## Heterogeneity and the Distance Puzzle

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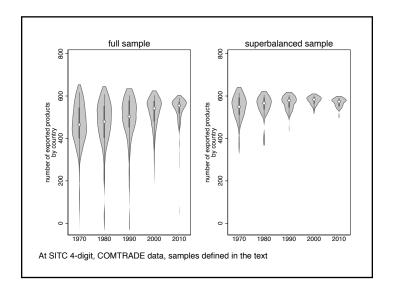
## Introduction: the paradox of distance

- ▶ The distance effect is increasing or stable through time in gravity models: Disdier & Head, (2008), Head & Mayer (2013)...
  - This seems counter-intuitive ("Death of distance")
- ▶ Various answers in the literature (see Head & Mayer (2013)):
  - Problem with the log-linear estimation strategy?
    - ▶ Not taking zeros into account + heteroskedasticity
    - ➤ PPML estimates (Santo Silva & Tenreyro (2006), , Bosquet & Boulhol (2015))
  - ► Composition effect (Larch et al. (2016))
  - ▶ Relative evolution of short-distance trade costs compared to long-distance trade costs (Buch & al. (2004), Krautheim (2012))
  - ▶ Input-output linkage (Daudin et al. (2011))
  - Network : the distance coefficient does not depend on trade costs (Chaney (2018))

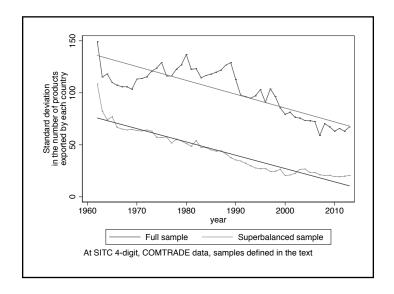
#### How do we contribute?

- Every theoretical fundation of the gravity equation delivers a relationship betwee the distance elasticity and a degree of structural heterogeneity in some model-specific structural dimension
  - → "Trade elasticity" (Arkolakis et al. (2012))
- ▶ The distance coefficient is the product of:
  - 1. The elasticity of distance to trade costs
  - 2. The elasticity of trade to trade costs
- Empirical evidence on the historical evolution of structural heterogeneity is notoriously scarce
  - ► The only other try we know of is Berthelon & Freund (2008) from the late 1980s to early 2000s
- ► We document over 1962-2013 how the increasing substituability of the bundles shipped out by each country (Armington framework) contributes to the paradox

# Dispersion of the number of products exported by each country



# Standard deviation in the number of products exported by each country



#### Overview

#### The distance puzzle in our data

Benchmark estimation Composition, sample, FTA effects

#### Interpreting properly the distance coefficient

The heterogeneity dimension in each model The heterogeneity dimension captured in our data

#### Estimation strategy and results

Benchmark estimation Robustness checks

#### Conclusion

# Summary of results

- Robustness of distance puzzle in 1962-2013: increase in distance coefficient
  - ▶ +5% controlling for estimation strategy
  - ▶ +31% controlling for composition and sample effects
- Evolution of heterogeneity parameter:
  - ??% increase in 1963-2013 (??% in 1970-2013)
  - this estimate is likely to be a lower bound
- Elasticity of trade costs to distance has not increased
  - ???% decrease in 1963-2013
  - ??% decrease in 1970-2013
- Which dimension of increased country similarity?
  - Result obtained within the Armington framework
  - Increased substitutability of traded product bundles

## Roadmap

#### The distance puzzle in our data

Benchmark estimation Composition, sample, FTA effects

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The heterogeneity dimension in each model
The heterogeneity dimension captured in our data

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# Estimation procedure (1)

- run gravity equations (obviously)
  - ► COMTRADE data, 1962-2009 STIC 4 digits (1962-2009)
  - cross section, no panel
  - focus on evolution of distance elasticity overtime
  - using the PPML estimator (consistency & efficiency)
- ▶ Microfounded gravity equation (Anderson & Wincoop (2003)):

$$X_{ij,t} = \left(\frac{Y_{i,t}Y_{j,t}}{Y_t}\right) \left(\frac{\tau_{ij,t}}{\Pi_{i,t}P_{j,t}}\right)^{\epsilon_t}$$

• heterogeneity parameter:  $\epsilon_t$ :  $1-\sigma_t$  in Armington (sector or firm productivity heterogeneity in Metlitz and Eaton & Kortum framworks )

