

Entrepôt

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Preliminary

Motivation

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- ▶ Stiff historic and contemporary competition to become entrepôts
 - ▶ Saudi Arabia: \$7bn to be the “major east-west marine transshipment location.” (FT 2015)
 - ▶ India: \$5bn in new ports to compete with established hubs (Reuters 2016)
 - ▶ Singapore: \$1bn to “stay ahead of the curve as a world-class hub port” (Int. Port Tech. 2018) following \$3bn in automation (Ship & Bunker 2012)

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3. What are the positive (or negative) regional spillovers of entrepôts?

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- ▶ Counterfactuals: evaluate effects of (1) Hard Brexit and (2) opening NE passage.
 - ▶ Network generates first-order, localized effects.

Contributions and Related Literature

- ▶ Provide evidence on how global shipping networks inform international trade
 - ▶ Previous papers only utilize data on ships calling at ports ([Kojaku et al \(2019\)](#), [Wang and Wang \(2011\)](#))
 - ▶ Endogenize transport costs as part of a global network of shipping routes ([Brancaccio et al \(2019\)](#), [Hummels \(2007\)](#), [Limao and Venables \(2001\)](#))
 - ▶ Network effects of the container shipping technology on international trade ([Bernhofen et al \(2016\)](#), [Cosar and Demir \(2018\)](#), [Rua \(2014\)](#), [Wong \(2019\)](#))
- ▶ Quantify the effects of global shipping networks through a GE economic geography model
 - ▶ Extend Armington route choice framework ([Allen and Arkolakis, AA \(2019\)](#)) to include Ricardian industry-level comparative advantage ([EK \(2002\)](#))
 - ▶ Trade cost changes and infrastructure investment at nodes (entrepôts) and where spillovers between nodes may be negative due to scale economies ([Fajgelbaum & Schaal \(2017\)](#), [Ducruet et al. \(2019\)](#))
 - ▶ Economies of scale in shipping by estimating a scale economy with respect to volume of traffic ([Anderson et al \(2016\)](#), [Holmes and Singer \(2018\)](#))

Data

Stylized Facts

Model: Overview

Estimation

Counterfactual

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Ports of Call

- ▶ AIS transponder information on (90% of) containership entry and exit into (1,200) ports



Dots represent the ports in our data set. Line represents containership journies between port pairs .

Ports of Call

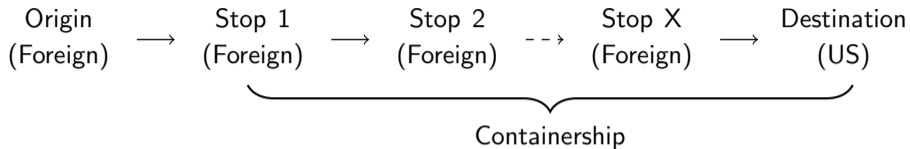
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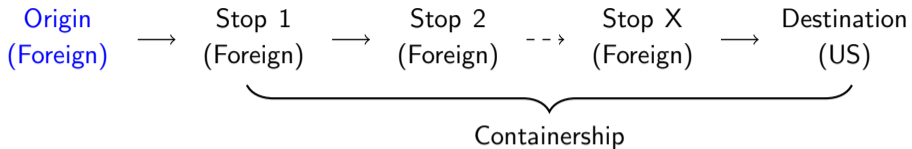
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- ▶ Containership movements do not necessarily capture the journey of container shipments.

