Introduction and Solving Equations

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Spring 2025

Model Meets Data: A "Typical" JMP Today

- Motivational facts: may include reduced-form results
- Model:
 - Setup and equilibrium
 - Theoretical results: key mechanisms
- Bring model to data
 - Estimate key parameters
 - Calibrate other parameters
- Counterfactuals
 - Quantify existing or proposed policies/shocks
 - Optimal policies

Why Guide Empirics by Theories?

- Lucas critique:
 - It is naive to try to predict the effects of a change in economic policy entirely on the basis of relationships observed in historical data
 - Policy evaluation should base on deep/structural parameters
 - How deep? Depend on your questions and theoretical interpretations
 - Tariffs ⇒ Employment: Firm productivities can be regarded as "deep"
 - Innovation subsidies ⇒ Employment: Firm productivities are not "deep" (an outcome of innovation investments)
- Example: Gravity equation

Gravity Equation

- The "naive" gravity equation proposed by Tinbergen (1962)

$$\log X_{in} = -\epsilon \log (1 + t_{in}) + D'_{in}\gamma + \alpha X_i + \beta X_n + u_{in}, \tag{1}$$

where X_{in} is the total export from country i to n, t_{in} is the tariff rates of country n on imports from country i, D_{in} is a vector of distance measures such as physical distance, common border, common language, and so on. X_i is the total expenditure of country i

- The fixed-effect gravity equation:

$$\log X_{in} = -\epsilon \log (1 + t_{in}) + D'_{in}\gamma + fe_i + fe_n + u_{in}, \qquad (2)$$

where the fixed effects fe_i and fe_n do not only absorb the total expenditure but also all factors that are exporter- or importer-specific

- Tariff ⇒ Trade flows
 - What does ϵ measure?
 - Tariff \Rightarrow Wage \Rightarrow Expenditure \Rightarrow Trade flows?

Structural gravity equation: Armington Model

 N countries. Each country i produces a distinctive variety of goods. Consumers in country n has a CES preference over all varieties:

$$U_{n} = \left[\sum_{i=1}^{N} C_{in}^{\frac{\sigma-1}{\sigma}}\right]^{\frac{\sigma}{\sigma-1}}, \quad \sigma > 1,$$
(3)

where C_{in} is the quantity of goods from country i consumed in country n

- Country i is endowed with L_i workers. The variety i is produced using labor under perfect competition. The unit cost of producing good i is

$$c_i = \frac{w_i}{A_i},\tag{4}$$

where w_i is the wage and A_i is the productivity in country i

- Exporting from country *i* to *n* incurs
 - an iceberg trade cost, $au_{in} \geq 1$, with $au_{ii} = 1$
 - an import tariff, $t_{in} \geq 0$, with $t_{ii} = 0$

Armington Model: Equilibrium

- Let X_n be the total expenditure in country n and X_{in} be the value of exports from i to n. Then bilateral trade share can be expressed as

$$\lambda_{in} \equiv \frac{X_{in}}{X_n} = \left[\frac{w_i \kappa_{in}}{A_i}\right]^{1-\sigma} P_n^{\sigma-1}, \quad \kappa_{in} \equiv \tau_{in} (1 + t_{in}), \tag{5}$$

where
$$P_n = \left[\sum_{i=1}^N \left[\frac{w_i \kappa_{in}}{A_i}\right]^{1-\sigma}\right]^{\frac{1}{1-\sigma}}$$

Labor market clearing:

$$w_i L_i = \sum_{n=1}^N \frac{1}{1 + t_{in}} \lambda_{in} X_n. \tag{6}$$

- Total expenditure equates total income:

$$X_n = w_n L_n + \sum_{i=1}^N \frac{t_{in}}{1 + t_{in}} \lambda_{in} X_n. \tag{7}$$

- Welfare measure: real income

$$U_i = \frac{X_i}{P_i}. (8)$$

Armington Model: Gravity

- Equation (5) can be re-written into

$$\log\left(X_{in}\right) = -\left(\sigma - 1\right)\log\left(1 + t_{in}\right) - \left(\sigma - 1\right)\log\tau_{in} + \left(\sigma - 1\right)\log\left(\frac{w_i}{A_i}\right) + \log\left(P_n^{\sigma - 1}X_n\right) \tag{9}$$

