

# An Unhealthy Trade Surplus

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# Introduction

## Welfare cost of current account

- Existing view: In the absence of distortions to savings or investment, current account surplus does not produce a welfare loss
- This paper: Even without distortions to savings or investment, a current account surplus may produce a welfare loss.
- Trade surplus is usually the largest component of current account surplus. Trade surplus
  - ▶ reduces shipping cost of inbound cargos
  - ▶ increases imports of heavy goods (such as scrap metals and other industrial waste)
  - ▶ additional pollution and cause significant welfare loss, especially in countries with low pollution control standard or weak enforcement

# Motivation

- China is the largest heavy goods import country in the world

Table 1: China and World's Heavy Goods Imports

|                                   | China | World |
|-----------------------------------|-------|-------|
| Ave weight (kg) per dollar import | 0.46  | 0.22  |
| Waste imports value(mil \$)       | 1,610 | 6,450 |
| Waste imports weight(mil ton)     | 313   | 1,260 |

Note: Waste goods = "waste" or "scrap" in hs6 goods description.

- China's waste imports are about \$1 or 250kg per person per year.
- Why does China import so much heavy goods, such as scraps? Is there any consequence of this import pattern?

# What we do

- Our story:
  - ▶ Trade surplus reduces the shipping cost of relatively heavy goods - those with a high weight to value ratio.
  - ▶ Waste goods are among the heavy goods.
- Estimation suggests:
  - ▶ a 10% increase in the trade surplus is associated with an increase in heavy goods imports by 1.2% ...
  - ▶ and an increase in the waste goods imports by 4.5%
- Welfare consequence:
  - ▶ The industries which use more heavy goods are more polluting industries.
  - ▶ Trade surplus gives these industries cheaper intermediate inputs. The pollution will increase.
  - ▶ Welfare loss: 0.4%

# Literature

- Non-iceberg trade cost: Hummels and Skiba (2004), Lashkaripour (2015) and Wong (2017) and Brancaccio et al. (2017). Do not highlight the trade surplus and heavy goods shipping cost relation.
- Welfare cost of trade imbalance: Dekle et al. (2007) and Epifani and Gancia (2017). Only focus on terms of trade.
- Pollution and trade: Frankel(2009), Kellenberg (2009, 2012), Lan et al. (2012) Bombardini and Li (2016). Focus on "pollution heaven hypothesis".

# Model

- Let  $\lambda$  be the Lagrangian multiplier of the back-haul capacity (price per weight). Assume  $\lambda$  is smaller when surplus is larger
- Shipping cost per unit is

$$c_i \propto \lambda w_i$$

where  $w$  is the weight per unit of good  $i$

- The trade cost **per value** (iceberg cost) is  $\tau_i$

$$\tau_i \propto t_i + \frac{c_i}{p_i} \propto t_i + \lambda \frac{w_i}{p_i}$$

where  $t_i$  is the ad-valorem trade tariff;  $p_i$  is the price per unit.

- Hypothesis to test:
  - ▶ The trade surplus pushes the cost per weight lower
  - ▶ A country imports more heavy (per value) goods if the cost per weight is low.
  - ▶ A country tends to import more heavy goods if it runs a larger trade surplus.

# Data

- Container freight cost (64 country-pairs) + UN goods trade data (2010-2017)
  - ▶ Use the container freight cost as the proxy of  $\lambda$ . The container usually has a maximum weight.
- Weight-value ratio: Columbia custom data (2007-2013).
  - ▶ The weight-value ratio is quite persistent over time (0.91) and the correlation is about 0.8 with Chinese weight-value ratio.

## Empirical Specification: Shipping cost

- We consider the following regression equation:

$$\ln \lambda_{ndt} = \alpha_0 + \alpha_1 \ln(\text{Imbalance})_{ndt} + \Omega_{ndt}^{\leftrightarrow} + e_{ndt} \quad (1)$$

(2)

- $n$  and  $d$  = the origin and destination;  $\text{Imbalance} = \text{export}/\text{import}$  of country  $d$
- $\Omega_{ndt}^{\leftrightarrow}$  is the country pair-year FE (no direction)
- Recognizing that the trade imbalance may be endogenous, we use the gov expenditure as the IV for trade imbalance.

$$\left\{ \left( \frac{\text{Import}_{dn2000}}{\text{Import}_{d2000}} \right) \times \text{Gov}_{dt} \right\} / \left\{ \left( \frac{\text{Import}_{nd2000}}{\text{Import}_{n2000}} \right) \times \text{Gov}_{nt} \right\}, \quad (3)$$

- We will further handle the endogeneity problems using some case studies.



