

Implicit Utility and the Canonical Gravity Model

Russell H. Hillberry and Anton C. Yang

Purdue University

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Gravity Trade Puzzle

- Structural estimation of gravity models of trade transparently maps regression coefficients to structural parameters.
- Puzzle of separating trade costs from trade responses.
- Visually, e.g., following coefficients on log distances:

$$\rho(1 - \sigma) \log d_{ij}.$$

- YANG (2019) shows identification using the CDE preferences with easily accessible population data.

Question and lessons studied

- HY discussions: Can we do this in CES-gravity models?
- Theoretical demand structure is alone sufficient:

$$(1 - \sigma) \log[\quad ? \quad] \longrightarrow \rho(1 - \sigma) \log d_{ij}.$$

Related Papers

- HANOCH (1975): implicitly additive demand system;
- ANDERSON (1979): gravity model theory;
- MCCALLUM (1995): high border costs;
- HUMMELS (1999): *ad valorem* freight charges + tariff rates;
- EATON AND KORTUM (2002): Ricardian Framework;
- ANDERSON AND VAN WINCOOP (2003)/AvW: $\sigma = 5, 10, 20$;
- BALISTRERI AND HILLBERRY (2007)/BH: $\sigma = 5$;
- SIMONOVSKA AND WAUGH (2014): disaggregate prices + trade-flow data $\rightarrow \sigma \approx 4$;
- CALIENDO AND PARRO (2015): tariff data + asymmetric border;
- HEID, LARCH AND YOTOV (2017): non-discriminatory trade-policy variables $\rightarrow \sigma \in [4.3, 6.9]$;
- FEENSTRA (1994), BRODA AND WEINSTEIN (2006), SODERBERY (2015): imported quantities + unit values $\rightarrow \sigma$ at the product-level;
- PRECKEL, CRANFIELD AND HERTEL (2010), YANG, GOUEL AND HERTEL (2018), GOUEL AND GUIMBARD (2019): implicit additive direct models (MAIDADS) + MLE $\rightarrow \kappa, \mathbf{U}_j$'s;
- YANG (2019): CDE + population + MPEC $\rightarrow \sigma_i$'s, \mathbf{U}_j 's;
- YANG AND PRECKEL (2020): CDE + MLE $\rightarrow \mathbf{U}_j$'s;
- HILLBERRY AND YANG (2020): CES + MPEC $\rightarrow \sigma$.

A Solution from the Demand Theory

We bridge [HANOCH \(1975\)](#)'s demand theories with trade.

The representative consumer preferences in a region j are modeled by the following direct CES function: [Demand Parameterization \(click here\)](#)

$$U_j \equiv \left[\sum_i \alpha_i^{(1-\sigma)/\sigma} \left(\frac{T_{ij}}{t_{ij}} \right)^{(\sigma-1)/\sigma} \right]^{\sigma/(\sigma-1)}. \quad (1)$$

- T_{ij} is the quantity of shipment from i to j , which is
- melted by the iceberg trade cost variable $t_{ij} > 0$;
- U_j 's are the representative agent's utility in region j ;
- α_i 's > 0 are taste parameters;
- $\sigma > 0$ is the constant elasticity of substitution.

Trade Flows with CES price index

Then under region j 's national budget constraint, the nominal trade flow equation is given by:

$$FOB_i T_{ij} = Y_j \left(\frac{\alpha_i FOB_i t_{ij}}{P_j} \right)^{1-\sigma}, \quad (2)$$

- FOB_i is the domestic price of output units in region i ;
- Y_j is the national income in region j ;
- P_j is the (aggregate) consumer price index in region j .

P_j equals the inverse of shadow price resulted from the utility maximization, and for any CES demand function:

$$U_j = \frac{Y_j}{P_j}. \quad (3)$$

Implicit Trade Flows

Writing Eq. (3) into (2) allows us to rewrite Eq. (2) that is embedded with utility U_j :

$$FOB_i T_{ij} = U_j^{1-\sigma} (\alpha_i t_{ij})^{1-\sigma} FOB_i^{1-\sigma} Y_j^\sigma. \quad (4)$$

Motivations of Eq. (4):

- σ controls counterfactuals of T_{ij} , FOB_i (or t_{ij}) and Y_j ;
- Furthermore, $T_{ij} = (\alpha_i U_j)^{1-\sigma} (FOB_i t_{ij})^{-\sigma} Y_j^\sigma$;
- With \bar{U}_j : 1% \uparrow of FOB_i or $t_{ij} \implies \sigma\%$ \downarrow of T_{ij} ;
- Due to the generic σ , in order to hold U_j and T_{ij} fixed while prices \uparrow by 1%, the region- j consumer must be compensated by exactly a 1% \uparrow in Y_j ;
- Holding FOB_i 's and U_j 's fixed, the econometric exercise exploits variations in T_{ij} and Y_j .

