

The World's Largest Open Access Agricultural & Applied Economics Digital Library

This document is discoverable and free to researchers across the globe due to the work of AgEcon Search.

Help ensure our sustainability.

Give to AgEcon Search

AgEcon Search
http://ageconsearch.umn.edu
aesearch@umn.edu

Papers downloaded from **AgEcon Search** may be used for non-commercial purposes and personal study only. No other use, including posting to another Internet site, is permitted without permission from the copyright owner (not AgEcon Search), or as allowed under the provisions of Fair Use, U.S. Copyright Act, Title 17 U.S.C.

No endorsement of AgEcon Search or its fundraising activities by the author(s) of the following work or their employer(s) is intended or implied.

Preferential Trading in Agriculture: New Insights from a Structural Gravity Analysis and Machine Learning
Dongin Kim
Selected presentation for the International Agricultural Trade Research Consortium's (IATRC's) 2022 Annual Meeting: Transforming Global Value
Chains, December 11-13, 2022, Clearwater Beach, FL.
Copyright 2022 by Dongin Kim. All rights reserved. Readers may make verbatim copies of this document for non-commercial purposes by any means, provided that this copyright notice appears on all such copies.

Preferential Trading in Agriculture: New Insights from a Structural Gravity Analysis and Machine Learning

Dongin Kim

Agricultural and Resource Economics
University of Connecticut

IATRC Annual Meeting, December 13, 2022



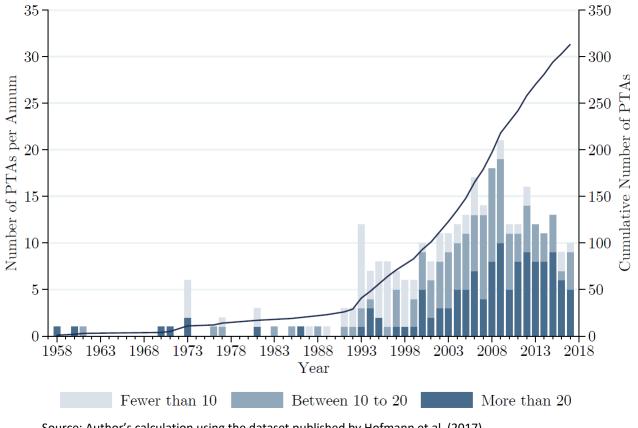
Overview

- ✓ I study the impact of Preferential Trading Agreements (PTAs) in agriculture.
 - Focusing on non-tariff provisions that might create heterogeneity
- ✓ Using a <u>machine learning selection algorithm</u> to overcome the potential econometrics challenges dealing with many policy variables
 - I find the 7 most relevant provisions affecting agricultural trade
 - Which capture most of PTA's agricultural trade effects.
- ✓ In the <u>structural gravity general equilibrium</u> context, the PTA may increase the agricultural GDP:
 - If PTA contains the above-mentioned selected provisions
 - And/or if countries have more PTA partners.

Motivation

- ✓ Countries have signed preferential trade agreements (PTAs)
 - Numbers have increased rapidly since 1990
- ✓ PTAs contain hundreds of provisions
 - Policy areas covered in PTAs increased
 - Number of provisions significantly varies across PTAs
- ✓ Non-tariff measures are more prominent in agriculture than in other sectors

PTA Enforcements by Year

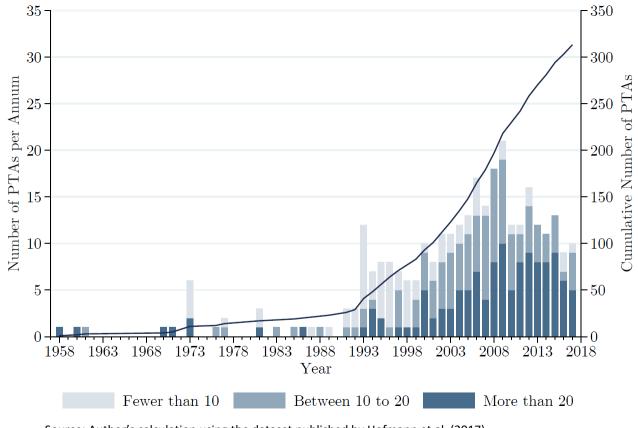


Source: Author's calculation using the dataset published by Hofmann et al. (2017).