# Estimation procedure (2)

- Trade costs:
  - ▶ distance parameter: ð<sub>ij</sub>
  - ▶ time-invariant cost vector of controls (adjacency,...): Z
  - time-varying cost vector of controls (policy: FTAs,...):  $S_t$
  - ▶ unobserved bilateral trade cost component assumed to have mean zero conditional on the observables:  $\nu_{ijt}$

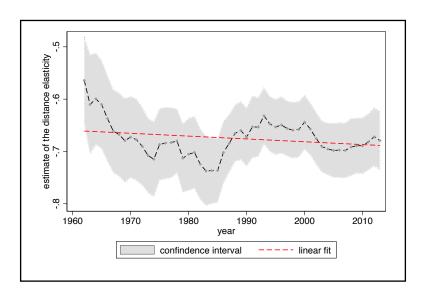
$$\tau_{ijt} = \exp\left\{\rho_t \ln \eth_{ij} + Z_{ij}\zeta_t + S_{ijt}\zeta_t + \nu_{ijt}\right\}$$

- $\rho_t$  is the 'world shrinkage' parameter i.e. elasticity of trade costs to distance
- Estimated equation:

$$X_{ij,t} = \exp\left(\xi_t - \delta_t \ln \eth_{ij} + Z_{ij}\tilde{\zeta}_t + S_{ijt}\tilde{\zeta}_t + f_{it} + f_{jt}\right)\eta_{ijt}$$

- $ightharpoonup f_{it}$  and  $f_{it}$  are fixed effects to control for price levels
- ξ<sub>ijt</sub> is a multiplicative error term which includes the exponentiated unobserved bilateral trade cost
- distance elasticity:  $-\delta_t = \epsilon_t \rho_t$

# Baseline regression (PPML)

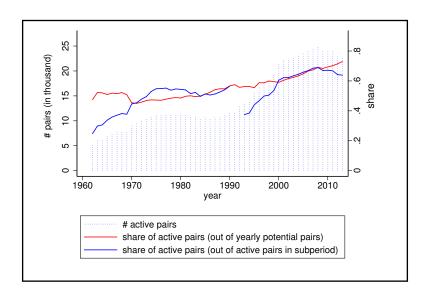


Decline of 4.5%: basically stable

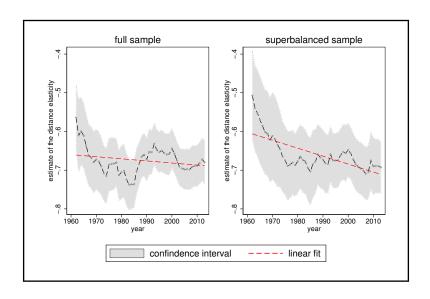
## Sample composition effect

- We know the country sample potentially matters
  - Increasing number of new low volume long-distance relationship
  - ▶ Potentially increases the distance elasticity of trade (Mayer et al. (2019), Head & Mayer (2013) )
  - ▶ Though it should be less of an issue with PPML
- There are big sample issues in the data
- ► Test: keep only trading pairs that have reciprocal non-zero trade every year from 1962 to 2009 ("Superbalanced sample")
  - ▶ It deepens the puzzle

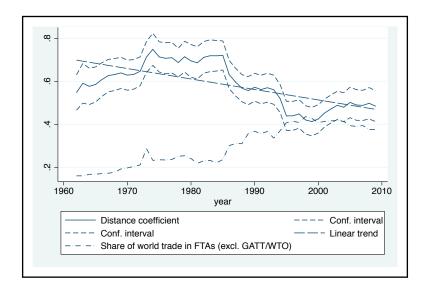
## Share of active pairs



# Superbalanced sample

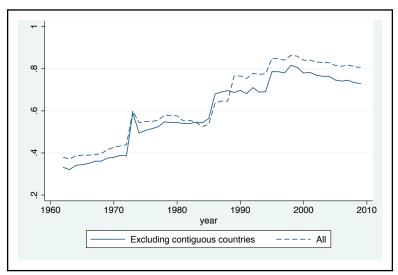


# FTAs(1)



# FTAs(2)

Figure: Share of intra-FTA trade among nearby countries (2000km or less)



# Sector composition

- ► Test: suppose the composition of trade constant i.e. at 1962 shares for 4 digit goods
- ▶ It deepens the puzzle (increse in manuf share)
- FTAs
- Test: introduce FTA variables
- It 'solves' the puzzle
- But what does it mean?
- Increasing number of proximity controls overtime
- ► Mechanically reduces the effect of distance