Combined AIS and Bill of Lading Data

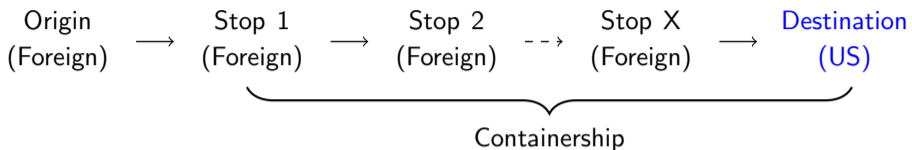


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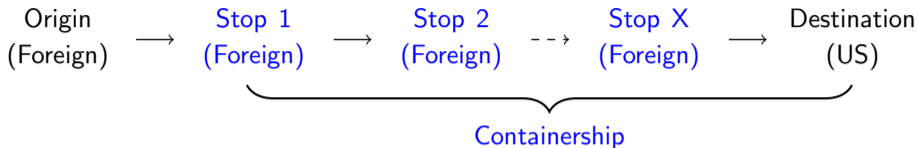
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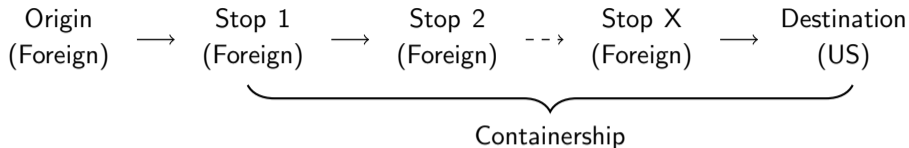
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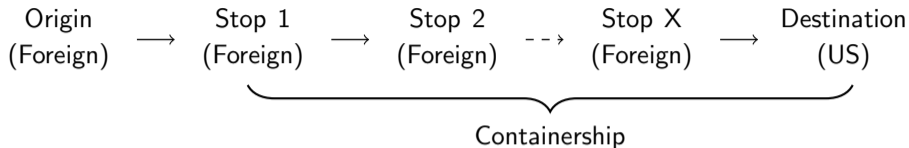
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- ▶ We match 90% of incoming containers

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Stylized Facts

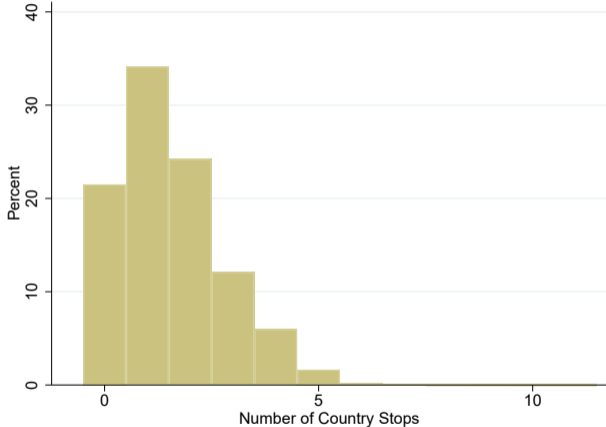
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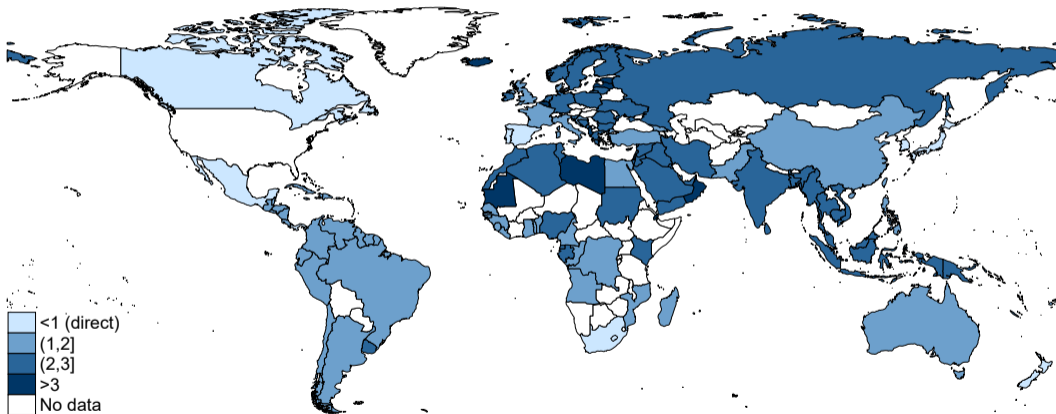


▶ >70% by volume are indirect.

Weight and Value Port Stops

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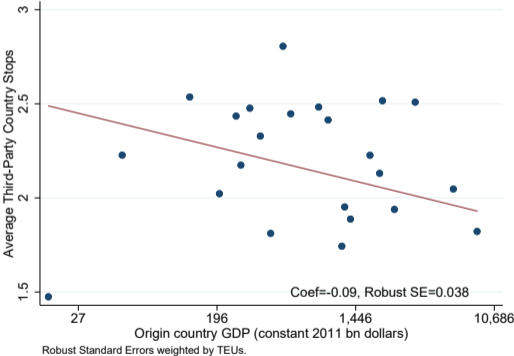
Stylized Fact 1: *The majority of containerized trade into the US is indirect.*



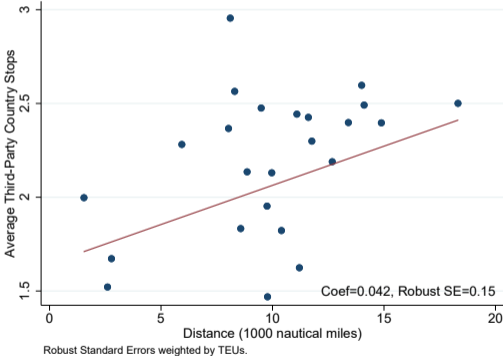
Average number of stops, by origin

Variation in Directness

Stylized Fact 2: *There is significant variation in this indirectness across countries—larger and closer countries are more likely to ship directly.*