Motivation

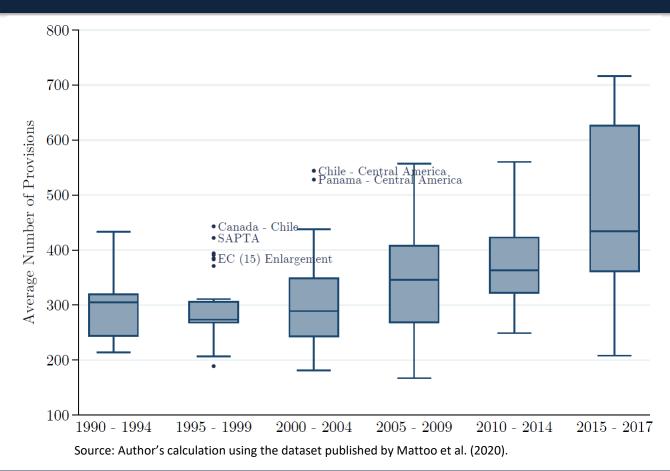
- ✓ Countries have signed preferential trade agreements (PTAs)
 - Numbers have increased rapidly since 1990
- ✓ PTAs contain hundreds of provisions
 - Policy areas covered in PTAs increased
 - Number of provisions significantly varies across PTAs
- ✓ Non-tariff measures are more prominent in agriculture than in other sectors

Policy Coverage



Source: Author's calculation using the dataset published by Hofmann et al. (2017).

Number of Non-tariff Provisions in PTAs



Motivation

- ✓ Countries have signed preferential trade agreements (PTAs)
 - Numbers have increased rapidly since 1990
- ✓ PTAs contain hundreds of provisions
 - Policy areas/Provisions covered in PTAs increased
 - Number of provisions significantly varies across PTAs
- ✓ Non-tariff measures are more prominent in agriculture than in other sectors

Motivation

- ✓ Non-tariff provisions that might affect agricultural trade
 - SPS and TBT:

Disdier et al. (2008), Peterson et al. (2013), Murina and Nicita (2017)

- Subsides and Competition Policies (c.f., non-discrimination principle):
 Bagwell and Staiger (2004), Limao (2006)
- **Intellectual Property Rights** (i.e., geographical indication): Huysmans and Swinnen (2019), Duvaleix et al. (2021), Curzi and Huysmans (2022)
- Others (e.g., environmental law)

Research Questions

- ✓ Are there heterogeneous trade effects of enforcing PTAs
 - Due to the different provisions?
 - What provisions affect agricultural trade?
- ✓ How much do countries gain after enforcing PTAs?
 - Partnered/non-partnered countries
 - Containing/not-containing relevant provisions
- ➤ I test with theoretically-consistent estimators using recently published datasets and a data-driven estimation strategy

Literature Review

- ✓ PTA's trade effects on agriculture
 - Grant and Lambert (2008), Sun and Reed (2010), Scoppola et al. (2018), etc.
 - What has been missing: Identify non-tariff impacts of PTAs
- ✓ Heterogeneous impacts of PTAs
 - Recent studies established the importance of non-tariff provisions in determining PTAs' overall impact (Hofmann et al., 2017; Hofmann et al., 2019)
 - Overfitting and multicollinearity issues arise due to many policy variables (Kohl and Garrestsen, 2016; Santeramo and Lamonaca, 2021)
 - > Reduce dimension with ad hoc assumptions
 - Select the provisions (e.g., Campi et al., 2019)
 - Create indices (e.g., Hofmann et al., 2017; Mattoo et al., 2022)
 - > Rely on data-driven approach
 - Breinlich et al., 2021

Estimation Strategy

- ✓ Estimation Strategy
 - 1. Reduce the policy variable dimension using data-driven method (Regularization regression)
 - Select the most relevant PTA provisions affecting agricultural trade
 - 2. Quantify the impact of PTA and the provisions on agricultural trade