# Estimates for the Main Regressions: Shipping Cost

Table 2: Estimates for the Main Regressions: Shipping Cost

|   | (1)                  | (2)                   | (3)                  | (4)                 |
|---|----------------------|-----------------------|----------------------|---------------------|
|   | $\ln \lambda_{ndt}$  | $\ln \lambda_{ndt}$   | $\ln \lambda_{ndt}$  | $\ln \lambda_{ndt}$ |
| $\ln(\text{Imbalance})_{ndt}$                               | -0.071***<br>(0.024) | -0.126***<br>(0.0349) | -0.027<br>(0.025)    | -0.082**<br>(0.039) |
| $\ln(\text{Imbalance})_{ndt} \times \text{Pervasive-route}$ |                      |                       | -0.279***<br>(0.063) | -0.192**<br>(0.082) |
| Country-pair-year FE  | Y                    | Y                     | Y                    | Y                   |
| IV  |                      | Y                     |                      | Y                   |
| Obs.  | 596                  | 434                   | 596                  | 434                 |
| R-squared   | 0.82                 | 0.83                  | 0.83                 | 0.83                |

Notes: Pervasive route=1 if the destination country runs an aggregate trade surplus and the origin country runs an aggregate trade deficit.

# Empirical Specification: Import Composition

- We test whether the low shipping cost will involve heavier goods import

$$\ln I_{i,ndt} = \beta_0 \ln \lambda_{ndt} + \beta_1 \ln \lambda_{ndt} \times \ln \left( \frac{w_i}{p_i} \right) + \eta_{i,nt} + \eta_{i,dt} + \epsilon_{i,ndt}, \quad (4)$$

- $I_{i,ndt}$  is the import good  $i$  from country  $n$  to country  $d$  in year  $t$
- To address the endogeneity problem, we use the gov expenditure as the IV.

# Estimates for the Main Regressions: Log Import Value

|   | (1)                  | (2)                  | (3)                  | (4)                  | (5)                  | (6)                  |
|---|----------------------|----------------------|----------------------|----------------------|----------------------|----------------------|
| $\ln \lambda_{ndt}$   | -1.210***<br>(0.029) | -1.493***<br>(0.034) | -1.189***<br>(0.029) |                      | -1.213***<br>(0.029) |                      |
| $\ln \lambda_{ndt} \times \ln \left( \frac{w_i}{p_i} \right)$                       | -0.127***<br>(0.011) | -0.169***<br>(0.013) | -0.120***<br>(0.011) | -0.117***<br>(0.011) | -0.135***<br>(0.011) |                      |
| $\ln \lambda_{ndt} \times \ln \left( \frac{w_i}{p_i} \right) \times \text{Persist}$ |                      |                      |                      |                      | -0.020***<br>(0.001) |                      |
| $\ln(\text{Imbalance}_{ndt}) \times \ln \left( \frac{w_i}{p_i} \right)$             |                      |                      |                      |                      |                      | 0.0147***<br>(0.005) |
| O-g-y FE  | Y                    | Y                    | Y                    | Y                    | Y                    | Y                    |
| D-g-y FE  | Y                    | Y                    | Y                    | Y                    | Y                    | Y                    |
| D-o-Y FE  |                      |                      |                      | Y                    |                      | Y                    |
| IV  |                      | Y                    |                      |                      |                      |                      |
| W/o bulky goods   |                      |                      | Y                    |                      |                      |                      |
| Obs.  | 873,074              | 873,074              | 861,216              | 873,074              | 873,074              | 873,074              |
| R-squared   | 0.82                 | 0.82                 | 0.82                 | 0.84                 | 0.83                 | 0.85                 |

## Further address the endogeneity issue

- We provide two more exercises besides the IV setup to address the endogeneity issue.
- Propose two exercises:
  - ▶ Mexico currency crisis as an exogenous shock to trade imbalance but not to the heavy goods import
  - ▶ China port import data: different ports in the same country so that unobserved country-pair component can be controlled

# Evidence from Mexican Currency Crisis

- Look for an exogenous shock to the trade imbalance.
- Assuming that the currency crisis only changes the trade imbalance, but is not directly correlated with the composition of goods imports (Aguiar, 2004).
- Estimate the following equation

$$\ln \left( \frac{\text{Imports}_{i,nt}}{\text{Imports}_{i,nt-1}} \right) = \beta \ln \frac{\text{Imbalance}_{nt}}{\text{Imbalance}_{nt-1}} \times \ln \left( \frac{w_i}{p_i} \right) + \eta_{it} + \eta_{nt} + \varepsilon_{i,nt}$$

- We use a 5-years window.

