



The Future of Agricultural Trade Policy Modeling: Insights from Meta-Analysis

Sebastian Hess & Stephan von Cramon-Taubadel
Georg-August University, Göttingen

1. Background and motivation

“Abolishing all trade barriers could boost global income by \$US2.8 trillion and lift 320 million people out of poverty by 2015.”

M. Moore, Director-General, WTO 2002

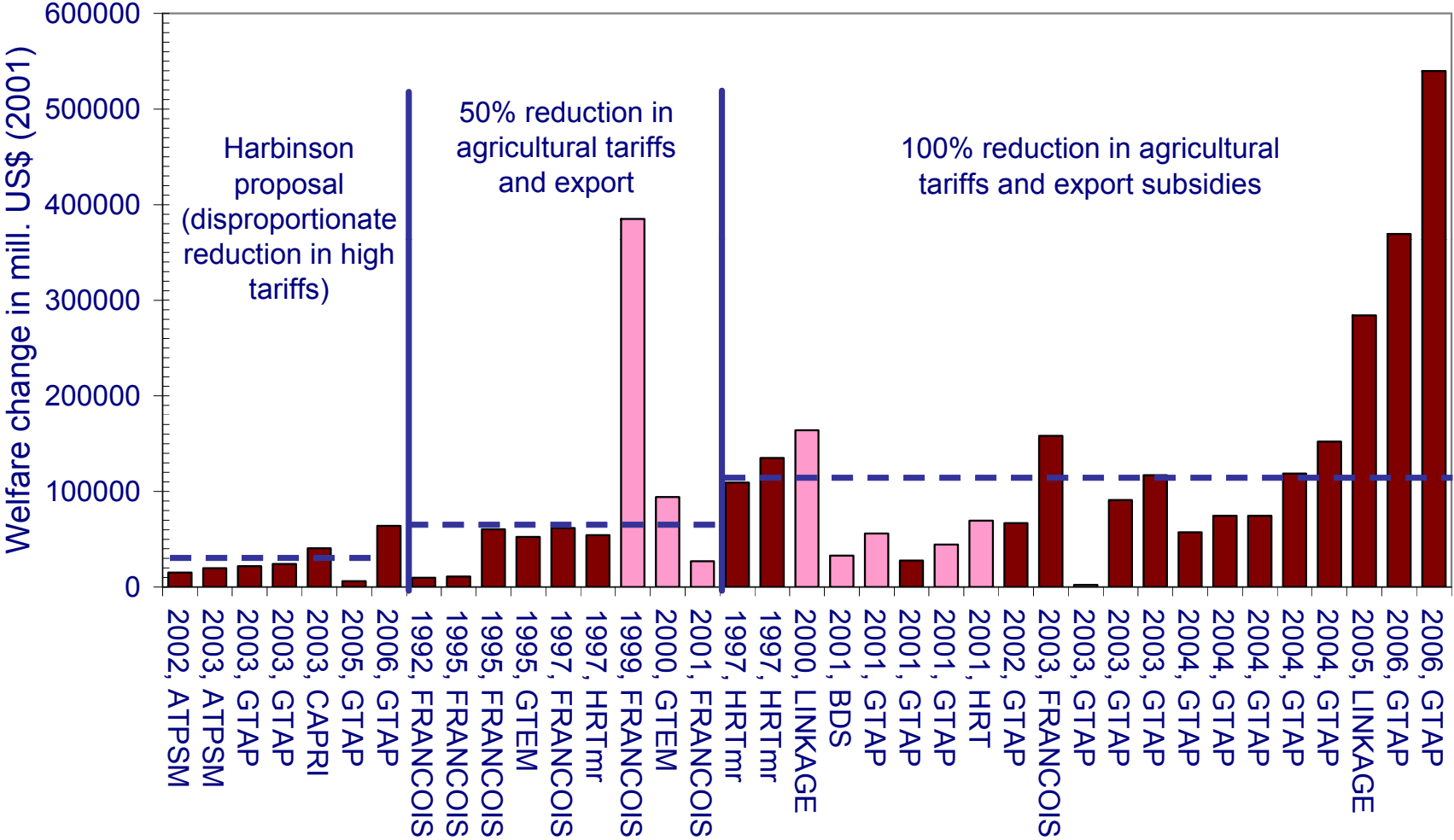
“I am confident that I can concoct a model to generate any result desired by a reader with a deep pocketbook.”

