The Price of Distance: Producer Heterogeneity, Pricing to Market, and Geographic Barriers

| Kazuko Kano | Takashi Kano | Kazu Takechi |
|-------------|-----------------|--------------|
| U. of Tokyo | Hitotsubashi U. | Hosei U. |

Comments Welcome

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Price of Distance: How much do you pay to a distant market?

- Transportation costs (measured by distance) increase price differentials across regions (countries): but how much?
- Regression exercises reveal statistically significant positive effect (the Law of One Price (LOP) literature, e.g., Broda and Weinstein 2008, Engel et al. 2007, Crucini et al. 2010).
- However, the size of the distance elasticity of price differential is estimated economically subtle less than 3 % ... the price of distance is too small?
- Good transportation infrastructure?

-Introduction

Distance Effect

- Empirical trade literature observes decent size of the distance elasticity of transportation costs using data of trade volumes and trade directions approximately 20-30 % (Anderson and van Wincoop (2003)).
- Even under the same specification of iceberg-type transportation costs as in the LOP literature.
- Distance elasticity is an important parameter (Crozet and Koening 2010; Balistreri, et al, 2011)
- The LOP literature (using price data) has identification problems of distance effect

-Introduction

What's new?

- Previous reduced form LOP exercises regress retail price differentials on distance
 - Interpretation: distance includes transportation costs, distribution costs, information costs, etc
- Unique Data: Using unique price data allows us to measure transportation costs

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 Control for potential biases caused by producer heterogeneity and pricing to market

Introduction

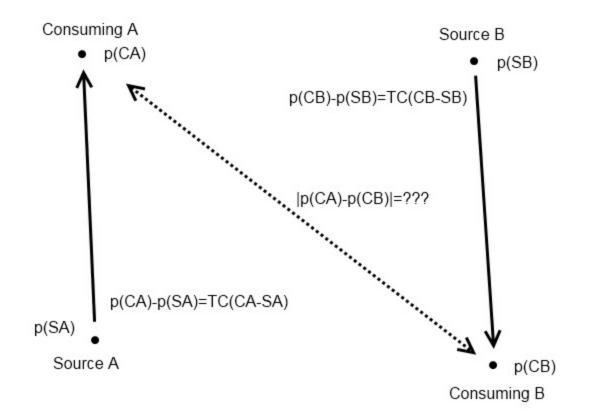
Results

- Large distance effects are found
- Elasticities = 0.46 ~ 0.768 (previous studies: 0.001 ~ 0.3)
- Price of distance is large
- Markets are not integrated yet (geographically separated)
- Implication: Improving transporation infrastructure have a large welfare impact
 - Intelligent Transport System: solve traffic jams, reduce traffic accident, ...

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What's new in this paper: data

- Focus on distance effect: regional price differentials within country (no effect of trade barriers and exchange rates (Parsely and Wei 1996))
- Unique daily data set of wholesale prices of agricultural products in Japan.
- Why unique? We can identify two crucial data aspects
 - 1. Source regions: in which regions are products made?
 - 2. **Product delivery patterns:** to which regions are products delivered from the sources?
- Why important?



What's new in this paper: source regions

- Need to know source regions of products in order to measure transportation costs correctly (Anderson and van Wincoop 2004).
- However, retail price data are not accompanied by information of the sources of products.
- Using wholesale prices and information on source regions, we can eliminate other costs associated with distance
- Donaldson (2010) and Kano et al (2010) use information about source region

Data description

- "Daily Wholesale Market Information on Fresh Fruits and Vegetables (Seikabutsu Hinmokubetsu Shikyo Joho)."
- Selected vegetables in 2007: cabbage, carrot, Chinese cabbage, lettuce, potato, shiitake mushroom, spinach, and welsh onion.
- High product categorization by sources, brands, sizes, and grades: "Identical" product shares the same brand, same size, same grade, same source, and same date.
- 55 wholesale markets across 47 prefectures in Japan: each prefecture has at least one wholesale market.
- Distances between prefectural head offices in prefectural capital cities.