- Suppose that $\tau_{in} = D'_{in}\tilde{\gamma} + \tilde{u}_{in}$. Then we have the fixed-effect gravity equation:

$$\log X_{in} = -(\sigma - 1)\log(1 + t_{in}) + D'_{in}\gamma + fe_i + fe_n + u_{in}.$$
 (10)

- The coefficient of $\log{(1+t_{in})}$ has a structural interpretation: $-(\sigma-1)$

Armington Model: Counterfactuals

- How do changes in (t_{in}) affect trade share λ_{in} , considering direct and indirect effects?
- Parameters: $(A_i, L_i, \tau_{in}, t_{in}; \sigma)$
- Equilibrium outcomes (w_i, X_i) such that

1

$$w_{i}L_{i} = \sum_{n=1}^{N} \frac{1}{1+t_{in}} \lambda_{in} X_{n}, \quad \lambda_{in} = \frac{\left(\frac{w_{i}\kappa_{in}}{A_{i}}\right)^{1-\sigma}}{\sum_{k=1}^{N} \left(\frac{w_{k}\kappa_{kn}}{A_{k}}\right)^{1-\sigma}}, \quad \kappa_{in} = \tau_{in} \left(1+t_{in}\right). \tag{11}$$

2

$$X_n = w_n L_n + \sum_{i=1}^N \frac{t_{in}}{1 + t_{in}} \lambda_{in} X_n. \tag{12}$$

- Counterfactual:
 - Solve (w_i, X_i) and (λ_{in}) under baseline (t_{in}) .
 - Do it again under alternative (t_{in}) .

Reduced-Form vs. Structural Trade Elasticities

- Reduced-form trade elasticity:
 - Fixed-effect gravity equation:

$$\log \lambda_{in} = -(\sigma - 1)\log(1 + t_{in}) + D'_{in}\gamma + fe_i + fe_n + u_{in}$$
(13)

- Ceteris paribus: capture, by design, the direct effect
- Structural trade elasticity:
 - $Structural\ interpretation$ of reduced-form trade elasticity: Elasticity of substitution \Rightarrow Deep parameters
 - GE counterfactual elasticity for policy evaluation: $\Delta(1+t_{in}) \Rightarrow_{\text{Overall effects}} \Delta X_{in}$?

Reduced-Form vs. Structural Trade Elasticities

- Log-linearizing the equilibrium system: for any Z > 0, denote $\tilde{Z} = d \log Z$

$$w_{i}L_{i} = \sum_{n=1}^{N} \frac{1}{1 + t_{in}} \lambda_{in} X_{n}, \quad \lambda_{in} = \frac{\left(\frac{w_{i} \kappa_{in}}{A_{i}}\right)^{1 - \sigma}}{\sum_{k=1}^{N} \left(\frac{w_{k} \kappa_{kn}}{A_{k}}\right)^{1 - \sigma}}, \quad \kappa_{in} = \tau_{in} \left(1 + t_{in}\right)$$

$$X_{n} = w_{n}L_{n} + \sum_{i=1}^{N} \frac{t_{in}}{1 + t_{in}} \lambda_{in} X_{n}$$

$$(14)$$

- Linearized system: Deriving $\frac{\partial \log \lambda_{in}}{\partial \log(1+t_n)}$ from a GE model

$$\tilde{w}_{i} + \tilde{L}_{i} = \sum_{n=1}^{N} \chi_{in} \left(\tilde{\lambda}_{in} + \tilde{X}_{n} - 1 + t_{in} \right), \quad \chi_{in} \equiv \frac{\frac{1}{1 + t_{in}} \lambda_{in} X_{n}}{w_{i} L_{i}}$$

$$\tilde{\lambda}_{in} = (1 - \sigma) \left(\tilde{w}_{i} + \tilde{\kappa}_{in} - \tilde{A}_{i} \right) - \sum_{k=1}^{N} \lambda_{kn} \left(\tilde{w}_{k} + \tilde{\kappa}_{kn} - \tilde{A}_{k} \right), \quad \tilde{\kappa}_{in} \equiv \tilde{\tau}_{in} + 1 + t_{in}$$

$$\tilde{X}_{n} = \frac{w_{n} L_{n}}{X_{n}} \left(\tilde{w}_{n} + \tilde{L}_{n} \right) + \sum_{k=1}^{N} \left[\frac{t_{in} \lambda_{in}}{1 + t_{in}} \left(\tilde{\lambda}_{in} + \tilde{X}_{n} \right) + \frac{\lambda_{in}}{1 + t_{in}} \right]$$
(15)