# Summary (PPML)

	% change relatively to baseline	Total change 1962-2009
Baseline		1.07
Sample effect	7%	1.14
Composition effect	7%	1.14
FTA effect	-54%	0.49
Composition + sample	7%	1.14
Composition + FTA	-29%	0.75
Sample + FTA	-59%	0.44
Sample + Composition + FTA	-54%	0.49

## Ingredients of the puzzle

- The distance coefficient is the elasticity of trade to distance
  - ▶ Trivial: the whole point of log-linear equations
  - Still the case in the Poisson specification
- It is a product of two coefficients:
  - Elasticity of trade flows to trade costs ζ
  - Elasticity of trade costs to distance  $\rho$
- ➤ The 'death of distance' intuition is really about the elasticity of trade costs to distance
- Which should be going down
- But it does not tell much about the heterogeneity dimension,
   i.e. the trade elasticity ζ

# Short incursion in microfoundations (1)

- The gravity equation can be justified by three families of theories:
- Ricardian framework
   Homogeneous goods
   Shop around the world for lowest cost supplier (intersectoral productivity heterogeneity)
- Heterogeneous firms framework:
   Trade because all firms produce different varieties
   A subset of firms enters export markets (intrasectoral productivity heterogeneity)
- Armington framework
   Trade because consumers value variety
   Country-specific goods (heterogeneity: degree of substitutability between bundles)

# Short incursion in microfoundations (2)

- Ricardian framework:
  - **Distance coefficient:**  $\rho\theta$
  - lacktriangledown heta captures intersectoral productivity dispersion
  - if sectors have similar productivity
    - $\rightarrow$  small differences in variable costs have a large effect on trade flows
    - $\rightarrow$  high elasticity of trade to trade costs
- Monopolistic competition between heterogeneous firms:
  - Distance coefficient:  $\rho\gamma$
  - $ightharpoonup \gamma$  captures productivity dispersion across firms (parameter of Pareto)
  - if distribution decays swiftly, higher probability that productivity cut-off for exporting is close to the mass of firms
    - $\rightarrow$  small differences in variable costs have large effect on entry
    - $\rightarrow$  high elasticity of trade to trade costs

# Short incursion in microfoundations (3)

- Armington framework
  - ▶ Distance coefficent:  $\rho(\sigma 1)$
  - $ightharpoonup \sigma$  captures degree of similarity between country-specific product bundles
  - if the set of goods produced by different countries is similar
    - → high Armington elasticity
    - $\rightarrow$  high elasticity of trade to trade costs
- ► In all cases: elasticity of trade flows to trade costs is inversely related to heterogeneity

# Measuring the trade elasticity

- Features of our data: information on bilateral trade flows and unit values
- To measure efficiency heterogeneity: need information on domestic prices
  - intuition: country-specific cut-off for entry common to all exporters
  - price distribution in destination across all sources needed to estimate shape parameter of productivity distribution
- However we can measure substitutability across frameworks
  - use variation of market shares of country-level composite goods across export markets
  - construct relative prices of product bundles
  - estimate the aggregate Armington elasticity in cross section
- ► The estimated parameter is the trade elasticity in the Armington framework

## Relative prices of product bundles

- Consistent aggregation procedure to get relative prices
  - ► CES preferences at inter- and intrasectoral level
  - ▶ Intra- and intersectoral elasticities assumed equal
  - ▶ Write sector-specific demand equation
  - Sum across all sectors
- ► Gives market share equation for aggregate bilateral trade as a function of the weighted average of sectoral relative prices of exporter in destination

$$\ln \left[ \frac{X_{ij}}{Y_j} \right] \approx -(\sigma - 1) \ln \left[ \sum_{k=1}^K \omega_j(k) \frac{P_{ij}(k)}{P_j(k)} \right]$$

Exponentiating gives equation estimated in Poisson:

$$X_{ij}/Y_j = \exp\left[\lambda_0 - (\sigma - 1)\ln\left(\sum_k \omega_k \frac{P_{ij}(k)}{P_j(k)}\right) + fe_i + fe_j\right]\eta_{ij}$$

# Dealing with missing unit values

- Trade flow observed, but information on quantities missing
- ▶ On average, this is the case for 14% of total trade
- Use stepwise price imputation procedure
  - construct relative prices at highest disaggregation level
  - construct next level relative price as weighted average of observed relative prices
  - destination-specific weights at each step
  - repeat at each aggregation level
- assumption: missing unit values can be best approximated by observed prices for similar goods