Data

- ✓ Bilateral trade data
 - > ITPD-E from USITC by Borchert et al. (2021)
 - From 1988 to 2017
 - 249 exporters and 251 importers
 - Include intra-national trade needed to obtain theory-consistent estimators
 - Used in many recent studies (e.g., Campos et al., 2021; Rideley et al., 2022; Timini et al., 2022)
- ✓ Policy data
 - > Deep Trade Agreements from World Bank by Mattoo et al. (2020)
 - 937 provisions under 18 most frequently covered policy areas for 282/317 PTAs
 - All provisions are dummy variables (only indicate the presence)
 - Excluded observations for the countries associated with the 35 unmapped-PTAs
 - \rightarrow the final data includes 213 exporters and importers.

Gravity Model

✓ Armington-CES Gravity Model:

$$X_{ij} = \tau_{ij}^{-\Theta} \frac{Y_i}{\pi_i^{-\Theta}} \frac{E_j}{\rho_i^{-\Theta}}$$

- X_{ij} is agricultural exports from country i to j
- τ_{ij} is the bilateral trade cost
- Y_i is country i's total agricultural production
- E_i is country j's aggregate expenditure
- π_i and ρ_i are the multilateral resistance terms
- θ represents the elasticity of substitution

Structural Gravity Model

✓ Theoretically-consistent Empirical Model

$$\mu_{ijt} \coloneqq E(X_{ijt}|\mathbf{PTA}'_{ijt-1},\alpha_{it},\gamma_{jt},\delta_{ij}) = \exp(\mathbf{PTA}'_{ijt-1}\boldsymbol{\beta}' + \alpha_{it} + \gamma_{jt} + \delta_{ij})$$

- X_{ijt} is bilateral agricultural trade from country i to j at year t
- α_{it} and γ_{it} encompass the time-varying country dummy variables
- δ_{ij} denotes the set of country-pair fixed effects
- PTA'_{ijt-1} represents the first-differenced vector of PTA provisions
- Poisson PML to estimate $\boldsymbol{\beta}'$
 - BUT we don't want to estimate with 937 variables
 - ightharpoonup We need to reduce the dimension $PTA_l \subseteq PTA'$

How Regularization Regression Works?

- ✓ How Lasso regression selects the variables?
 - It is a method for penalizing extra variables that are not significantly improve the model specification

$$\arg\min_{\beta_l} \quad RSS + \lambda \sum_{l=1}^m |\beta_l|$$
 "Adding one more variable": $\quad - \quad \quad + \quad \quad + \quad \quad$

✓ I use PPML version of Lasso suggested by Breinlich et al. (2021)

Results: Selected PTA Provisions (Agriculture)

	PPML	Lasso	Post-Lasso	PPML
PTA - Presence of PTA	0.240*** (0.036)			0.022 (0.028)
ENV.19 - Prevent Pollution by Ships		0.052	-0.048 (0.042)	-0.040 (0.045)
ET.06 - Require Scheduling of EQ/QR		0.014	0.176*** (0.061)	0.174*** (0.062)
SPS.44 - Include SPS Chapter		0.052	0.079*** (0.018)	0.096*** (0.027)
STE.39 - Include Sector-specific Discipline on STE		0.107	0.162*** (0.056)	0.175*** (0.060)
SUB.16 - Require for Prior Notification on Subsidies		0.004	0.189*** (0.030)	0.191*** (0.030)
TBT.06/33 - Use Regional Standards		0.091	0.197** (0.095)	0.191** (0.096)
TF.32 - Mutual Recognition of AOs		0.251	0.451** (0.058)	0.448** (0.058)
Observation	798,239	786,621	786,821	786,821
Pseudo R^2	0.994		0.994	0.994

Note: Heteroskedasticity-robust standard errors are in the parentheses, clustered by exporter-importer pair. For the estimated parameters, the semi-elasticity is given by $100*(\exp(\beta) - 1)\%$.