J. Markusen

1. Background and motivation

- Quantitative models are arguably the single most visible input into trade policy debates that economists provide
- However, quantitative models are frequently criticised:
 - Weak empirical and theoretical foundations
 - Complexity, ‘black-box’ character
 - Results vary widely even across similar experiments

1. Background and motivation



1. Background and motivation

This:

- is confusing for policy makers
- is water on the mills of those who question liberalisation
- reflects poorly on our credibility as a profession

Goal: To identify characteristics of simulations that systematically influence results, and to derive quantitative estimates of these influences

Outline

1. Background and motivation
2. What drives model simulation results?
3. Meta analysis
 - with a literature sample
 - with synthetic data
4. Conclusions

2. What drives simulation results?

$$I = f\{ MC, LE, DB, RC \}$$

where:

I = simulated impact of the liberalisation experiment

MC = a vector of model characteristics

LE = the nature of the liberalisation experiment

DB = the database employed

RC = the research context

2. What drives simulation results?

Modellers are, of course, aware of these issues:

1. Sensitivity analysis
 - Usually uni- or bi-variate
 - Propensity to document ‘robustness’

2. Qualitative review
 - Subjective selection of studies for analysis
 - Again, not multivariate

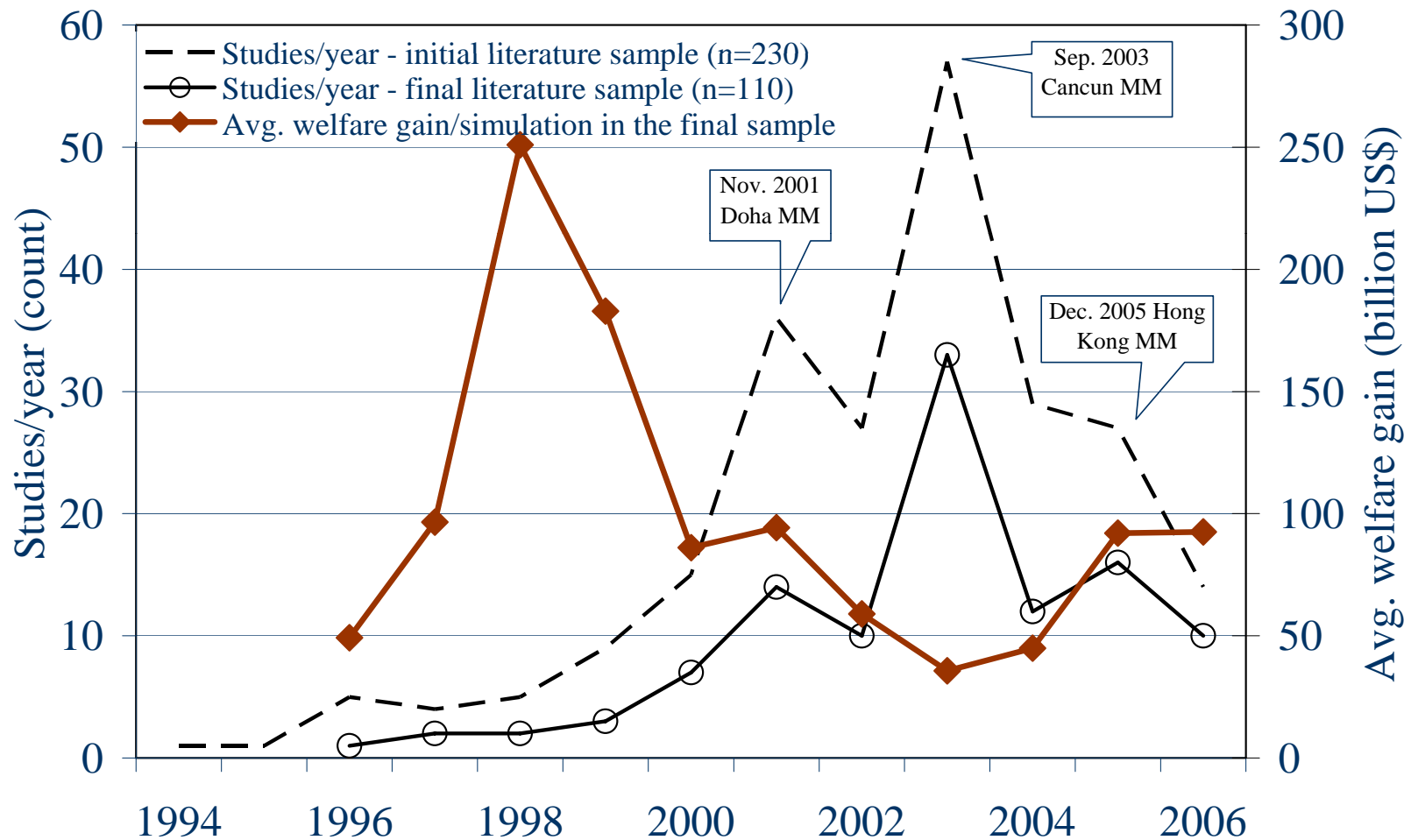
3. Meta analysis

- Main uses:
 - Combining evidence to obtain more precise estimates of an effect size (e.g. Medicine, but also economics)
 - Evaluating methods: quantifying the share of variance in a given set of estimates that is due to differences in methods, assumptions and other factors
- $I = f\{MC, LE, DB, \cancel{RC}, u\}$

3. Meta-analysis: Literature sample

- Exhaustive literature search: 160 combinations of 10 'Doha' and 16 'modelling' keywords, 18 databases
- Result → 1200 studies
- Eliminating:
 - repetition and redundancies
 - studies that do not report welfare changes
 - studies that provide no information on the model used
→ 230 studies
- Eliminating:
 - studies that do not focus on Doha Round liberalisation
 - studies that do not document key model characteristics
→ 110 studies

3. Meta-analysis: Literature sample



3. Meta-analysis: Literature sample

- Each of the 110 studies produces an average of 53 individual observations $\rightarrow n = 5835$