Structural Modeling

- Theoretical model: $F(Y, X, \theta) = 0$: Y: equilibrium outcomes; X: policies; θ : deep parameters
- Model characterization: properties of the equilibrium; elasticities $\nabla_X Y$, given θ
- Empirical model: $G(Y, X, \theta, \varepsilon) = 0$: ε : structural errors to fit the data
- Solve the model: Given θ and ε , solve (X,Y) from $G(Y,X,\theta,\varepsilon)=0$
- Bring the model to the data: $G(Y, X, \theta, \varepsilon) = 0 + \text{Data on } (X, Y) + \text{Assumptions} \Rightarrow \theta$
- Counterfactuals: Given θ , changes in $X \Rightarrow$ changes in Y

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- #5 Economic models can be used to address sample selection problems in regressions
 - e.g. Irarrazabal, Moxnes, and Opromolla (2013)

Trade-offs

- Structural modeling:
 - Answering more questions in interest
 - Useful in policy evaluation (both actual and hypothetical policies)
 - With stronger assumptions (some may not be easily justifiable)
- Structural modeling excels in the topics with:
 - Well-established and uncontroversial theories
 - Rich data that can characterize agents' entire behaviors
 - Relevant to important policy questions

Solving Non-Linear Equations

- General problem:

$$f(\mathbf{x}) = 0, \quad \mathbf{x} \in \mathcal{X} \subseteq \mathbb{R}^N, \quad f(\mathbf{x}) \in \mathbb{R}^N, \forall \mathbf{x} \in \mathcal{X}$$
 (16)

Example from Burden and Faires

$$3x_1 - \cos(x_2 x_3) - \frac{1}{2} = 0$$

$$x_1^2 - 81(x_2 + 0.1)^2 + \sin x_3 + 1.06 = 0$$

$$\exp[-x_1 x_2] + 20x_3 + \frac{10\pi - 3}{3} = 0$$
(17)

Solving Non-Linear Equations

- Jacobian matrix:

$$J(\mathbf{x}) = \begin{bmatrix} \frac{\partial f_1(\mathbf{x})}{\partial x_1} & \frac{\partial f_1(\mathbf{x})}{\partial x_2} & \dots & \frac{\partial f_1(\mathbf{x})}{\partial x_N} \\ \frac{\partial f_2(\mathbf{x})}{\partial x_1} & \frac{\partial f_2(\mathbf{x})}{\partial x_2} & \dots & \frac{\partial f_2(\mathbf{x})}{\partial x_N} \\ \vdots & \vdots & \dots & \vdots \\ \frac{\partial f_N(\mathbf{x})}{\partial x_1} & \frac{\partial f_N(\mathbf{x})}{\partial x_2} & \dots & \frac{\partial f_N(\mathbf{x})}{\partial x_N} \end{bmatrix}$$

$$(18)$$

- Jacobian matrix for the example from Burden and Faires

$$J(\mathbf{x}) = \begin{bmatrix} 3 & x_3 \sin(x_2 x_3) & x_2 \sin(x_2 x_3) \\ 2x_1 & -162(x_2 + 0.1) & \cos x_3 \\ -x_2 \exp[-x_1 x_2] & -x_1 \exp[-x_1 x_2] & 20 \end{bmatrix}$$
(19)

Simple Iteration

- 1. Initial guess $\mathbf{x}^{(0)}$
- 2. Compute $f\left(\mathbf{x}^{(0)}\right)$
- 3. Update $\mathbf{x}^{(1)} = \mathbf{x}^{(0)} f(\mathbf{x}^{(0)})$
- 4. Iterate until $\mathbf{x}^{(k)} = \mathbf{x}^{(k-1)}$ where $k = 1, 2, \dots$