Results: Other Sectors (Processed Food)

	PPML	Lasso	Post-Lasso	PPML
PTA - Presence of PTA	0.189*** (0.036)			-0.011 (0.023)
CP.31 - Create Regional/Agreement-specific Authority		0.058	0.104* (0.062)	$0.100 \\ (0.062)$
ENV.19 - Prevent Pollution by Ships		0.018	0.052 (0.034)	0.056 (0.036)
ET.06 - Require Scheduling of EQ/QR		0.131	0.198*** (0.047)	$0.201^{***} \ (0.048)$
IPR.43 - Designate the List of GI Products		0.012	0.203*** (0.019)	0.195*** (0.023)
SPS.44 - Include SPS Chapter		0.014	0.111*** (0.016)	0.119*** (0.023)
TBT.06/33 - Use Regional Standards		0.438	0.397*** (0.057)	0.398*** (0.057)
TF.32 - Mutual Recognition of AOs		0.069	0.239*** (0.042)	0.241*** (0.043)
Observation	957,886	946,580	946,580	946,580
Pseudo R^2	0.996		0.996	0.996

Note: Heteroskedasticity-robust standard errors are in the parentheses, clustered by exporter-importer pair. For the estimated parameters, the semi-elasticity is given by $100*(\exp(\beta) - 1)\%$.

Results: Other Sectors (Non-food Manufacturing)

	PPML	Lasso	Post-Lasso	PPML
PTA - Presence of PTA	0.109 (0.077)			0.013 (0.091)
CP.31 - Create Regional/Agreement-specific Authority		0.031	0.294*** (0.055)	0.287*** (0.075)
ET.06 - Require Scheduling of EQ/QR		0.008	0.188*** (0.051)	0.184*** (0.057)
Observation	1,217,498	1,206,138	1,206,138	1,206,138
Pseudo R^2	0.996		0.991	0.991

Note: Heteroskedasticity-robust standard errors are in the parentheses, clustered by exporter-importer pair. For the estimated parameters, the semi-elasticity is given by $100*(exp(\beta) - 1)\%$.

General Equilibrium Analysis

✓ Structural Gravity System (*conditional GE described in Yotov et al., 2016):

$$X_{ij} = \tau_{ij}^{-\theta} \frac{Y_i}{\pi_i^{-\theta}} \frac{E_j}{\rho_j^{-\theta}}; \quad \pi_i^{-\theta} = \sum_{j=1}^N \frac{E_j \tau_{ij}^{-\theta}}{\rho_j^{-\theta}}; \quad \rho_j^{-\theta} = \sum_{i=1}^N \frac{Y_i \tau_{ij}^{-\theta}}{\pi_i^{-\theta}}$$

- In the above partial equilibrium estimates, we considered Y_i , E_j , $\pi_i^{-\theta}$, and $\rho_j^{-\theta}$ being constant when $\tau_{ij}^{-\theta}$ got affected by PTA'_{ijt}
- So, it assumes that enforcing PTA only affects the partnered countries;
 - No impact on the other countries
- But $\pi_i^{-\theta}$ and $\rho_j^{-\theta}$ can be defined by E_j , Y_i , and $\tau_{ij}^{-\theta}$;
 - Accounts for changes in all other trading partners

General Equilibrium Analysis

✓ By solving the structural system (Baier et al., 2019):

$$\hat{X}_{ij} = \frac{\hat{w}_i^{-\theta} \exp(\beta' PT A'_{ij})}{\hat{P}_j^{-\theta}} \hat{E}_j;$$
$$\hat{W}_i = \frac{\hat{E}_i}{\hat{P}_i}$$

- \widehat{w} and \widehat{P} are predicted wage and price levels decided endogenously in the system.
- I iterate my calculation over the years and provide average annual effects of PTA enforcement in agricultural welfare from 1990 to 2017.

Results: Annual Welfare Gain/Loss

	No PTA	PTA with no Relevant Provisions	PTA with Fewer Relevant Provisions	PTA with More Relevant Provisions
Worldwide	-0.13%	+0.62%	+0.62%	+1.41%
North	-0.04%	+0.90%	+1.11%	+2.27%
South	-0.15%	+0.41%	+0.37%	+0.46%

Note: This table shows the annual gain/loss ($\%\Delta \widehat{W}_i$) of agricultural welfare.