# Dealing with zero trade flows

- Under model assumptions some trade would be observed in every sector between each pair
- ➤ Zero trade flows prevalent: from 86-90% of possible observations at 4-digit level
- Assumption: statistical, not structural zeros linked to data collection thresholds
- Same stepwise procedure used for price imputation
- Corresponds to assumption that unobserved relative price equal to observed
- ▶ Problem: unobserved prices much higher than imputed prices

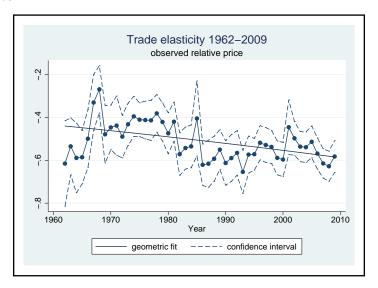
# Proportion of zero trade flows as a function of market share

Share of ZTF		
ms	-0.0427 <b>***</b> (0.0001)	-0.2573 <b>***</b> (0.013)
year	-0.0033 <b>***</b> (0.0000)	-0.0024 <b>***</b> (0.000)
ms * year		0.0001*** (0.000)
constant	6.0976 <b>***</b> (0.0366)	4.2515 <b>***</b> (0.134)
Observations	657001	657001

#### Overestimation bias

- Underestimation factor not constant across exporters
  - share of ztf decreasing in market share
  - reduction in share of ztf proceeds at quicker pace for small exporters
- Relative price underestimated by more for small exporters
- For given distribution of market shares, true underlying distribution of prices is greater than observed distribution
- Estimated parameter overestimates the true substitutability parameter
- But less so overtime
- If estimated elasticity increases, this is a lower bound on true parameter evolution

### Results



- ▶ 33% increase in parameter 1962-2009
- corresponds to annual increase of .6% per year

# Changing the dataset: BACI

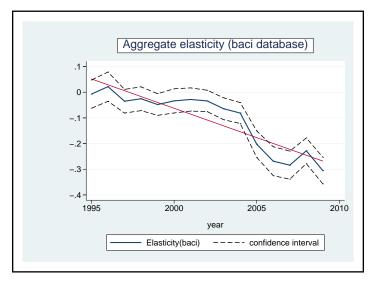


Figure: Estimated  $(1 - \tilde{\sigma})$ , BACI database

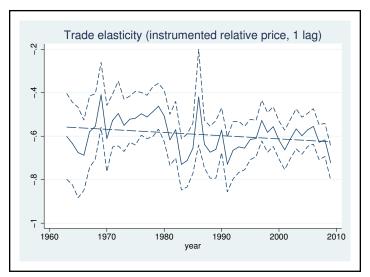
## Instrumenting: motivation

- Results subject to caution?
  - attenuation bias (if supply schedules not horizontal)
  - ▶ matters not only for level, but for evolution (Feenstra(1994))
- ► Objective: capture exporter-specific shocks to the price of the composite good which are not demand-driven
- Indicator: GDP price level (Penn World Tables: 189 countries, 1950-2009)

## Instrumenting: procedure

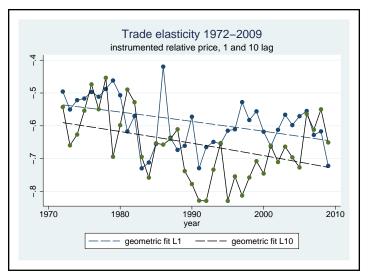
- compute relative prices for exporter-specific composite goods
- compute evolution of GDP price levels of trading partners, weighted by market shares (common currency)
- compute hypothetical relative price in t for each exporter as:
  - product of its relative price in (t s)
  - evolution of its GDP price level between t and (t s) relatively to all other partners
- predict relative price of each exporter in t: regress observed relative price on hypothetical relative price.
- Idea: get an instrumented relative price which depends on past relative price and relative evolution of GDP price level.
- Estimate market share equation using instrumented relative prices

# Instrumenting: one lag



- reassuring: level of parameter increases by 9%
- results on evolution hold: 13% increase

# Increasing the number of lags



- ▶ level increases with number of lags: 22% for 10 lags
- results on evolution hold: 23% increase in 1972-2009

## Is there a distance puzzle left?

- empirical evidence on 13% increase in substitutability parameter
- this is aggregate trade elasticity in Armington framework
- combining with 7% increase in distance elasticity
- provides a direct explanation of the distance puzzle
- economic interpretation of increased perceived substitutability of product bundles?
  - increasing similarity in set of traded goods across countries
  - composition effects (changes in range or shares of traded goods)
- Not done: separate out net effect of increased perceived similarity
- ▶ Not feasible? parameter estimated on aggregate data is not a weighted average of sectoral parameters