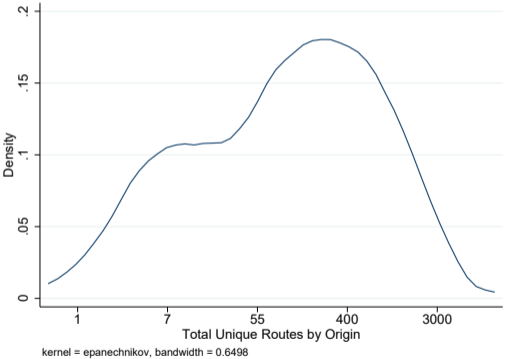
(a) Stops vs. Country size



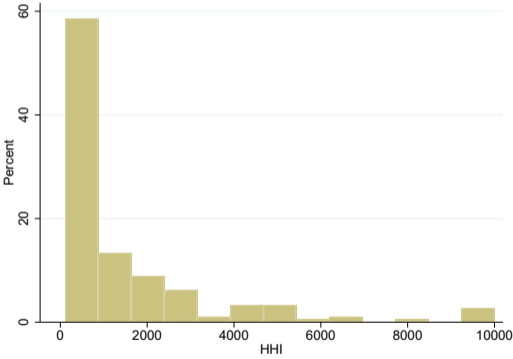
(b) Stops vs Distance

Within-country variation

Stylized Fact 3: *From a single origin, trade is on averaged dispersed through a large number of routes.*



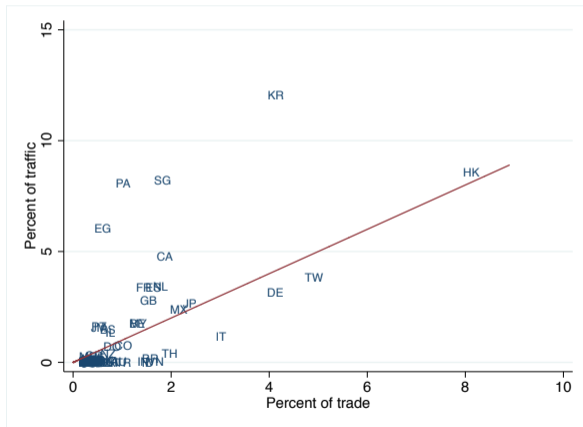
(a) Distribution of Unique Routes



(b) Distribution of Route Concentration

Concentration of Through-Shipments

Stylized Fact 4: *Shipping is concentrated through a minority of countries which account for a disproportionate share of third-party stops.*



Traffic vs. Trade

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 1. Multilateral resistance
 2. Non-transportation trade costs
 3. Multiple industries with variable trade and production costs
- ▶ Estimating equation backs out the costs of traveling each link in network from the observed traffic and trade volumes

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- ▶ To export to any j , competitive producers pay tariffs κ_{ijn} and iceberg transport cost $\tau_{nijr}(\omega)$ that depends on their chosen shipping route r :

$$p_{ijn}(\omega) = c_{in}\kappa_{ijn}\tau_{nijr}(\omega)$$

Endogenous Transport Costs (AA 2019)

- ▶ Total transport cost involves $\tilde{\tau}_{nijr}$ and a route-specific idiosyncratic cost shock

$$\tau_{nijr}(\omega) = \frac{1}{\epsilon_{ijnr}(\omega)} \tilde{\tau}_{nijr}$$

- ▶ The common transport cost from i to j on shipping route r is $\tilde{\tau}_{ijr}$

$$\tilde{\tau}_{ijr} = \prod_{k=1}^{K_r} t_{k_r-1, k_r}$$

where t_{k_r-1, k_r} is the leg-specific cost going directly from $k_r - 1$ to k_r

Equilibrium Traffic

- ▶ Summing across routes r that goes through leg k, l , express share of exports in industry n from origin i to destination j that pass through leg k, l as

$$\pi_{ijn}^{kl} = [(c_{in} \kappa_{ijn}) \cdot \tau_{nik} t_{nkl} \tau_{nlj}]^{-\theta} \cdot \Phi_{jn}^{-1}$$

- ▶ τ_{nij} is the average cost to ship from i to j
- ▶ $\Phi_{jn} = \sum_{i'} (c_{i'n} \kappa_{i'jn} \tau_{i'j})^{-\theta}$ is multilateral resistance, accounts for costs and connectivity of all other competitors i'

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- ▶ For a set of industries that share transport costs, total traffic between k and l :

$$\Xi_{klN} \equiv \sum_i \sum_j X_{ijN} \cdot [\tau_{ikN} t_{kln} \tau_{ljN} \tau_{ijN}^{-1}]^{-\theta}$$