- I = simulated Δ welfare in million US\$ (dummies account for different measures)
- MC = a vector of 23 model characteristics (e.g. CRTS vs. IRTS, size of Armingtons, etc.)
- LE = a vector of 5 measures of the size of the simulated liberalisation (e.g. tariff reduction, reduction in NTBs, shift to blue box spending)
- DB = a vector of 4 dummies (GTAP-4, GTAP-5, etc.)

3. Meta-analysis: Literature sample

- This set of explanatory variables is the result of a specification search using:
 - stepwise regression
 - regression trees
 - an earlier sample of 53 studies
 - roughly 150 potential explanatory variables
- Unweighted and weighted OLS used to account for different number of observations per study

3. Meta-analysis: Literature sample

Results (1)

- Standards of documentation and transparency in the modelling literature are low
- Not seldom that the authors of a simulation are unable, a few years later, to reconstruct exactly how it was produced
- Modelling community small:
 - 288 authors contribute to the 230 studies
 - 25 most frequent authors involved in 176 (76%) of these studies

3. Meta-analysis: Literature sample

Results (2)

- R^2 in unweighted regression 24%, in weighted 46%
- Signs and significance of coefficients robust
- Most coefficients have expected signs and plausible magnitudes, e.g.:
 - Larger liberalisation steps → larger welfare gains
 - Higher Armington elasticities → larger welfare gains
 - Shocks to technical change → larger welfare gains

3. Meta-analysis: Literature sample

Dummies for model interactions				Coeff.	Std.err.	Signif.
Dynamic	CRTS/ perfect comp.	Cap. fixed	Arm's high	12342	939	***
		Cap. accum.	Arm's low	2106	723	***
			Arm's high	12290	707	***
	IRTS/ imperfect comp.	Cap. accum.	Arm's low	451	1102	
			Arm's high	14483	1863	***
	Comparative static	CRTS/ perfect comp.	Cap. accum.	Arm's low	4518	278
Arm's high				4708	830	***
IRTS/ imperfect comp.		Cap. fixed	Arm's high	20746	4766	***
		Cap. accum.	Arm's low	5717	555	***
			Arm's high	6090	2586	**

3. Meta-analysis: Literature sample

Problems

- Many important variables can only be captured as dummies
- Functional form, interaction terms
- Many studies cannot be included due to incomplete documentation
- DDA studies focus on a limited sub-set of the relevant variable space (e.g. mostly tariff reductions)

Therefore: Synthetic data

3. Meta-analysis: Synthetic data

PE Model “GSIM”

(Francois&Hall 2003)