Matlab Programming

- Matrix manipulation
- Function
- Iteration

Newton's Method

- First-order Taylor expansion at $\mathbf{x} = \mathbf{x}^{(0)}$:

$$f(\mathbf{x}) = f(\mathbf{x}_0) + J(\mathbf{x}^{(0)}) \left[\mathbf{x} - \mathbf{x}^{(0)}\right]$$
 (20)

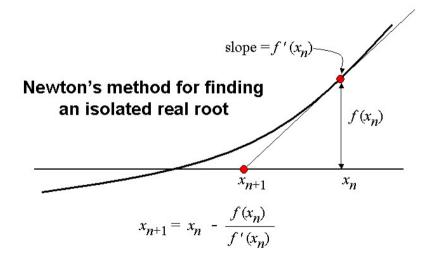
- Since $f(\mathbf{x}) = 0$, we approximate \mathbf{x} by

$$\mathbf{x}^{(1)} = \mathbf{x}^{(0)} - J(\mathbf{x}^{(0)})^{-1} f(\mathbf{x}^{(0)})$$
 (21)

- Newton's iterative method:

$$\mathbf{x}^{(k)} = \mathbf{x}^{(k-1)} - J(\mathbf{x}^{(k-1)})^{-1} f(\mathbf{x}^{(k-1)}), \quad k = 1, 2, ...$$
 (22)

Newton's Method



Newton's Method

- 1. Initial guess $\mathbf{x}^{(0)}$
- 2. Compute $f\left(\mathbf{x}^{(0)}\right)$ and $J\left(\mathbf{x}^{(0)}\right)$
- 3. Update $\mathbf{x}^{(1)} = \mathbf{x}^{(0)} J\left(\mathbf{x}^{(0)}\right)^{-1} f\left(\mathbf{x}^{(0)}\right)$
- 4. Iterate until $\mathbf{x}^{(k)} = \mathbf{x}^{(k-1)}$ where $k = 1, 2, \dots$

Broyden's Method (Quansi-Newton)

- 1. Approximate the inverse of the Jacobian matrix: Initial guess $\mathbf{x}^{(0)}$ and $B_0 = \mathbf{I}$
- 2. Update $\mathbf{x}^{(1)} = \mathbf{x}^{(0)} B_0 f(\mathbf{x}^{(0)})$
- 3. Let $\mathbf{h} = \mathbf{x}^{(1)} \mathbf{x}^{(0)}$ and $\mathbf{y} = f\left(\mathbf{x}^{(1)}\right) f\left(\mathbf{x}^{(0)}\right)$. By the definition of the Jacobian matrix, we want to obtain B_1 such that $B_1\mathbf{y} = \mathbf{h}$
- 4. Update $B_1 = B_0 + \frac{1}{\mathbf{h}'B_0\mathbf{y}}(\mathbf{h} B_0\mathbf{y})\mathbf{h}'B_0$. It is straightforward to verify that $B_1\mathbf{y} = \mathbf{h}$
- 5. Iterate until $\mathbf{x}^{(k)} = \mathbf{x}^{(k-1)}$ where k = 1, 2, ...

Armington Model: Counterfactuals

- Equilibrium outcomes (w_i, X_i) such that

1

$$w_{i}L_{i} = \sum_{n=1}^{N} \frac{1}{1+t_{in}} \lambda_{in} X_{n}, \quad \lambda_{in} = \frac{\left(\frac{w_{i}\kappa_{in}}{A_{i}}\right)^{1-\sigma}}{\sum_{k=1}^{N} \left(\frac{w_{k}\kappa_{kn}}{A_{k}}\right)^{1-\sigma}}, \quad \kappa_{in} = \tau_{in} \left(1+t_{in}\right). \tag{23}$$

2

$$X_{n} = w_{n}L_{n} + \sum_{i=1}^{N} \frac{t_{in}}{1 + t_{in}} \lambda_{in} X_{n}.$$
 (24)

- A toy example:
 - N=3 and $\sigma=4$
 - A = [2, 1, 1], L = [1, 2, 4]
 - $\tau_{in} = 2$ for all $i \neq n$
 - Initially $t_{in}=0$ for all i,n