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Estimation

- ▶ Objective: estimate transport cost between locations
 - ▶ One issue: land borders. Solution: parameterize

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- ▶ Estimation routine:

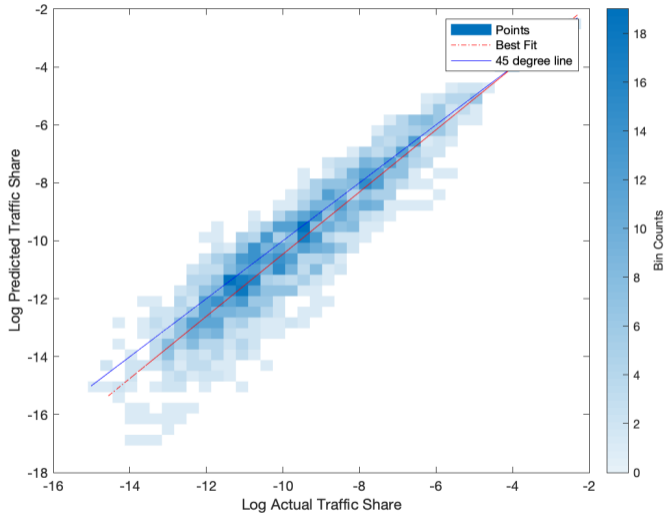
1. Guess β
2. Find t_{kl}, τ_{kl}
3. Find predicted traffic:

$$\Xi_{kl}^{predicted} \equiv \sum_i \sum_j X_{ij} \cdot \left[\tau_{ik} t_{kl} \tau_{lj} \tau_{ij}^{-1} \right]^{-\theta}$$

4. Minimize difference between predicted and observed:

$$\arg_{\beta} \min \sum_{kl \neq \text{land borders}} \left| \Xi_{kl}^{observed} - \Xi_{kl}^{predicted} \right|$$

Model Fit



Route Cost Estimates



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4. Multilateral resistance: change in t_{kl} also stiffens competitions at j by allowing other countries better access to j

When $t_{kl} \downarrow$, first 3 terms will increase X_{ij} . The 4th term shows potential decrease if the shift differentially favors trade and production costs from other countries to j

Counterfactuals

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- ▶ Calculate trade flow and welfare changes using hat algebra (Dekle, Eaton, & Kortum (2008))

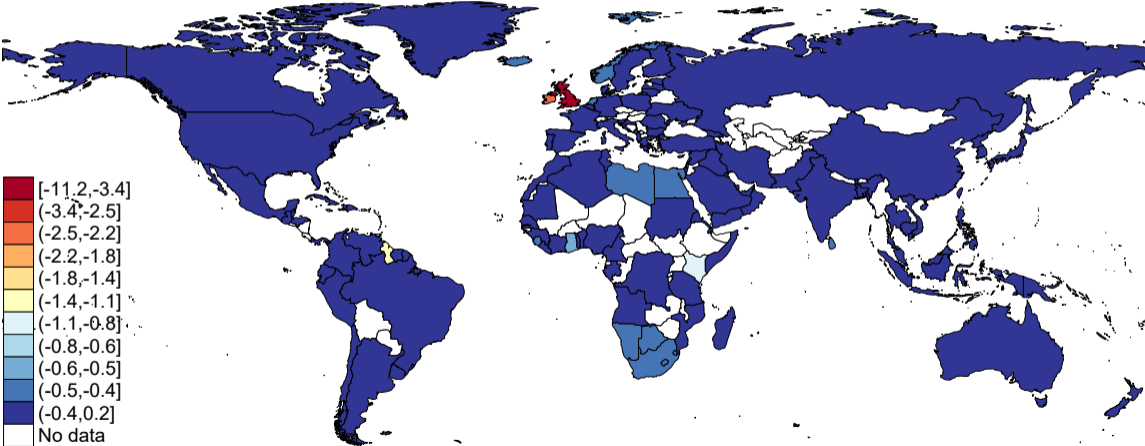
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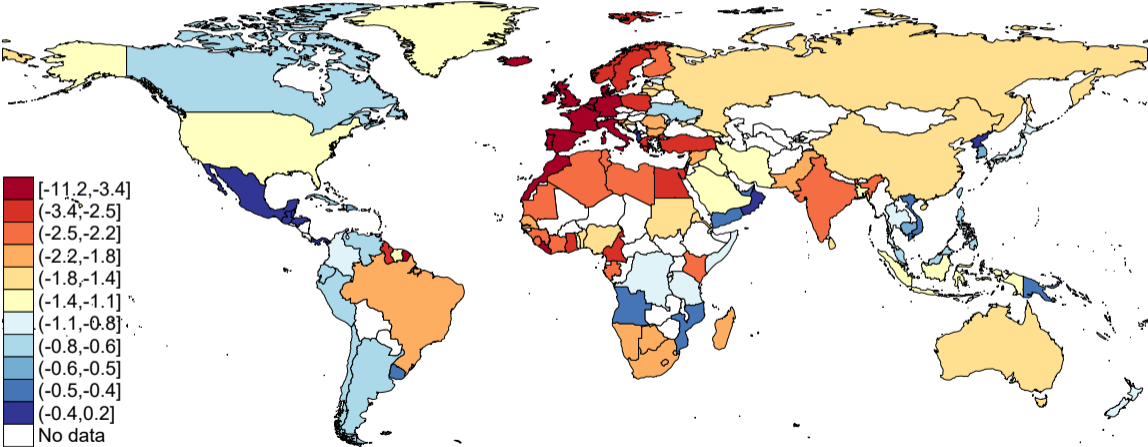
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- ▶ Scale economies will change results