## Estimates for the Mexican data

|  | (1)                  | (2)                   |
|--|----------------------|-----------------------|
|  | $\ln l_{i,nt}$       | $\Delta \ln l_{i,nt}$ |
| $\ln(\text{Imbalance})_{nt} \times \ln\left(\frac{w_i}{p_i}\right)$        | 0.0595***<br>(0.003) |                       |
| $\Delta \ln(\text{Imbalance})_{nt} \times \ln\left(\frac{w_i}{p_i}\right)$ |                      | 0.0145***<br>(0.005)  |
| Product-Year FE  | Y                    | Y                     |
| Origin-Year FE   | Y                    | Y                     |
| Obs.   | 192,356              | 133,459               |
| R-squared  | 0.50                 | 0.20                  |

# Evidence from Chinese Ports Data

- From the Chinese custom data (2000-2006), we can compute the port-country bilateral trade volumes.
- If a carrier must return to its home port, our story should hold as well (spoke-hub).
- Use Chinese ports data, estimate the following equation

$$\ln \text{Imports}_{i,mnt} = \beta \ln \text{Imbalance}_{mnt} \times \ln \left( \frac{w_i}{p_i} \right) + \eta_{imt} + \eta_{int} + \eta_{mn} + \varepsilon_{i,mnt}$$

where  $m$  is a port,  $n$  is an origin country

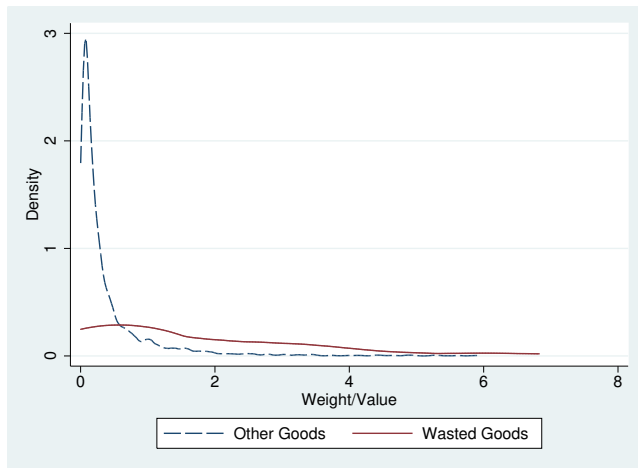
## Estimates for the Chinese Port data

|  | (1)                  | (2)                  |
|--|----------------------|----------------------|
|  | $\ln I_{i,nmt}$      | $\ln I_{i,nmt}$      |
| $\ln(\text{Imbalance})_{nmt}$  | 0.051***<br>(0.002)  |                      |
| $\ln(\text{Imbalance})_{nmt} \times \ln\left(\frac{w_i}{p_i}\right)$ | 0.0098***<br>(0.001) | 0.0057***<br>(0.001) |
| Port-Product-Year FE   | Y                    | Y                    |
| Origin-Product-Year FE   | Y                    | Y                    |
| Port-Origin-year FE  |                      | Y                    |
| Obs.   | 4,970,457            | 4,970,457            |
| R-squared  | 0.79                 | 0.81                 |



# Application 1: Waste Goods Imports

Figure 1: The Density of Weight to Value (kg/usd) Ratio



NOTE: This figure shows the density of the weight to value ratio.

# Waste Imports across Countries

Table 3: Estimates for Waste-Goods Regressions

|  | (1)                  | (2)                  | (3)                  | (4)                   | (5)                  | (6)                  |
|--|----------------------|----------------------|----------------------|-----------------------|----------------------|----------------------|
| $\ln \lambda_{ndt}$  | -0.943***<br>(0.015) | -1.206***<br>(0.017) | -1.136***<br>(0.018) | -1.484***<br>(0.029)  | -1.814***<br>(0.097) | -2.846***<br>(0.135) |
| $\ln \lambda_{ndt} \times \text{Waste}$                          | -0.454***<br>(0.179) | -0.298<br>(0.179)    | -0.718***<br>(0.203) | -0.511*<br>(0.304)    | -0.452**<br>(0.179)  | -0.720***<br>(0.205) |
| $\ln \lambda_{ndt} \times \ln \left( \frac{w_i}{p_i} \right)$    |                      | -0.125***<br>(0.011) |                      | -0.166***<br>(0.0133) |                      |                      |
| $\ln \lambda_{ndt} \times \text{Waste} \times \ln(\text{GDP}_d)$ |                      |                      |                      |                       | 0.090***<br>(0.010)  | 0.174***<br>(0.014)  |
| O-g-y  | Y                    | Y                    | Y                    | Y                     | Y                    | Y                    |
| D-g-y  | Y                    | Y                    | Y                    | Y                     | Y                    | Y                    |
| IV   |                      |                      | Y                    | Y                     |                      | Y                    |
| Obs.   | 870,119              | 870,044              | 870,119              | 870,044               | 870,119              | 870,119              |
| R-squared  | 0.82                 | 0.82                 | 0.82                 | 0.82                  | 0.82                 | 0.82                 |