Implicit Utility into the GE Framework

In equilibrium, each region j 's aggregate income must equal the total value of **CIF** goods purchased by region j 's consumer:

$$Y_j = \sum_i FOB_i T_{ij}. \quad (5)$$

Combining Eq. (5) with (4), it can be shown that

$$Y_j = \sum_i U_j^{1-\sigma} (\alpha_i t_{ij})^{1-\sigma} FOB_i^{1-\sigma} Y_j^\sigma. \quad (6)$$

It implies a representation of the implicit additive indirect function associated with the **AvW's gravity** + **Hanoch's implicit utility**:

$$\sum_i \alpha_i^{1-\sigma} U_j^{1-\sigma} \left(\frac{FOB_i t_{ij}}{Y_j} \right)^{1-\sigma} \equiv 1. \quad (7)$$

Trade costs into the GE Framework

Following BH, we account for both distance and asymmetric border effects on trade costs:

$$t_{ij} = d_{ij}^{\rho} \left[\exp \left(\frac{\mathbf{A}}{1 - \sigma} \right) \right]^{1 - \delta_{ij}} . \quad (8)$$

- d_{ij} is the distance between i and j observed from the data;
- ρ is the elasticity of trade costs with respect to distance;
- $\mathbf{A} = (1 - \sigma) \ln b_{ij}$ are the border coefficients;
- b_{ij} 's equal 1 plus tariff equivalent of border costs:
 $TR^{equiv} = b_{ij} - 1$.
- δ_{ij} 's are the dummy variables equaling 0 if shipments cross border, equaling 1 otherwise.

Identification

Let $X_{ij} = FOB_i T_{ij}$ denote the value of flows between i and j , it can be shown that the empirical form of gravity equation is given by

$$\log(X_{ij}) = (1 - \sigma) \log \alpha_i + (1 - \sigma) \log U_j + (1 - \sigma) \log FOB_i + \rho(1 - \sigma) \log d_{ij} + (1 - \delta_{ij})(1 - \sigma)b_{ij} + \sigma \log Y_j. \quad (9)$$

Identification:

- If we know the cardinal value of U_j , then we can identify σ ;
- With σ being pinned down, we can obtain the estimates of α_i 's, b_{ij} and ρ given the information on bilateral distances.
- Impossible in reduced-form approaches with U_j unobserved.
- This is because (standard CES-gravity) fixed effects would sweep out Y_j and U_j , and leave the product of ρ and $1 - \sigma$ unidentified.
- For this reason, we follow [YANG \(2019\)](#)'s structural approach to evaluate U_j simultaneously, using an MPEC algorithm.

BH (JIE, 2007)'s Computable General Equilibrium

BH constructs a $4n$ system of equations that is an operational GE.

(1) Income definition:

$$Y_i = FOB_i E_i^0.$$

(2) Goods market-clearing condition:

$$E_i^0 = \sum_j \left[\frac{Y_j}{FOB_i} \left(\frac{\alpha_i FOB_i t_{ij}}{P_j} \right)^{1-\sigma} \right].$$

(3) Unit expenditure function:

$$P_j = \left[\sum_i (\alpha_i FOB_i t_{ij})^{1-\sigma} \right]^{1/(1-\sigma)}.$$

(4) Income balance:

$$U_i P_i = Y_i.$$

Linkage with HY's 3n CGE in the Estimation System:

- BH model the $4n$ -system equations s.t. LS in their estimation system.
- We apply Hanoch's implicit utility theorem and reduce the system to $3n$.
- We explicitly evaluate U given the system of constraints (next slide).

3n GE System of Equations as Constraints

(1) National income definition (**income**):

$$Y_i = FOB_i E_i^0 \quad \longrightarrow \quad \text{MCP.Y}$$

(2) National endowment identity (**supply**):

$$E_i^0 = \sum_j U_j^{1-\sigma} (\alpha_i t_{ij})^{1-\sigma} FOB_i^{-\sigma} Y_j^\sigma \quad \longrightarrow \quad \text{MCP.FOB}$$

(3) Preferences definition (**demand**):

$$\sum_i \alpha_i^{1-\sigma} U_j^{1-\sigma} \left(\frac{FOB_i t_{ij}}{Y_j} \right)^{1-\sigma} \equiv \kappa \quad \longrightarrow \quad \text{MCP.P} \quad \left[\text{MCP.U} = \frac{Y}{P} \right]$$

- E_i^0 is region i 's fixed endowment; **red**: complementary variables;
- Equation (3) is both a GE environment and normalization;
- We let the computation algorithm determine the data-generating scaling factor of U_j , expressed by some $\kappa > 0$;
- This system formulates a mixed complementarity problem (**MCP**).