Results: Top 10 Agricultural Exporters

Country	PTA (Count)	Effective PTA (Count)	Welfare Gain/Loss (%)	⊿Exports (%)
USA	20	0	+0.65	-1.33
BRA	43	0	+0.04	-1.71
NLD	94	50	+2.21	+1.86
CAN	14	1	+0.09	-2.53
CHN	24	1	+1.59	-0.77
AUS	27	0	-1.22	-0.21
ESP	90	50	+5.25	-0.03
IND	52	0	-5.09	-0.24
DEU	98	50	+9.33	+0.16
FRA	96	50	+10.84	+0.22

Note: This simulation is based on 2016.

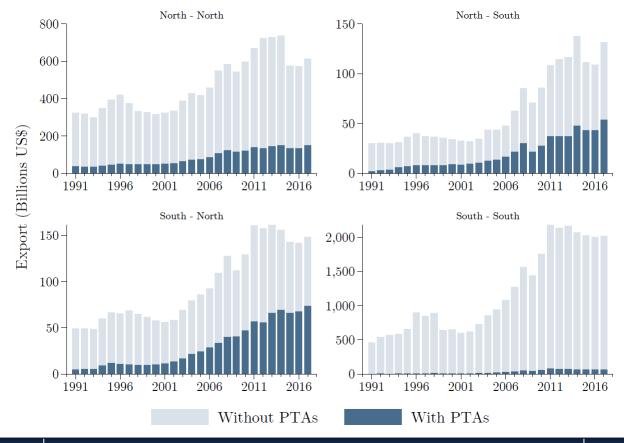
Summary

- ✓ PTA's heterogeneous trade effects caused by non-tariff provisions
 - 6 most relevant provisions with policy areas: Anti-discriminatory (Export Taxes, Subsidy, State Enterprises), SPS and TBT measures, and Trade Facilitations
 - PTAs without these provisions have no impact on agricultural trade
- ✓ The welfare gain by PTA enforcement is more significant
 - If PTA contains more relevant provisions
 - For the developed countries
 - For the countries have more PTA partners

Next Step

- ✓ The model cannot fully control endogeneity of policy
 - Adding reliable confounding variables
- ✓ Tariff impacts
 - Different time-lags (tariff reduction effects)
 - Sub-industry level analysis with average tariff
- ✓ Prediction
 - Full endowment model
 - More scenarios -> policy implications

Supplementary: Bilateral Agricultural Trade between PTA Partners



Supplementary: Selected PTA Provisions and Bundles

ENV.19	ET.06	SPS.44	STE.39	SUB.16	TBT.06/33	TF.32
-4.7%	19.2%	8.2%	17.6%	20.8%	21.8%	57.0%
ENV.06 (0.54)	CVD.01 (0.62)	ROR.06 (0.81)	CP.07 (0.62)	SUB.26 (0.65)	TBT.16 (0.66)	TBT.18 (0.67)
	ET.03 (0.55)	SPS.56 (0.62)	CP.15 (0.60)	SUB.34 (0.75)	TBT.21 (0.58)	TF.20 (0.60)
	ET.15 (0.61)		STE.27 (0.72)	TBT.22 (0.65)		TF.26 (0.67)
	STE.32 (0.58)		STE.28 (0.65)	TF.11 (0.59)		TF.29 (0.93)
	SUB.14 (0.53)		STE.46 (0.62)	TF.38 (0.60)		
			SUB.03 (0.67)			
			SUB.29 (0.88)			

Supplementary: Including/excluding Intra-national Trade

	Including Do	omestic Sales	Excluding Do	omestic Sales
PTA - Presence of PTA	0.240*** (0.036)	0.022 (0.028)	0.035 (0.031)	-0.020 (0.066)
ENV.19 - Prevent Pollution by Ships		$-0.040 \\ (0.045)$		$0.018 \\ (0.090)$
ET.06 - Require Scheduling of EQ/QR		0.174*** (0.062)		0.099 (0.126)
SPS.44 - Include SPS Chapter		0.096*** (0.027)		$0.101 \\ (0.068)$
STE.39 - Include Sector-specific Discipline on STE		0.175*** (0.060)		0.321*** (0.119)
SUB.16 - Require for Prior Notification on Subsidies		0.191*** (0.030)		$-0.107^* \ (0.058)$
TBT.06/33 - Use Regional Standards		0.191** (0.096)		0.153 (0.122)
TF.32 - Mutual Recognition of AOs		0.448** (0.058)		0.312*** (0.130)
Observation	798,239	786,821	790,978	779,560
Pseudo R^2	0.994	0.994	0.963	0.964