## Application 2: Trade Surplus and Pollution

- Construct the weight per input value using the China 2012 I-O table.
- The pollution intensity per dollar value of output for each industry is obtained from the data by the World Bank's Industrial Projection System (IPPS).

**Table 4:** The Correlation Between Pollution Intensity and Input's Weight across Chinese Industries

|                        | $\ln(\text{SO}_2)$ | $\ln(\text{NO}_2)$ | $\ln(\text{TSP})$ |
|------------------------|--------------------|--------------------|-------------------|
| Weight-per-input-value | 2.424**<br>(1.273) | 1.845*<br>(1.128)  | 1.974*<br>(1.178) |
| Obs.                   | 74                 | 74                 | 74                |
| R-square               | 0.05               | 0.04               | 0.04              |

Notes: This table shows the correlation between three pollution intensities and the weight-per-input-value across Chinese industries. \*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ .

# Polluting Sectors and Trade Surplus

- We want to test whether the increase of trade surplus would make polluting industries dis-proportionally expand through our mechanism.
- We take China as an example.

$$\ln Y_{i,t} = \beta_0 \ln imbalance_t * PollutingSector_i + \epsilon_{i,t}, \quad (5)$$

where  $i$  is a 4-digit industry and  $t$  is from 1999-2017.

- Use Chinese biggest 3 trading partners' gov expenditures (Japan, Korea and US) as an IV for trade imbalance.
- $Y_{i,t}$  is the domestic sales of industry  $i$  in year  $t$ .

Table 5: Estimates for the Industry Output and Trade Imbalance

|  | (1)                        | (2)                        | (3)                        |
|--|----------------------------|----------------------------|----------------------------|
|  | $\ln(\text{Output}_{i,t})$ | $\ln(\text{Output}_{i,t})$ | $\ln(\text{Output}_{i,t})$ |
| $\ln(\text{Imbalance}_t) \times$<br>Heavy-sector <sub><i>i</i></sub>     |                            |                            | 0.921**<br>(0.374)         |
| $\ln(\text{Imbalance}_t) \times$<br>Polluting-sector <sub><i>i</i></sub> | 0.905***<br>(0.421)        | 0.983**<br>(0.500)         | 0.666<br>(0.410)           |
| Year FE  | Y                          | Y                          | Y                          |
| Industry FE  | Y                          | Y                          | Y                          |
| IV   |                            | Y                          |                            |
| Obs.   | 6,630                      | 6,630                      | 6,630                      |
| R-square   | 0.98                       | 0.98                       | 0.98                       |

Notes:  $\text{Output}_i$  means a Chinese industry domestic output in year  $t$ .  $\text{Imbalance}_t$  is Chinese trade imbalance measured by export-import ratio in year  $t$ . Heavy-sector<sub>*i*</sub> and Polluting-sector<sub>*i*</sub> are dummy variables defined in section ???. We use the government

# Discussion

- If the gov can collect tax to correct the pollution effect from the trade surplus, there is no welfare loss.
- However, we find that there is no correlation between the environmental regulation and trade surplus or heavy goods import.

# Welfare loss of trade surplus: A quantitative analysis

## Consumer

- A small open economy; two periods; complete capital market
- Final consumption comes from two sectors output: polluting sector  $q_t$  and output from the green sector  $y_t$ .
- Three trade-able materials: heavy material (H), light material(M) and scrap (K)
- One period utility:  $\ln c - \eta x$ ,  $c$ =consumption,  $x$  =pollution emission

$$c_t = \alpha^{-\alpha} (1 - \alpha)^{-(1-\alpha)} y_t^\alpha q_t^{1-\alpha} \quad (6)$$

- $\phi$  share of final consumption converts to scrap goods

$$K_t = \phi c_t \quad (7)$$

- Consumer are endowed with H and M every period.