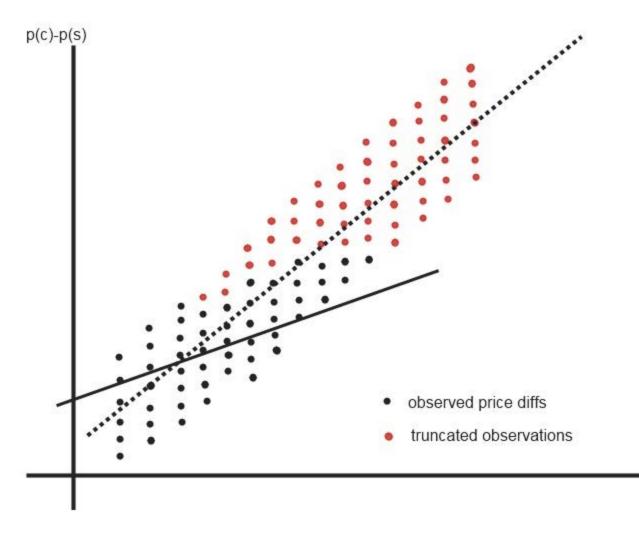
Data description

| | Cabbage | Carrot |
|--|---------|---------|
| Product entry | | |
| No. of Varieties | 3 | 10 |
| No. of size categories | 63 | 62 |
| No. of grade categories | 34 | 66 |
| No. of distinct product entries | 1027 | 1186 |
| Data truncation | | |
| No. of $T_{ij}(l) = 0$ or 1 | 369,343 | 198,129 |
| No. of $T_{ij}(l) = 1$ | 15,841 | 8,395 |
| Mean of log distance over $T_{ij}(l) = 0$ or 1 | 5.939 | 6.027 |
| Mean of log distance of $T_{ij}(l) = 1$ | 3.705 | 3.99 |
| Price differential | | |
| Mean log price differenetial $q_{ij}(l)$ | 0.039 | 0.075 |
| SD. Log price differential $q_{ji}(l)$ | 0.167 | 0.285 |

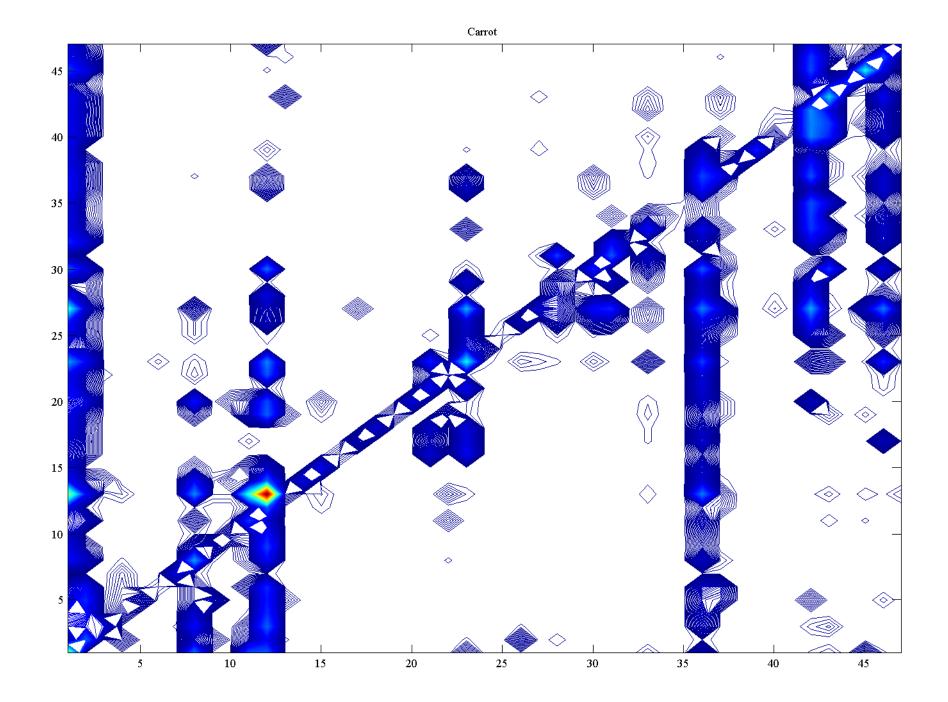
Product-delivery patterns (also done in KKT(2010))

- Two roles of transportation costs:
 - 1. intensive margin: increase price differential
 - 2. extensive margin: decrease chance of product delivery
- Transportation costs make product delivery concentrated around local areas neighboring source regions: Data truncation of price differentials.
- Estimates of distance elasticity using price data alone could be biased downwards due to sample selection.

Data truncation due to delivery choice might result in a sample selection bias.



