

CGE Models in GAMS: History and Current State of the Art

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History: Solving CGE Models

- Various approaches
 - Programming models
 - LP: Chenery, Evans, Norton-Scandizzo
 - NLP: Ginsburgh-Waelbroeck, Ginsburgh-Kayzer
 - Systems of equations: “square models”
 - Scarf, Shoven-Whalley
 - Johansen, Dixon (Orani), Hertel (GTAP)
 - Adelman-Robinson, Dervis-deMelo-Robinson

History: Solving CGE Models

- The Scarf algorithm died out quickly.
 - Empirically very inefficient and impractical.
- The programming approaches are still around, but little used.
- Direct solution approaches have prevailed. Basically, two schools:
 - ORANI: CGE model in log differentials.
 - Direct methods: CGE model in “levels”.

Solving CGE Models in Levels

- Tatonnement methods
 - Gauss-Seidel/Jacobi algorithms
- Jacobian methods
 - Newton-Rapheson
 - M.J.D. Powell: Powell algorithm
- Fortran programs
 - 1970s and early 1980s

CGE Models in GAMS

- Mid 1980s: convert to GAMS
- Modeling “systems”
 - Hercules (Pyatt-Drud)
 - MPSGE (Rutherford)
- “Cameroun” model
 - Devarajan; Robinson (USDA/ERS)
 - Devarajan, Lewis, Robinson: a “standard” CGE model

CGE Models in GAMS

- IFPRI “standard” CGE model
 - Lofgren, Harris, and Robinson
- SAM-based model
 - New UN-SNA SAM
 - CGE model consistent with new data framework

IFPRI CGE Model in GAMS

- The paper is accompanied by a modeling system: a set of GAMS input files (model, data, simulations, reports)
- The system includes extensive reports.
- The system has been applied to a large number of countries.

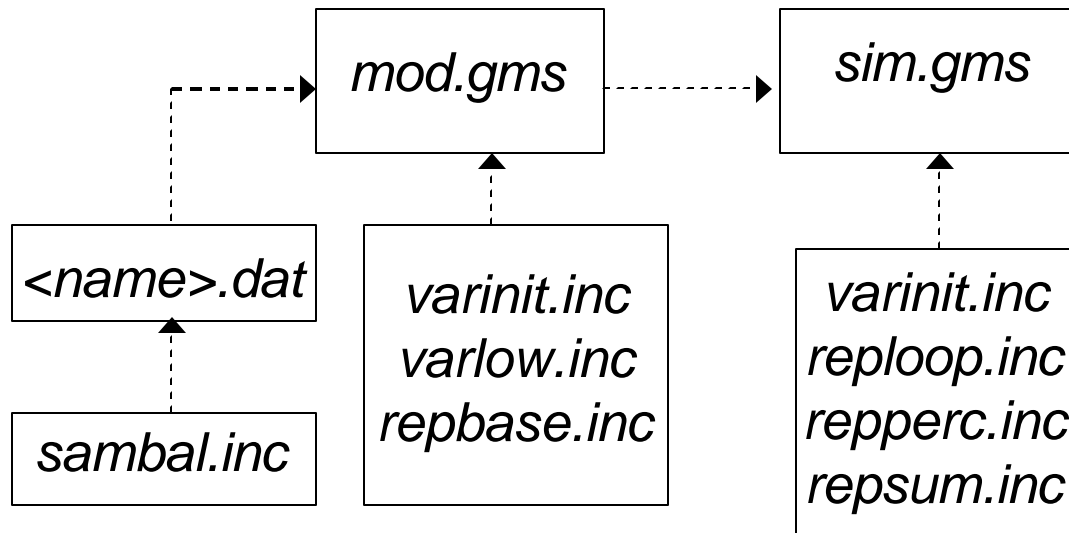
Application Countries

- Argentina
- Bangladesh
- Bolivia
- Botswana
- Brazil
- Chile
- China
- Colombia
- Costa Rica
- Dominican Republic
- Ecuador
- Egypt
- Honduras
- Indonesia
- Jamaica
- Madagascar
- Malawi
- Mexico

Application Countries

- Morocco
- Mozambique
- North Korea
- Paraguay
- Philippines
- Peru
- Russia
- South Africa
- Swaziland
- Tanzania
- Thailand
- Turkey
- Uganda
- United States
- Uruguay,
- Venezuela
- Vietnam
- Zambia
- Zimbabwe

GAMS Modeling System



The Model in GAMS

- This system may be used in the following ways:
 1. Simulations with existing data set without changing the modeling structure (*sim.gms*).
 2. Simulations with new data set:
 - a. construct data set (*<name>.dat*)
 - b. simulations (*sim.gms*)
 3. Simulations with new data set and model structure:
 - a. construct data set (*<name>.dat*)
 - b. adjust model structure (*mod.gms*) and selected include files (*? .inc*)
 - c. simulations (*sim.gms*)

GAMS Solvers for CGE Models

- NLP Solvers
 - Minos: Stanford
 - Conopt: Drud
 - Used to solve “square” CGE models or models expressed as NLP problems
- MCP: “Square” solvers
 - Miles: Rutherford
 - Path: Ferris, Munson, Dirkse

Algorithms and Models

- Interaction between developers of CGE models, developers of solution algorithms, and GAMS language features.
- Productive interaction.
 - Special properties of CGE models.
 - Test problems.