# Firm

- The green sector uses labor and light materials to produce

$$y_t = A_{y,t} \left( m_t^\omega m_t^{*(1-\omega)} \right)^\theta L_{y,t}^{1-\theta}$$

- All start variables denote the imports from foreign countries.
- The polluting sector uses scrap, heavy material and labor to produce. One unit of output will generate  $b$  ton pollutants.

$$q_t = A_{q,t} \left( h_t^\beta h_t^{*(1-\beta)} \right)^\sigma \left( k_t^\gamma k_t^{*(1-\gamma)} \right)^\lambda L_{q,t}^{1-\sigma-\lambda}$$

- The pollution abatement cost is  $\frac{\xi}{2} \bar{x}_t^2$ ,  $\bar{x}_t$  is the reduction of pollution
- Pollution emission  $x_t = by_{gt} - \bar{x}_t$
- Pollution tax  $T_t x_t$



# Trade Cost

- The trade cost of light material is assumed to be exogenous.
- The trade cost of the heavy and scrap are endogenously determined by the trade surplus.

$$\tau_{h,t}^* = \bar{\tau}_{h,t}^* \left(1 + \frac{S_t}{w_t}\right)^{-\nu} I(S_t > 0)$$

$$\tau_{k,t}^* = \bar{\tau}_{k,t}^* \left(1 + \frac{S_t}{w_t}\right)^{-\nu} I(S_t > 0)$$

# Calibration

Assume  $T = 0$ . Calibrate the model to match period 1 with 2012 Chinese economy.

Table 6: Parameters

| Parameters | Value | Moments                            | Model | Data  |
|------------|-------|------------------------------------|-------|-------|
| $L$        | 0.897 | wage per capita                    | 1     | 1     |
| $\rho$     | 1.021 | surplus per capita/w per capita    | 0.045 | 0.045 |
| $\omega$   | 0.659 | light import/total expenditure     | 0.092 | 0.092 |
| $\sigma$   | 0.461 | labor share in polluting industry  | 0.52  | 0.52  |
| $\lambda$  | 0.019 | scrap import/total expenditure     | 0.005 | 0.005 |
| $\beta$    | 0.333 | heavy import/total expenditure     | 0.123 | 0.123 |
| $M$        | 0.616 | light export/total expenditure     | 0.13  | 0.13  |
| $H$        | 0.275 | heavy export/total expenditure     | 0.117 | 0.117 |
| $\phi$     | 0.031 | scrap export/total expenditure     | 0     | 0     |
| $b$        | 2.73  | SO2 ton emission/total expenditure | 0.001 | 0.001 |
| $\xi$      | 0.56  | SO2 ton trade price                | 0.46  | 0.46  |

# Results

Table 7: Calibration Results

|                           | (1)  | (2)    | (3)    | (4)    | (5)    |
|---------------------------|------|--------|--------|--------|--------|
| Pollution                 | 100  | 97.90  | 74.00  | 97.57  | 49.83  |
| Pollution intensity       | 100  | 98.88  | 85.26  | 98.71  | 68.99  |
| Scrap imp t=1             | 100  | 96.92  | 0      | 0      | 50.62  |
| Heavy imp t=1             | 100  | 96.92  | 99.97  | 98.93  | 60.84  |
| Surplus/w %               | 4.50 | 4.50   | 4.51   | 5.38   | 3.31   |
| Pc t=1                    | 100  | 100.40 | 105.81 | 100.10 | 141.39 |
| Utility change from c (%) | 0    | -4.51  | -9.99  | -4.61  | -9.20  |
| Utility change (%)        | 0    | 0.40   | 1.14   | 0.34   | 8.25   |

Notes: This table shows the model predictions in different counter-factual studies: (1) Benchmark model, (2) Symmetric trade cost ( $\nu = 0$ ), (3) Banning scrap import, (4) Banning scrap import + high elasticity of substitution ( $\omega_k = 10$ ) (5) Optimal environmental regulation.

# Conclusion

- This is the first paper that explores how a trade imbalance may affect welfare by altering the composition of imports
- Consistent with our theory, we find that countries with a trade surplus tends to import more heavy goods, especially industrial waste.
- This suggests a concrete channel for a trade surplus to generate a welfare loss, especially in developing countries that have a lax environmental standard.