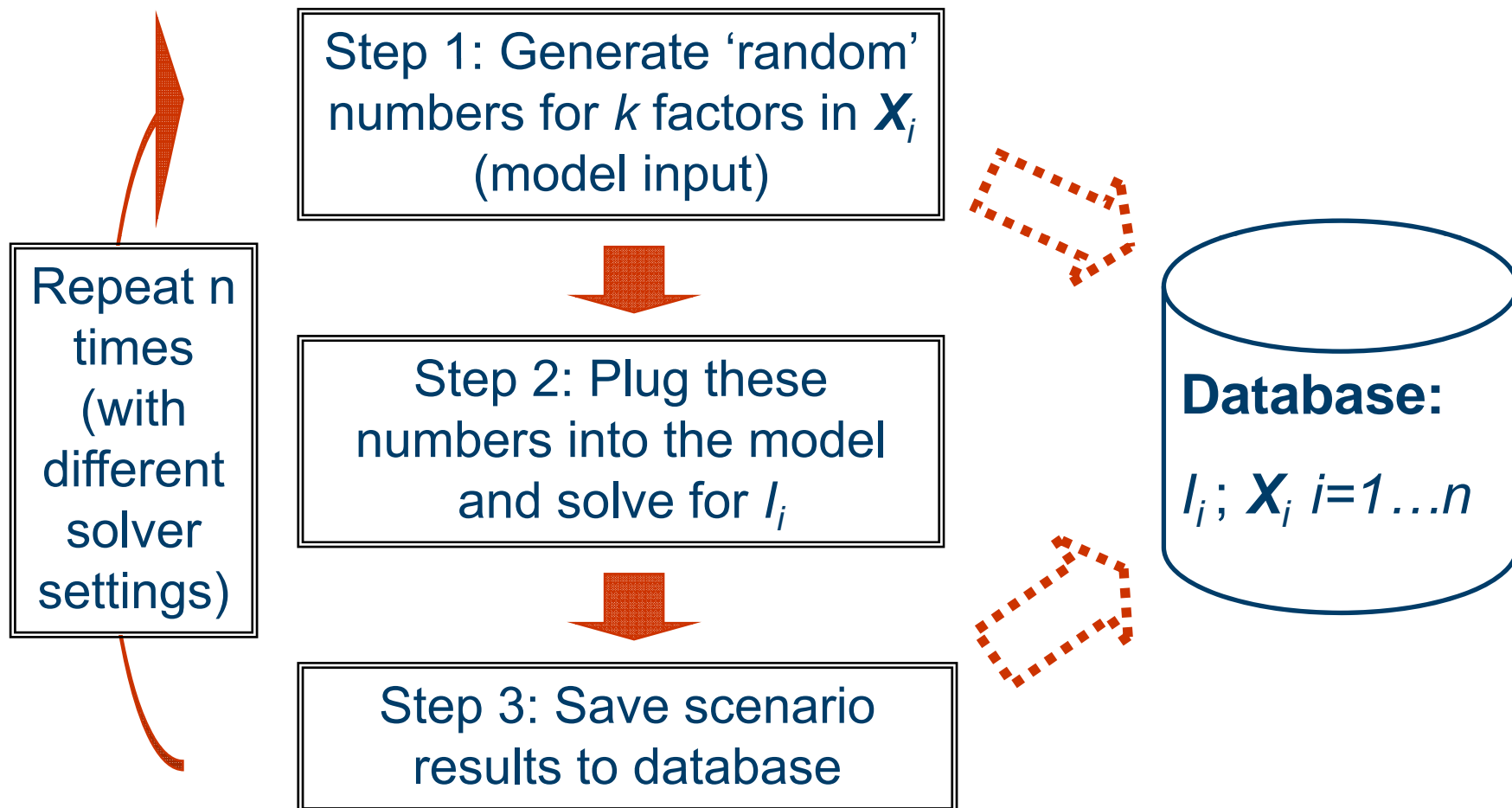
- Wheat
- Canada, US, EU, ROW
- Data = GTAP-5
- Armington-style import differentiation
- Bilateral trade flows and tariffs
- Elasticities: demand, supply, import substitution

Single Country CGE

(van der Mensbrugghe, 2000)

- Agriculture & All other
- CAN & ROW
- SAM=GTAP-5
- Armington import substitution
- Import tariffs, export subsidies, MFN
- CES production, CET exp./dom., ELES

3. Meta-analysis: Synthetic data



3. Meta-analysis: Synthetic data

- Generalized Additive Model (GAM)

- $$g(m) = \underbrace{\beta \mathbf{X}_{ji}}_{\text{parametric}} + \underbrace{f_1(\mathbf{X}_{mi}) + f_2(\mathbf{X}_{li}, \mathbf{X}_{li})}_{\text{smooth functions, non parametric}} + \varepsilon$$

parametric smooth functions, non parametric

- $g(m) = E\{I_j\}$, $\varepsilon \sim N(0, \sigma^2)$, model selection according to R^2