MPEC and MCPs

- We formulate the constrained optimization problem using the algorithm of mathematical programming with equilibrium constraints (MPEC). [A PPML-MPEC constrained optimization \(click here\)](#)
- Popular in solving optimization of engineering problems.
- An appropriate candidate of solving constraints that are highly non-linear and are MCPs.
- The MCPs, in turn, verify that our $3n$ GE system is operational, via homogeneity test and check of Walras' Law.
- Recent literature using MPEC to solve GE gravity models of trade include: BH, [BALISTRERI, HILLBERRY AND RUTHERFORD \(2011\)](#), [TAN \(2012\)](#), [YANG \(2019\)](#).
- The MPEC program is implemented in GAMS version 31.1.1 with the help of the preprocessor using GAMS-F tool.

Empirical Verification of Theoretical Approach

- To verify that our approach is consistent with empirical findings, we replicate BH's results who structurally estimate AvW's coefficient $a1 = -1.44$.
- The authors use least squares (LS) as objective, and fixed $\sigma = 5$. Since $a1 = \rho(1 - \sigma)$, this implies $\rho = 0.36$.
- Thus, our theoretical structure must hypothetically yield exactly the **same result**, provided that we use the same data, econometric models, and identically exogenize the level of σ .

Relevant procedures:

- Step 1: we first replicate BH's model using explicit direct CES.
- Step 2: we replicate the model again using our implicit model.
- Step 3: confirmed that our model consistently yields $\rho = 0.36$.
- Step 4: change objectives to PPML, and repeat step 1 and 2.
- Step 5: verified empirical consistency again in both models.

Freeing σ with Symmetric Border Costs

- We directly solve the canonical AvW's model of aggregate trade.
- In this step, we release σ and directly estimate σ , ρ , and b under the same equilibrium constraints and normalization scheme that would have replicated BH/AvW if σ were fixed;
- In this exercise, we will use LS estimator as in BH;
- Defining the fitted value

$$\hat{z}_{ij} = \log\left(\frac{X_{ij}}{Y_j}\right)$$

$$= (1 - \sigma)[\log \alpha_i + \log(U_j) + \log(FOB_i) + \log t_{ij} - \log(Y_j)];$$

- Econometric specification given by $\min \sum_i \sum_j [z_{ij} - \hat{z}_{ij}]^2$;
- Constrained by (i): Eq (1) - (3);
- and (ii) AvW/BH's normalization for the scale of utility;
- **AcW's data**: 30 US states, 10 Canadian provinces, 1 rest of the US, 1,551 non-zero trade-flows observations.

Check the Excess Degrees of Freedom

- Result 1: replicated that $\rho = 0.36$, when $\sigma = 5$ is held fixed;
- Result 2: with LS, $\sigma = 1.62$, $\rho = 2.31$, $b = 2.96$.
- We stop here, and following [YANG AND PRECKEL \(2020\)](#) to ask whether there are additional degrees of freedom in the parameter spaces that can be removed.
- That is, whether additional normalization is needed as we are moving to freeing σ ? Latitude for changing parameters?
- Thus, we construct a new problem by maximizing and minimizing σ while including the original estimation problem and requiring that the sum of squared residual is at least as small as computed from the estimation problem.
- We then repeat this procedure for ρ and b .
- Conclusion: No, the parameter values cannot be changed significantly without increasing the residuals.

Structural Estimation and Bootstrapping

Table 1: Structural estimation with implicit and explicit representation

| | BH replication with explicit representation | BH replication with implicit representation | Structural estimation with implicit representation | Implicit representation with regional specific σ 's |
|---|---|---|--|--|
| | (1) | (2) | (3) | (4) |
| $a1 = (1 - \sigma)\rho$ | -1.44 | -1.44 | -1.44 | |
| $a2 = (1 - \sigma) \ln b_{US-CA}$ | -1.85 | -1.85 | -1.85 | |
| $a3 = (1 - \sigma) \ln b_{CA-US}$ | -1.85 | -1.85 | -1.85 | |
| $\bar{a}1 = (1 - \bar{\sigma})\rho$ | | | | -1.47 |
| $\bar{a}2 = (1 - \bar{\sigma}) \ln b_{US-CA}$ | | | | -1.39 |
| $\bar{a}3 = (1 - \bar{\sigma}) \ln b_{CA-US}$ | | | | -1.39 |
| σ | 5 (assigned) | 5 (assigned) | 1.62 (0.005) | |
| $\bar{\sigma}$ | | | | 1.81 (0.02) |
| ρ | 0.36 (0.005) | 0.36 (0.005) | 2.31 (0.03) | 1.82 (0.06) |
| $\ln b_{US-CA}$ | 0.46 (0.02) | 0.46 (0.02) | 2.96 (0.14) | 1.72 (0.12) |
| $\ln b_{CA-US}$ | 0.46 (0.02) | 0.46 (0.02) | 2.96 (0.14) | 1.72 (0.12) |
| N | 1511 | 1511 | 1511 | 1511 |
| Sum of squared residuals | 2262.84 | 2262.84 | 2262.84 | 1286.03 |

Standard errors across columns in “()” obtained from 2,000 bootstrap resamples.