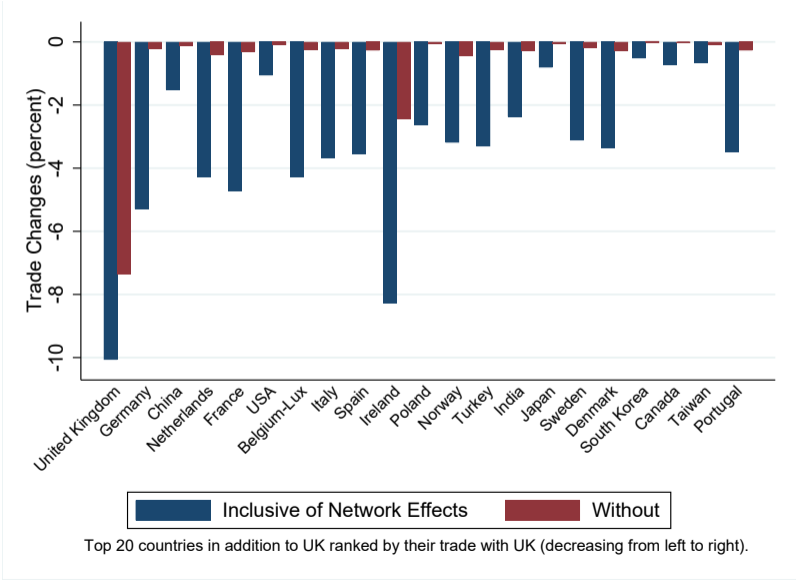
Hard Brexit: No Network Effects



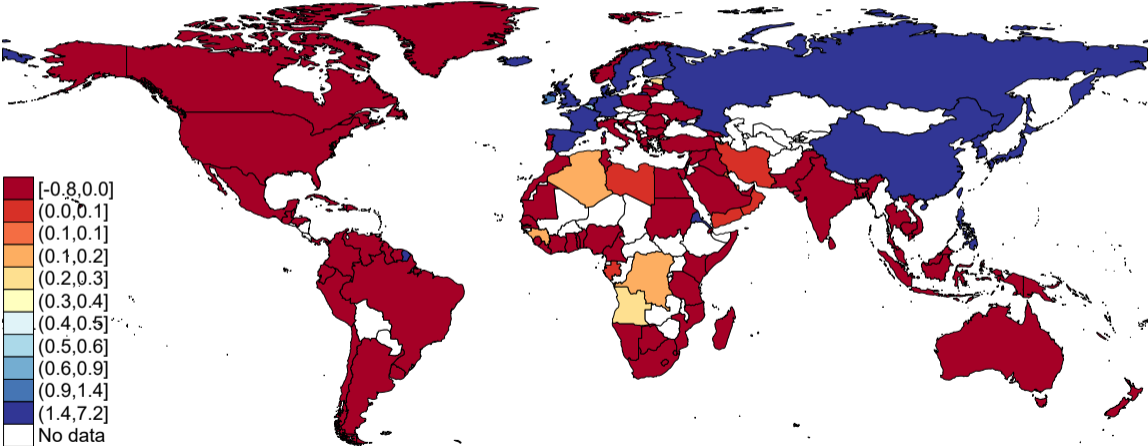
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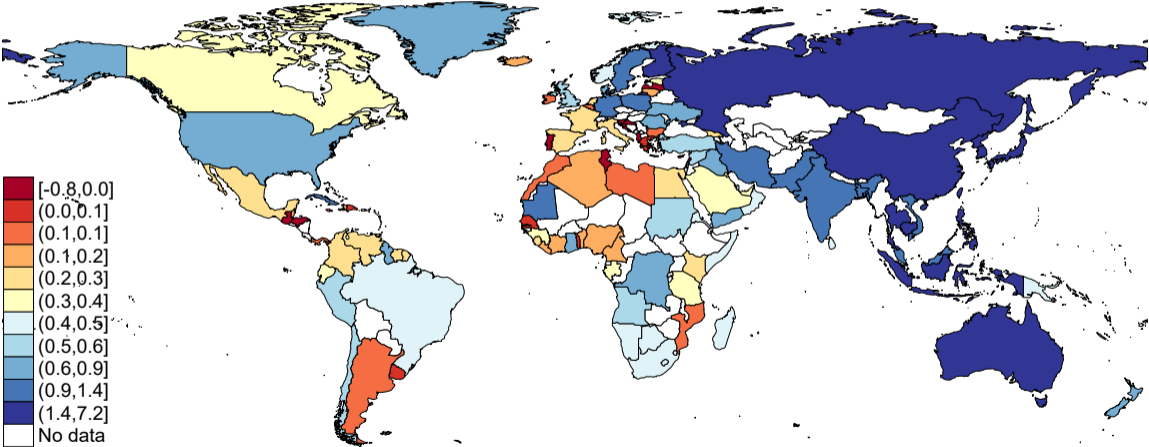
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NE Passage: No Network Effects