Note: Heteroskedasticity-robust standard errors are in the parentheses, clustered by exporter-importer pair. For the estimated parameters, the semi-elasticity is given by $100*(exp(\beta) - 1)\%$.

Supplementary: Different Time-lags

	Simple Dummy	w/ Provisions
PTAt-1	0.094*** (0.036)	-0.028 (0.043)
PTAt-3	0.036 (0.036)	$0.040 \\ (0.025)$
PTAt-5	0.251*** (0.029)	-0.005 (0.030)
ENV.19 - Prevent Pollution by Ships		0.031 (0.065)
ET.06 - Require Scheduling of EQ/QR		0.069 (0.131)
SPS.44 - Include SPS Chapter		0.095* (0.052)
STE.39 - Include Sector-specific Discipline on STE		0.330*** (0.116)
SUB.16 - Require for Prior Notification on Subsidies		$-0.150^{**} \ (0.058)$
TBT.06/33 - Use Regional Standards		0.165 (0.102)
TF.32 - Mutual Recognition of AOs		0.202 (0.157)
Observation	672,041	657,870
Pseudo R ²	0.995	0.966

Note: Heteroskedasticity-robust standard errors are in the parentheses, clustered by exporterimporter pair. For the estimated parameters, the semi-elasticity is given by 100*(exp(β) - 1)%.

Supplementary: Variable Selection Method in Detail

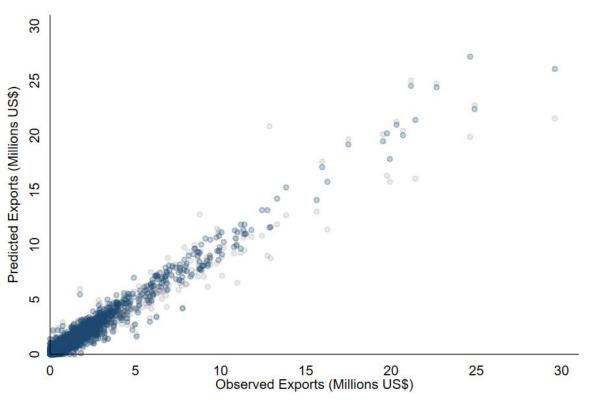
✓ PPML-Lasso regression model (Breinlich et al., 2021):

$$(\widehat{\boldsymbol{\beta}}, \widehat{\alpha}, \widehat{\gamma}, \widehat{\delta}) \coloneqq \arg\min_{\beta, \alpha, \gamma, \delta} \quad \frac{1}{N} \sum_{i,j,t} (\mu_{ijt} - X_{ijt} \ln \mu_{ijt}) + \frac{1}{N} \sum_{l=1}^{M} \widehat{\phi}_{l} \lambda |\beta_{l}|$$

$$\frac{1}{N} \sum_{i,j,t} (\mu_{ijt} - X_{ijt} \ln \mu_{ijt}) + \frac{1}{N} \sum_{l=1}^{M} \widehat{\phi}_{l} \lambda |\beta_{l}|$$
Penalty Term

- $\mu_{i,it}$ is the conditional mean, defined in the previous slide
- *n* is the number of observations (relies on countries and year)
- $\lambda \geq 0$ is the standard tuning parameter for any regularization regression
- $\hat{\phi}_l \ge 0$ is the regressor-specific penalty term suggested by Belloni et al. (2016)
- Used the plug-in algorithms to find λ and $\widehat{\phi}_l$

Supplementary: Predicted vs. Observed Trade



Note: The gray dots represent the model fit of the simple dummy approach. The blue dots represent the model fit of including the selected provisions.