⇒ Detect functional forms, interactions of \mathbf{X}

⇒ Interpretation: Linear framework

3. Meta-analysis: Synthetic data

Results (1)

GAM order Model	Linear (X+X)	GAM smooth	Linear Interaction (X*X)	Linear (X+X), (X*X), (X ²)
PE GSIM $\Delta(CS+PS)$	0.77	0.78	0.87	0.87
CGE $\Delta real\ GDP$	0.49	0.55	0.80	0.70

(Recall: R^2 unweighted literature sample $\approx 24\%$)

3. Meta-analysis: Synthetic data

Results (2): SAM

	Agriculture	Others	...
Agriculture	7924 (-0.031)	23188 (0.015)	...
Others	18736 (-0.007)	464809 (0.001)	...
Labour	8424 (-0.024)	312781 (0.004)	...
Capital	6977 (-0.038)	213616 (0.003)	...
Income Tax	-2105 (0.101)	12695 (-0.000)	...
...

3. Meta-analysis: Synthetic data

Comparing PE and GE models

1. Separate equations:

$$I_i^{GE} = \beta^{GE} X_i + \phi Z_i + u_i$$

$$I_i^{PE} = \beta^{PE} X_i + \gamma W_i + v_i$$

2. Single equation, fixed effects:

$$I_i = \beta X_i + \alpha \begin{pmatrix} D=0, GE \\ D=1, PE \end{pmatrix} + u_i$$

3. Single equation, imputed missing values, fixed effects:

$$I_i = \beta X_i + \phi \begin{pmatrix} Z_i, GE \\ \bar{Z}_i, PE \end{pmatrix} + \gamma \begin{pmatrix} \bar{W}_i, GE \\ W_i, PE \end{pmatrix} + \alpha \begin{pmatrix} D=0, GE \\ D=1, PE \end{pmatrix} + u_i$$

3. Meta-analysis: Synthetic data

$$I_i = \beta X_i + \phi \begin{pmatrix} Z_i, GE \\ \bar{Z}_i, PE \end{pmatrix} + \gamma \begin{pmatrix} \bar{W}_i, GE \\ W_i, PE \end{pmatrix} + \alpha \begin{pmatrix} D = 0, GE \\ D = 1, PE \end{pmatrix} + u_i$$

- $I_i = \Delta CS$ for PE model; = $\Delta Utility$ for GE model
- $R^2 = 0.77$
- Estimate of $\alpha = 4.76$ bill. US\$ (Prob $>|t| = 0.000$)

4. Conclusions

- Meta-analysis provides important insights beyond conventional sensitivity analysis and qualitative comparisons of model results
- Results with synthetic data confirm basic insights of literature data, but R^2 much higher
- Importance of interactions between model characteristics and liberalisation experiments revealed by synthetic data
- Armington elasticities – weak empirical and theoretical basis, powerful impact on results



4. Conclusions

- First stab at assessing the impact of variations in base data (SAM)
- Indications that solvers can add considerable variability to simulation results
- Important: Our results do not point to the 'true' model
- Simulation results must be interpreted with great caution
- Document, document, document!

Thank you!



The Economist

Literature search

- Exhaustive literature search: 160 combinations of 10 'Doha' and 16 'modelling' keywords, 18 on-line databases

Databases employed in the literature search

AgEconSearch	COPAC	Econis
EconLit	Google/ Google Scholar	IBZ
JADE	JSTOR	KVK
OCLC Journal Articles ECO	OCLC Papers	OCLC Proceedings
OCLC World Cat	OLC	RePec Articles
RePec Working Papers	SSCI	WTO House Library

Literature search

Modelling keywords

applied model	CGE	comput*	dynamic
economy wide	equilibrium	forward* model	general model
global model	international model	mathemat*	multi commodity
multi region*	multi sector*	partial model	static

Doha keywords

doha agric* market*	doha develop*
doha domestic support	export subsid*
doha fair trad*	doha food security
doha market access	doha market* trad*
doha non trade	doha trade distort*

Fixed effects (bill. US\$)

Country/region	Fixed effect (std. error)
ACP (intercept)	-11.6 (1.5)
Australia	5.0 (1.2)
Canada	7.3 (0.9)
India	8.3 (0.8)
EU-25	14.9 (2.5)
USA	16.9 (2.2)