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 - ▶ Instrument for traffic Ξ_{kl} using

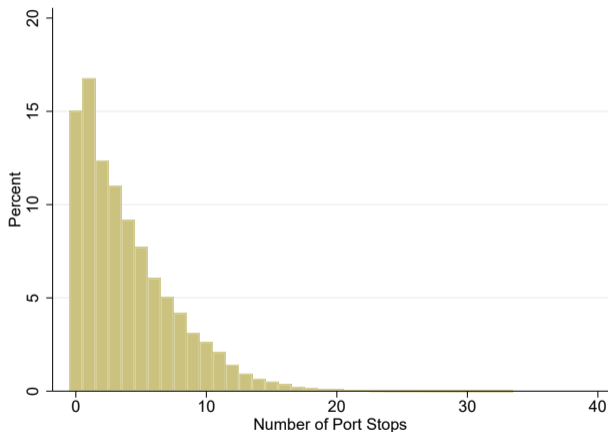
$$z_{kl} = \sum_i Pop_i \sum_j Pop_j \frac{d_{ij}}{d_{ik} d_{lj}}$$

Conclusion

- ▶ International trade is often indirect, varied, and concentrated through entrepôts
- ▶ Changes in trade costs of a node or links in the transportation network result in regional trade and welfare spillovers
- ▶ Brexit and NE passage counterfactuals: large network effects, network-localized
- ▶ Further work: integrate scale economies into counterfactuals

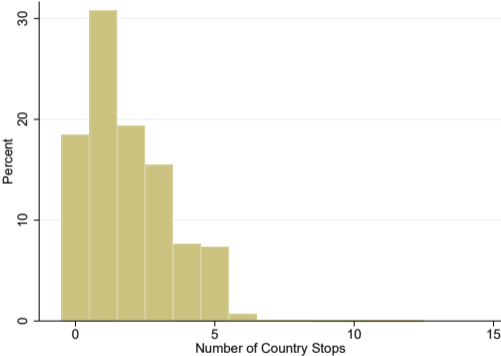
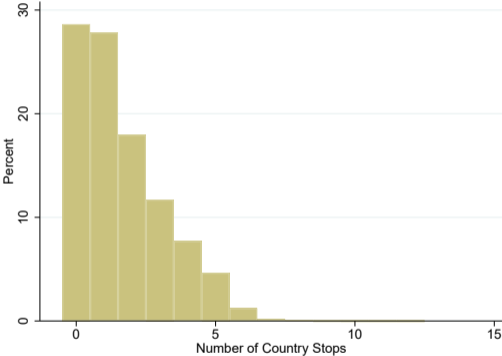
Indirectness of Trade

Number of port stops per TEU



About 15% of containers (TEUs) are direct, making no stops along the way, and the average number of port stops is 5.5 [Back](#)

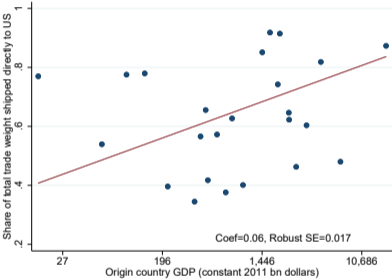
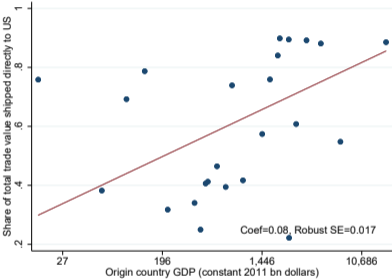
Indirectness of Trade by Weight and Value