distance



- Empirical framework

What's new in this paper: a simple heterogeneity model

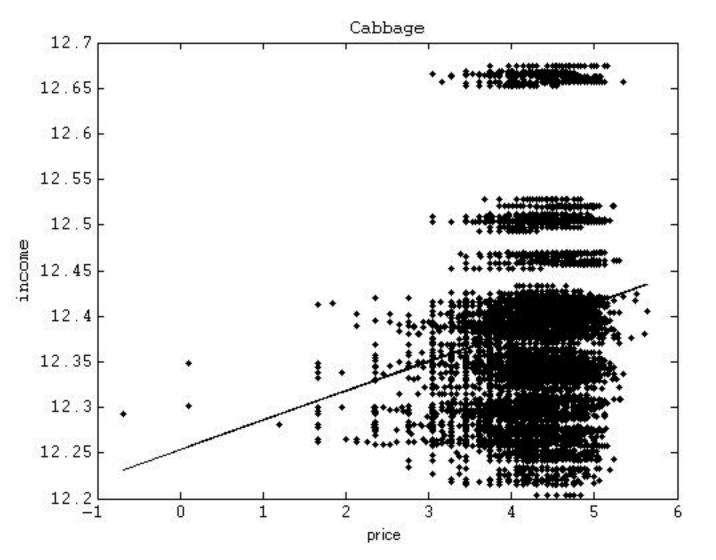
- A simple model of monopolistically competitive, heterogeneous producers facing local demand curves (Helpman, Melitz, and Rubinstein (2008)).
- They produce products in the source regions and choose which consuming regions to deliver their products with fixed costs (KKT (2010)).
- ► CES-monopolistic competition = constant markups → variable markups

- Empirical framework

Previous Literature

• Preference: non-homethetic preference \rightarrow pricing-to-market

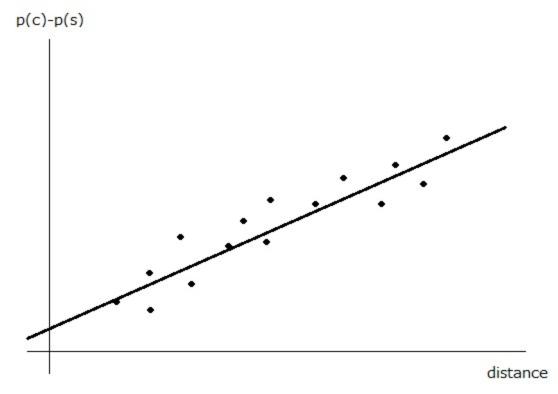
- Melitz and Ottaviano (2008)
- Simonovska (2010)
- The positive relationship between price and income per capita
 - Balassa-Samuelson
 - Variable markups
- Our focus is on the relationship between price differentials and distance... but in fact the positive relationship exits

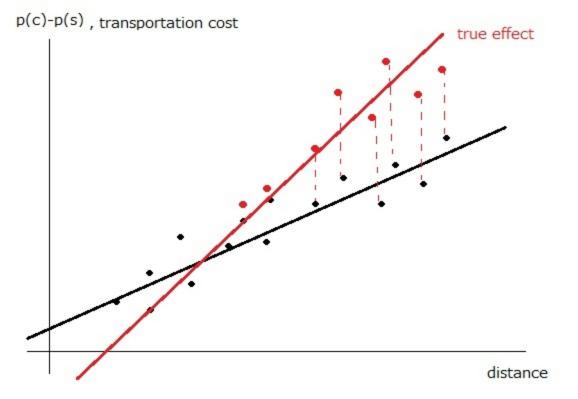


Empirical framework

Pricing Behavior and Transportation Costs

- Pricing to market → price differentials reflect local market conditions (in CES, only transportation cost)
- The relationship between price differentials and distance can be biased by the presence of producer heterogeneity and market characteristics
- Transportation costs reduce profitability in a remote market
 - As the productivity threshold increases, only highly productive and thus low-price-setting firms supply
- The increase in price differentials are relatively lower for remote markets
 - This may create biases





- Empirical framework

Non-homothetic Preference

Simonovska (2010)'s framework

$$u_i = \int_{\omega \in \Omega} \ln(q(\omega) + \bar{q}) d\omega$$

Then the demand function is:

$$q(\omega) = \frac{w_i + \bar{q} \int p(\omega)}{N_i p(\omega)} - \bar{q},$$

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- Empirical framework

Producers

- Each producer: monopolistic competition
- Producer's profit maximization problem:

$$\max \pi_{ij}(\phi) = p_{ij}q_{ij} - \frac{\tau_{ij}w_j}{\phi}q_{ij}$$

The optimal price:

$$p_{ij} = \left(\frac{\tau_{ij}w_j(w_i + \bar{q}P_i)}{\phi N_i \bar{q}}\right)^{1/2}$$

Cut-off value (zero demand):

$$\phi_i^* j = \frac{\tau_{ij} w_j N_i \bar{q}}{w_i + \bar{q} P_i} \to p_{ij} = \frac{\tau_{ij} w_j}{\phi^{1/2} \phi_{ij}^{* 1/2}}$$

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- Empirical framework

Reminder: optimal price in a CES case

• Profit =
$$p_{ij}x_{ij} - (\tau_{ij}w_j/\phi)x_{ij} - f_{ij}$$

Optimal price is given by mark-up price

$$p_{ij} = \tau_{ij} \frac{w_j}{\phi \alpha}$$

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- Empirical framework

Price Differentials

 Nonhomothetic: price depends on threshold value (heterogeneity matters)

$$p_{ij}/p_{jj} = \tau_{ij}\phi_{jj}^{*\,1/2}/\phi_{ij}^{*\,1/2} = \tau_{ij}^{1/2} \frac{(w_i + \bar{q}P_i)^{1/2}}{(w_j + \bar{q}P_j)^{1/2}} (\frac{N_j}{N_i})^{1/2}$$