Bringing the Armington Model to Data

- Challenges:
 - High-dimensional (A_i, L_i, τ_{in})
 - How to get σ ?
- Data:
 - Bilateral trade flows: X_{in} for all i, n
 - Bilateral tariffs: t_{in} for all i, n
 - Bilateral distance measures: D_{in} for all $i \neq n$

Exact-Hat Algebra

- Starting from the observed economy, how would changes in tariffs affect trade flows, wages, and welfare?
 - Observed economy: a set of parameters, e.g. $(A_i, L_i, \tau_{in}) \Leftarrow \mathsf{Data}$, e.g. (X_{in}, t_{in})
 - Shortcut: computing counterfactuals based on Data, without explicitly obtaining parameter values
- Notation: For any variable Z > 0
 - Let Z' be the value of Z after shocks in interest
 - Let $\hat{Z} = Z'/Z$
 - WTH: $(\hat{w}_i, \hat{X}_i, \hat{P}_i)$ under $\widehat{(1+t_{in})}$

Exact-Hat Algebra

- Equilibrium in relative changes: (\hat{w}_i, \hat{X}_i) satisfy

1

$$\hat{w}_{i}\hat{L}_{i}w_{i}L_{i} = \sum_{n=1}^{N} \frac{1}{1 + t_{in}'} \hat{\lambda}_{in}\hat{X}_{n}\lambda_{in}X_{n}, \quad \hat{\lambda}_{in} = \frac{\left(\frac{\hat{w}_{i}\hat{\kappa}_{in}}{\hat{A}_{i}}\right)^{1-\sigma}}{\sum_{k=1}^{N} \lambda_{kn} \left(\frac{\hat{w}_{k}\hat{\kappa}_{kn}}{\hat{A}_{k}}\right)^{1-\sigma}}, \quad \hat{\kappa}_{in} = \hat{\tau}_{in}\widehat{1 + t_{in}}. \quad (25)$$

2

$$\hat{X}_{n}X_{n} = \hat{w}_{n}\hat{L}_{n}w_{n}L_{n} + \sum_{i=1}^{N} \frac{t'_{in}}{1 + t'_{in}}\hat{\lambda}_{in}\hat{X}_{n}\lambda_{in}X_{n}. \tag{26}$$

- Welfare: measured by real income
$$\hat{U}_i \equiv \frac{\hat{\chi}_i}{\hat{P}_i}$$
 where $\hat{P}_i = \left[\sum_{k=1}^N \lambda_{ki} \left(\frac{\hat{w}_k \hat{\kappa}_{ki}}{\hat{A}_k}\right)^{1-\sigma}\right]^{\frac{1}{1-\sigma}}$

- Exogenous shocks:
$$\left(\widehat{1+t_{in}};\widehat{ au}_{in},\widehat{A}_{i},\widehat{L}_{i}\right)$$

Exact-Hat Algebra

- Data: (X_{in}, t_{in}, D_{in})
- Exact-hat algebra requires

-
$$\lambda_{\mathit{in}} \equiv rac{X_{\mathit{in}}}{X_{\mathit{n}}}$$
 where $X_{\mathit{n}} = \sum_{k=1}^{\mathit{N}} X_{\mathit{kn}}$

$$- w_i L_i = \sum_{n=1}^{N} \frac{1}{1+t_{in}} X_{in}$$

- σ : estimated by the gravity equation
- Trade imbalances in the data?
 - Exogenous trade imbalances
 - Eliminate trade imbalances in the data, starting from a balanced world

Summary of Structural Gravity Model

- Question: Tariff changes \Rightarrow Equilibrium outcomes: trade, wages, welfare
- Armington model
- Model characterization: equilibrium existence and uniqueness; equilibrium in relative changes; stylized version of the model
- Estimation and calibration: data; estimation of σ
- Counterfactual experiments
- Sensitivity analysis: e.g. the value of σ

Summary

- Reduced-form vs. Structural modeling
- Example: Armington model
 - Build an equilibrium system
 - Numerically solve the equilibrium system
 - Bring the model to the data
- Structural modeling: sketch