About 70% of shipment weight and more than 80% of shipment value is indirect

Variation in Indirect Trade

Origin country GDP vs trade share at first stop



- ▶ By value and by weight, the share of direct shipments are more likely to be higher from bigger countries

Shipping: Endogenous Transport Costs (AA 2019)

- ▶ Using familiar derivations pioneered in EK (2002), express expected trade cost τ_{ij} from i to j as

$$\tau_{ij} = \mathbf{c} \left(\sum_{K=0}^{\infty} \sum_{p \in P_K} \tilde{\tau}_{ij}(\mathbf{p})^{-\theta} \right)^{-\frac{1}{\theta}} = \mathbf{c} \left(\sum_{K=0}^{\infty} \sum_{p \in P_K} \prod_{k=1}^{K_r} t_{p_{k-1}, p_k}^{-\theta} \right)^{-\frac{1}{\theta}}$$

where $\mathbf{c} \equiv \Gamma \left(\frac{\theta-1}{\theta} \right)$

- ▶ Characterize weighted adjacency matrix $\mathbf{A} = [a_{ij} \equiv t_{ij}^{-\theta}]$, $a_{ij} \in [0, 1]$ where 0 is no connection between i and j and 1 is cost-less link
- ▶ Sum over all paths of length K :

$$\tau_{ij}^{-\theta} = \mathbf{c}^{-\theta} \sum_{K=0}^{\infty} \left(\sum_{k_1=1}^N \sum_{k_2=1}^N \dots \sum_{k_{K-1}=1}^N a_{i, k_1} \times a_{k_1, k_2} \times \dots \times a_{k_{K-2}, k_{K-1}} \times a_{k_{K-1}, j} \right)$$

where k_n is sub-index for the n^{th} location reached on a particular path

Shipping: Endogenous Transport Costs (AA 2019)

- ▶ Expression in parenthesis equivalent to:

$$\tau_{ij}^{-\theta} = \mathbf{c}^{-\theta} \sum_{K=0}^{\infty} \mathbf{A}_{ij}^K$$

where $\mathbf{A}^K = [\mathbf{A}_{ij}^K]$: \mathbf{A}_{ij}^K is the (i, j) element of matrix \mathbf{A} to power K

- ▶ Express geometric sum of matrix \mathbf{A} as $\sum_{K=0}^{\infty} \mathbf{A}_{ij}^K = (\mathbf{I} - \mathbf{A})^{-1} \equiv \mathbf{B}$ where $\mathbf{B} = [b_{ij}]$ is the route cost matrix Sufficient Condition
- ▶ Write expected trade cost from i to j as function of route cost matrix:

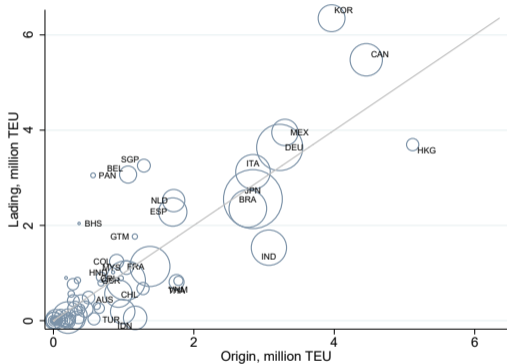
$$\tau_{ij} = \mathbf{c} b_{ij}^{-\frac{1}{\theta}}$$

which provides an analytical relationship between any given route network and the resulting bilateral trade cost between all locations

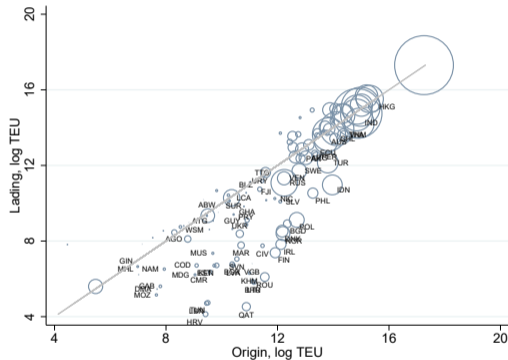
Shipping: Endogenous Transport Costs (AA 2019)

- ▶ The geometric sum of matrix \mathbf{A} is $\sum_{K=0}^{\infty} \mathbf{A}_{ij}^K = (\mathbf{I} - \mathbf{A})^{-1} \equiv \mathbf{B}$ as long as the spectral radius of \mathbf{A} is less than one
- ▶ A sufficient condition for this is if $\sum_j t_{ij}^{-\theta} < 1$ for all i
- ▶ This will necessarily be the case if either
 1. Trade costs between connected locations are sufficiently large
 2. Adjacency matrix is sufficiently sparse (i.e. many locations are not directly connected)
 3. Heterogeneity across traders are sufficiently small (i.e. θ is sufficiently large)