Fitted into PPML with Trade Flows Objectives

- Following GOURIEROUX, MONFORT AND TROGNON (1984) and SANTOS SILVA AND TENREYRO (2006), with asymmetric border costs b_{ij} , the PPML estimator is given by

$$\begin{aligned}
 X_{ij} &= \exp \left\{ \left| (1 - \sigma) \log \alpha_i \right| + (1 - \sigma) \log U_j + (1 - \sigma) \log FOB_i \right. \\
 &\quad \left. + \rho(1 - \sigma) \log d_{ij} + (1 - \delta_{ij})(1 - \sigma)b_{ij} + \sigma \log Y_j \right\} + \varepsilon_{ij} \\
 &= \exp(x_{ij}g_i) + \varepsilon_{ij}.
 \end{aligned}$$

- $x_{ij}g_i$ is a proxy representing everything inside the curved bracket;
- ε_{ij} is the disturbance term;
- we repeat the check of robustness and bootstrapping as for LS;
- we release the symmetric border assumption and repeat all steps.

Concluding Remarks

- We bridge the demand theory with trade literature and show that an alternative but identical CES-gravity can achieve identification via an application of a canonical gravity model.
- We demonstrate that theoretical structure is alone sufficient for identifying σ , ρ and b , without adding any more data.
- We generalize the standard trade flows to implicit trade flows.
- The procedure allows evaluation of the utility index, which is critical to identifying structural parameters:

$$(1 - \sigma) \log[U_j] \longrightarrow \rho(1 - \sigma) \log d_{ij}.$$

- We show that the MPEC algorithm is useful to calculate cardinal value of utility (even with the CES function) with MCPs.

Thank You!

Questions or comments?

Demand Parameterization

- HANOCH (1975): $G(\frac{p}{w}, u) = \sum_i \beta_i u^{e_i(1-\alpha_i)} (\frac{p_i}{E})^{1-\alpha_i} \equiv 1$
- The following indirect CES function is its special case:

$$U \equiv \left[\sum_i \beta_i \left(\frac{p_i}{E} \right)^{1-\sigma} \right]^{1/\sigma-1},$$

- which is also a parametric transformation from the preferences in BALISTRERI AND HILLBERRY (2007):

$$U \equiv \left[\sum_i \alpha_i^{(1-\sigma)/\sigma} Q_i^{(\sigma-1)/\sigma} \right]^{\sigma/(\sigma-1)},$$
$$\alpha_i^{1-\sigma} = \beta_i > 0 \quad \forall i.$$

A PPML Constrained Optimization

The PPML-MPEC problem is formulated as follows:

$$\max_{\mathbf{g}_i = \{\alpha_i, \sigma, \rho\}, U_j, b_{ij}, 0 < \kappa < c} L(b_i) = \text{constant} - \sum_i \sum_j \exp(x_{ij} g_i) + \sum_i \sum_j y_{ij} x_{ij} g_i$$

s.t.

- (i) GE($\mathbf{g}_i, U_j, b_{ij}$) [set of GE constraints]
- (ii) CES($\mathbf{g}_i, U_j, b_{ij}$) $\equiv \kappa$ [CES additivity constraint]
- (iii) $\alpha_i, \sigma > 0$
- (iv) $U_j, \rho > 0$
- (v) $b_{ij} \geq 0$
- (vi) $\left(\frac{Y_{\text{Alabama}}}{U_{\text{Alabama}}} \right)^{1-\sigma} = \sum_i \left[\frac{U_i^{\text{Alabama}, j}}{\sum_j Y_j} \right]^{1-\sigma}$.

- The last equilibrium constraint is a specific normalization that is equivalent to AvW and BH.

[Head back to MPEC \(click here\)](#)

[Head back to PPML \(click here\)](#)

Inequality Constraints as an MCP1

The market clearing conditions imply that the strict equalities would hold if and only if the associated goods are free of charge

$$E_i^0 \geq \sum_j U_j^{1-\sigma} (\alpha_i t_{ij})^{1-\sigma} FOB_i^{-\sigma} Y_j^\sigma \quad \perp \quad FOB_i \geq 0.$$

[Head back \(click here\)](#)