# Most(Least) 10 Heavy Goods

Table 8: 10 Most Heavy and Light Goods

| Heavy Goods             | Light Goods             |
|-------------------------|-------------------------|
| Bitumen and asphalt     | Diamond                 |
| Limestone flux          | Precious metal          |
| Wasted slag from iron   | Gold                    |
| Ceramic building bricks | Halogenated derivatives |
| Glass scrap             | Watch                   |
| Pebbles and flint       | Emeralds                |
| Sea-iron products       | Gold powder             |
| Smelting steel waste    | Cocaine esters          |
| Tungsten-bearing waste  | Lysergic acid           |
| Gypsum                  | Alkaloids               |

Note: This table shows 10 goods which have the highest (lowest) weight-to-value ratio from the transaction level data on Colombian imports in 2007-2013 period.

## Estimates for the Main Regressions: Shipping Cost (cont')

- Concern: Multinational shipment arrangement by shipping companies may mitigate the negative relationship between trade imbalance and the shipping cost differential.
- First, negative elasticity of trade imbalance suggests that this concern is not strong enough to over turn the sign.
- Second, if a country runs trade surplus against most of its trade partners (like China), it is difficult to find such a multinational shipment arrangement to full load the empty vessel.

# Estimates for the Main Regressions: Shipping Cost (cont')

|  | (1)                 | (2)                 | (3)                 | (4)                 |
|--|---------------------|---------------------|---------------------|---------------------|
|  | $\ln \lambda_{ndt}$ | $\ln \lambda_{ndt}$ | $\ln \lambda_{ndt}$ | $\ln \lambda_{ndt}$ |
| $\ln(\text{Imbalance})_{ndt}$                        | -0.088<br>(0.06)    | -0.141**<br>(0.067) | -0.024<br>(0.073)   | -0.042<br>(0.084)   |
| $\ln(\text{Imbalance})_{ndt} \times \text{Sol.pair}$ |                     |                     | -0.188<br>(0.125)   | -0.237*<br>(0.136)  |
| Country-pair FE                                      | Y                   | Y                   | Y                   | Y                   |
| IV   |                     | Y                   |                     | Y                   |
| Obs.   | 84                  | 84                  | 84                  | 84                  |
| R-squared  | 0.77                | 0.76                | 0.78                | 0.78                |

## City-level evidence: Pollutant emission (cont')

| VARIABLES                      | (1)<br>ln(water pollutant) | (2)<br>ln(so2)      | (3)<br>ln(dust)   |
|--------------------------------|----------------------------|---------------------|-------------------|
| ln(textile waste import)       | 0.543**<br>(0.237)         | 0.919***<br>(0.299) | 0.382<br>(0.321)  |
| ln(metal waste import)         | 0.407*<br>(0.217)          | -0.0717<br>(0.273)  | -0.164<br>(0.294) |
| ln(plastic waste import)       | 0.0328<br>(0.224)          | 0.0495<br>(0.283)   | 0.500<br>(0.304)  |
| GDP, Manufacture Y, Population | Y                          | Y                   | Y                 |
| Year FE                        | Y                          | Y                   | Y                 |
| City FE                        | Y                          | Y                   | Y                 |
| Observations                   | 994                        | 993                 | 994               |
| R-squared                      | 0.966                      | 0.954               | 0.947             |

All coefficients are scaled by 0.01.



# Waste Imports and Pollution

- Lots of scientific evidence: Waste goods recycling might hurt people's health, e.g., increasing the cancer rate.
- Is the pollution externality generated by waste import properly corrected via government regulation?
  - ▶ OECD environmental regulation report (2002-2004)

# Results of Regulation and Trade Imbalance

Table 9: Results of Regulation and Waste Imports

|                    | (1)<br>ERS          | (2)<br>Environment<br>tax | (3)<br>Regulation<br>Standard |
|--------------------|---------------------|---------------------------|-------------------------------|
| ln(Waste Import)   | -0.011<br>(0.013)   | -0.015<br>(0.042)         | -0.017<br>(0.012)             |
| ln <i>GDP</i>      | -0.199<br>(1.311)   | -5.089<br>(4.243)         | 5.351***<br>(1.184)           |
| Corruption         | -0.727**<br>(0.339) | -1.017<br>(1.069)         | -0.098<br>(0.306)             |
| Regulation Quality | 0.187<br>(0.279)    | -1.005<br>(0.901)         | 0.451*<br>(0.252)             |
| Country FE         | Y                   | Y                         | Y                             |
| Year FE            | Y                   | Y                         | Y                             |
| Obs.               | 93                  | 96                        | 93                            |
| R-squared          | 0.94                | 0.85                      | 0.96                          |