Concentration of Through-Shipments



(a) First stop v. trade



(b) First stop v. trade, log scale

Concentration of Through-Shipments

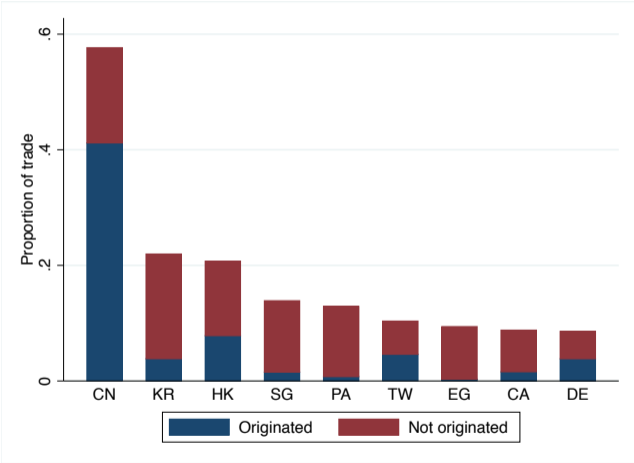
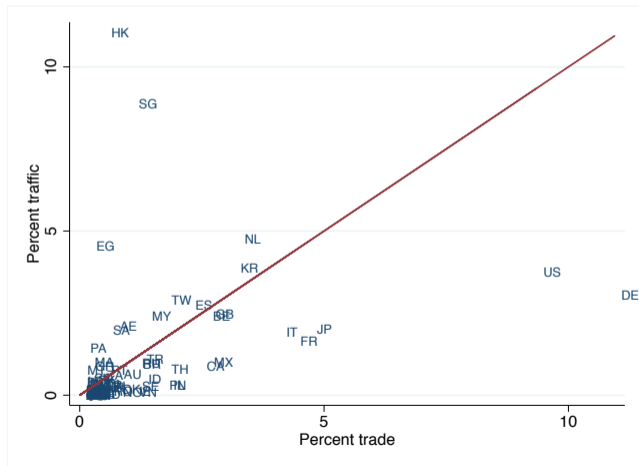


Figure: Percent of shipments making stops, by country

Concentration of Through-Shipments

Stylized Fact 4: *Shipping is concentrated through a minority of countries which account for a disproportionate share of third-party stops.*



Traffic vs. Trade

Route Cost Estimates: IV

- ▶ New problem: we don't actually observe P , but $\hat{P} = P(Q)$.

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$$\hat{P}_{kl} = \alpha_0 + \alpha_1 \cdot Q_{kl} + \epsilon_{kl} + d_{kl}$$

where $d_{kl} = P_{kl}^{true} - \hat{P}_{kl}$.

- ▶ 2 (!) exclusion restrictions: $Cov(Z_{kl}, \epsilon_{kl}) = 0$, $Cov(Z_{kl}, d_{kl}) = 0$

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- ▶ 2 (!) exclusion restrictions: $Cov(Z_{kl}, \epsilon_{kl}) = 0$, $Cov(Z_{kl}, d_{kl}) = 0$
- ▶ Two proposed fixes:

1. Test model validity to minimize scope for contamination through d_{kl} :

$$\hat{P}_{kl}^{external} = \hat{P}_{kl} + d_{kl}$$

2. Test exclusion restriction 2:

$$Cov(\hat{P}_{kl}^{external} - \hat{P}_{kl}, z_{kl}) = 0$$

Scale Estimates

Table: Scale Estimates

	(1)	(2)	(3)	(4)
	OLS	RF	FS	IV
	$\text{Log } t_{kl}^{-\theta}$	$\text{Log } t_{kl}^{-\theta}$	$\text{Log } Vol_{kl}$	$\text{Log } t_{kl}^{-\theta}$
$\text{Log } Vol_{kl}$	0.8000 (0.0108)	0.1048 (0.0222)		0.4625 (0.0549)
$\text{Log } z_{kl}$			0.2267 (0.0238)	
$\text{Log } d_{kl}$	-0.5759 -0.0244	-0.8235 (0.0579)	-0.3287 (0.0425)	-0.6714 (0.0311)
Constant	-9.3310 (0.2746)	-8.0010 (0.7400)	-1.7422 (3.2490)	-5.1450 (0.6931)
F-statistic			67	
Observations	2,284	2,284	2,284	2,284
R-squared	0.89	0.18	0.05	0.96

Note. Robust standard errors in parentheses clustered by node k