CES:

$$p_{ij}/p_{ii} = \tau_{ij}$$

- In a CES framework, productivity does not affect price differential
- φ^{*}_{ij} depends on τ_{ij}: omitted variable bias (Helpman, Melitz, and Rubinstein (2008))

- Empirical framework

Profits and Heterogeneity

- Price differential is observed only when there is delivery
- Timing: pay fixed cost to draw φ, then decide to deliver after realization
- Assuming that productivity follows Patero distribution $(G(\phi) = 1 b^{\theta}/\phi^{\theta})$
- Expected profit

$$(1 - G(\phi^*)) \int \pi \mu d\phi = \frac{b^{\theta} \tau_{ij} w_i \bar{q}}{(2\theta + 1)(\theta + 1)\phi^{*\theta + 1}}$$

Delivery decision

$$Z_{ij} = (b^{\theta} \frac{\tau_{ij} w_j \bar{q}}{(2\theta + 1)(\theta + 1)\phi_{ij}^{*\,\theta + 1}}) / f_{ij} > 1$$

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Empirical framework

Profits: CES case

Gross profits/delivery costs:

$$Z_{ij} = \frac{(1-\alpha)[\frac{\tau_{ij}w_j}{\alpha P_i \phi}]^{1-\epsilon}w_i}{f_{ij}}$$

Deliver to market i:

$$Z_{ij} \ge 1$$

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- Empirical framework

Transportation Costs

- Iceberg type
- Parametric specification of transportation costs τ_{ij} with distance D_{ij}

$$\tau_{ij} = D_{ij}^{\gamma} \exp(\mu + u_{ij}), \quad u_{ij} \sim N(0, \sigma_u^2)$$

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γ is the distance elasticity parameter

- Empirical framework

Empirical Framework: sample selection

Price differential

$$\ln p_{ij} - \ln p_{jj} = (1/2)\mu + (1/2)\gamma \ln D_{ij} + (1/2)\ln(1+N_i) - (1/2)\ln(1+N_j) + (1/2)c_4dum_j - (1/2)c_5dum_i + (1/2)u_{ij}$$

Delivery decision

$$z_{ij} = -\ln f_e + \theta(\ln b - \bar{q}) - \theta\mu - \ln(2\theta + 1)(\theta + 1) + \theta\gamma \ln D_{ij} - \theta \ln w_j - (\theta + 1)\ln N_i + (\theta + 1)\ln(w_i + \bar{q}P_i) = c_0 + \theta c_1 + \theta\gamma \ln D_{ij} - \theta \ln w_i - (\theta + 1)\ln N_j + (\theta + 1)c_2 dum_j + c_3 dum_i + \eta_{ij}$$

Estimate by maximum likelihood

Empirical framework

Estimation results

| Point estimates and s.e. | Cabbage | Carrot | Lettuce |
|---------------------------|------------|------------|------------|
| $\hat{\gamma}_{non-homo}$ | 0.46 | 0.627 | 0.687 |
| (s.e.) | (0.003) | (0.006) | (0.006) |
| $\hat{	heta}$ | 2.013 | 1.169 | 1.203 |
| (s.e.) | (0.011) | (0.009) | (0.008) |
| $\hat{ ho}$ | -0.83 | -0.868 | -0.854 |
| (s.e.) | (0.003) | (0.003) | (0.003) |
| log likelihood | -19646.143 | -17034.691 | -21025.084 |
| No. of observations | 369,343 | 198,129 | 239,703 |
| | | | |
| $\hat{\gamma}_{CES}$ | 0.301 | 0.362 | 0.426 |
| $\hat{\gamma}_{OLS}$ | 0.033 | 0.051 | 0.022 |
| | | | |

- Empirical framework

Estimation Results

- Large distance effect
- Ignoring producer heterogeneity and pricing to market causes biased estimates of distance

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 The price of geographic barriers (distance) is still large for regional transportation

- Conclusions

Conclusions

- Conventional estimate of distance effects only using retail price data is heavily biased downwards due to 3 flaws:
 - 1. misspecification inevitable by no information and identification of the source regions of products.
 - 2. ignoring the underlying delivery choice that has to be affected by transportation costs too.

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- 3. pricing-to-market behavior
- After correcting these flaws, we observe the large price of distance (geographic barriers) on regional price differentials.

- Conclusions

Policy Implications

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- Geographic barriers still large
- New type of transportation system: ITS
- Automakers start doing research
- More efficient, safe road is required.