0.037	-0.037
FRA	-0.022	1.014	0.040	-0.026	-0.037	-0.037
IDN	0.027	0.952	0.037	0.071	0.032	0.032
IND	0.763	0.916	0.036	1.549	0.732	0.732
JPN	-0.030	1.021	0.040	-0.042	-0.037	-0.037
KHM	0.017	0.881	0.034	0.051	0.025	0.025
KOR	-0.033	0.985	0.038	-0.049	-0.037	-0.037
LAO	0.021	0.884	0.034	0.060	0.032	0.032
LKA	-0.043	0.998	0.039	-0.068	-0.037	-0.037
MMR	0.081	0.905	0.035	0.179	0.117	0.117
MYS	0.018	0.898	0.035	0.053	0.026	0.026
NZL	-0.026	1.074	0.042	-0.036	-0.037	-0.037
PAK	-0.097	0.966	0.038	-0.177	-0.038	-0.038
PHL	0.040	0.962	0.038	0.098	0.045	0.045
RUS	-0.029	1.023	0.040	-0.041	-0.037	-0.037
SGP	0.030	0.924	0.036	0.078	0.038	0.038
THA	0.017	0.880	0.034	0.051	0.030	0.030
USA	-0.017	1.000	0.039	-0.017	-0.037	-0.037
VNM	0.020	0.891	0.035	0.057	0.026	0.026
ZAF	-0.030	1.071	0.042	-0.043	-0.037	-0.037



Exporter	Importer	Baseline Modeled Trade	Experiment Modeled Trade	Trade Change
IND	KHM	70246681	74041362	5.40
IND	THA	76122285	80234281	5.40
IND	MYS	71569920	75433540	5.40
IND	VNM	71114514	74948988	5.39
IND	LAO	78601597	82836848	5.39
IND	BRN	77632140	81803553	5.37
IND	IDN	78575988	82795187	5.37
IND	SGP	85300611	89872504	5.36
IND	PHL	93410888	98387478	5.33
IND	MMR	174334148	183400146	5.20
CHN	PAK	116507317	116907316	0.34
PAK	CHN	116507317	116907316	0.34
LKA	PAK	120284147	120682983	0.33
PAK	LKA	120284147	120682983	0.33
PAK	ARE	185804799	186402546	0.32
ARE	PAK	185804799	186402546	0.32
PAK	KOR	86615339	86877455	0.30
KOR	PAK	86615339	86877455	0.30
PAK	ZAF	101254098	101551898	0.29
ZAF	PAK	101254098	101551898	0.29

We observed that the trade change is positive and more than 5% for each and every IND & ASEAN member nation. While the trade change for other countries' pairs is less than 1%.

If India joins ASEAN, this will trigger a positive shift in factory gate prices for India & ASEAN countries.

GDP change is positive for the ASEAN member nations. This is due to market expansion, larger consumer base and potential for higher export markups.

While for the non-members there are reduced trade flows. This also incentivizes the non-members to pursue bilateral agreements with India.

Both foreign exports and foreign imports increase for the ASEAN member nations.

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