CGE Model

- (1) $QS(a) = \sum_a aq(a) * PROD(f, QF(f, a) * betaq(f, a)) ;$
- (2) $WF(f) * QF(f, a) = \sum_a betaq(f, a) * PQ(a) * QS(a) ;$
- (3) $YH = \sum_f WF(f) * qfs(f) ;$
- (4) $PQ(a) * QD(a) = \sum_a betah(a) * YH ;$
- (5) $QD(a) - QS(a) = \sum_a 0 ;$
- (6) $SUM(f, a, QF(f, a)) - qfs(f) = \sum_f 0 ;$
- (7) $SUM(a, alpha(a) * PQ(a)) = \sum_a pbar ;$

Identities:

$$YH \equiv \sum_a PQ(a) * QD(a)$$

$$PQ(a) * QS(a) \equiv \sum_f WF(f) * QF(f, a)$$

$$\sum_a PQ(a) * (QD(a) - QS(a)) +$$

$$\sum_f WF(f) * (SUM(a, QF(f, a)) - qfs(f)) \equiv 0$$

Equations and Variables

Equation:

- (1) Product supply
- (2) Factor demand
- (3) Income
- (4) Product demand
- (5) Product market
- (6) Factor market
- (7) Price numeraire

Variables:

- $QS(a)$
 $QF(f, a)$
 YH
 $QD(a)$
 $PQ(a)$
 $WGF(f)$

Square CGE Model

- The model has one more equation than number of variables.
- The third identity (Walras' Law) implies that the excess demand equations are functionally dependent.
 - Jacobian is singular.
- Drop one excess demand equation.
 - MCP will solve this model.
 - NLP solvers will also work fine, with arbitrary objective function.

NLP Solvers

- Problem is to find unique feasible basis.
- Objective function is irrelevant, but judicious choice of objective function may improve solver performance.
 - Sum of squared excess demands.
- Feasible to loosen model and optimize.
 - Social welfare function.

Benign Singularity

- Add an identity. One more equation than number of variables.
 - Or do not drop an excess demand equation.
- The solution is unchanged—benign singularity.
- NLP solvers will solve this model.
 - Possible scaling/precision problems, but new solvers seem unfazed.
- MCP solver will fail—model is not square.

Benign Rectangularity

- Some factors do not appear in some sectors. They are excluded from production functions and factor demand equations.
- But, perhaps not excluded from factor market clearing equations.
- Result: model with excess variables.

Benign Rectangularity

$$(1) \text{QS}(a) \stackrel{E}{=} aq(a) * \text{PROD}(f \text{\$beta}q(f, a), \\ \text{QF}(f, a) ** \text{beta}q(f, a)) ;$$

$$(2) \text{\$beta}q(f, a) .. \\ \text{WF}(f) * \text{QF}(f, a) \stackrel{E}{=} \text{beta}q(f, a) * \text{PQ}(a) * \text{QS}(a) ;$$

$$(6) \text{SUM}((f, a), \text{QF}(f, a)) - \text{qfs}(f) \stackrel{E}{=} 0 ;$$

Benign Rectangularity

- Excess variables must all be zero at solution. So, solution does not change.
- NLP solvers will solve this model.
- MCP will fail—model is not square.

MCP Test

- Add one identity and one benign excess variable.
 - Benign singularity.
- Model is square, but Jacobian is singular.
- NLP solvers will solve this model.
- MCP will also solve it.

Walras' Law

- Walras' Law implies a singular Jacobian of excess demand equations.
 - Standard approach: drop one equation.
- Alternative: Add a dummy variable
 - “Walras” variable must be zero at solution.
 - Model is square.
- Jacobian is singular at solution.
- NLP and MCP solvers all work.

Conclusion

- MCP solvers
 - Facilitate development of CGE models incorporating complementarity relationships
 - Minimum wage
 - TRQs and import rationing
 - LP technology specification
 - Agricultural policy regimes
 - “Regime shift” macro models

Conclusion

- NLP solvers
 - Dynamic optimization: optimal growth models.
 - Rational expectations models.
 - Macro models.
 - Welfare analysis: welfare maximization with policy instruments.