

# Protectionism, Evasion, and Household Welfare: Evidence from Nigeria's Import Bans

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Banca d'Italia, European Central Bank and World Bank Conference  
Trade, value chains and financial linkages in the global economy

# Motivation

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(Atkin and Khandelwal, 2020; Irwin 2019; Olken and Pande 2012)

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than developed countries
- ▶ **They also face greater challenges enforcing trade policy**  
(Atkin and Khandelwal, 2020; Irwin 2019; Olken and Pande 2012)
- ▶ **How does enforcement shape the welfare impact of trade policy?**

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- ▶ Simple model calibrated using the 2018/2019 Nigerian Living Standards Survey

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- ▶ **Weak trade policy enforcement**
- ▶ **High quality price and household survey data**

# Preview of Main Findings

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- ▶ **Import bans are not effectively enforced:**
  - ▶ Bans trigger **increases in evasion**.
- ▶ **Ban imposition increases prices by 9.9% on average, but inflation is attenuated by evasion**
  - ▶ Higher increases for goods with lowest evadability
- ▶ **Bans disproportionately hurt the rich**
  - ▶ Poorer households derive a larger share of their income from banned goods
  - ▶ Enhanced enforcement would disproportionately hurt the rich:  
⇒ **evasion is welfare-enhancing but regressive**

# Outline Rest of the Talk

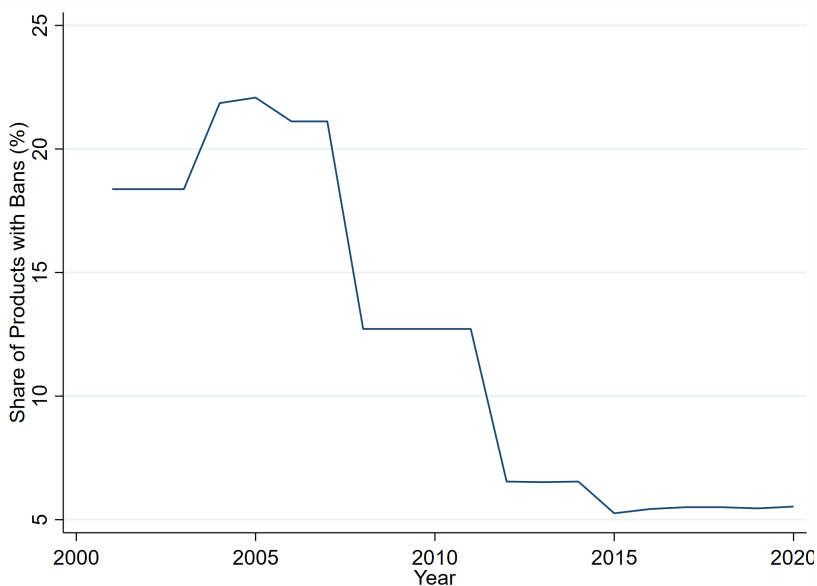
1. Introduction
2. Data and context
  - ▶ Data
  - ▶ Context
  - ▶ Measuring evasion
3. Were bans effectively enforced?
4. Did bans impact prices?
  - ▶ Average impacts
  - ▶ Measuring evadability
  - ▶ Did price impacts vary with ease of evasion?
5. Distributional impacts
6. Conclusion

# Data and Context

# Data

- ▶ **Import Bans and Taxes (tariffs, levies, VAT)** (HS6 level)  
*(Law firm)*
- ▶ **Trade Flows** (*COMTRADE*)
- ▶ **CPI Micro Data** (*Nigerian Bureau of Statistics*)
- ▶ **2018-2019 Nigeria Living Standards Survey**
  - ▶ 22,110 households, nationally representative

# Evolution of Import Bans in Nigeria, 2001-2020



# Measuring Evasion

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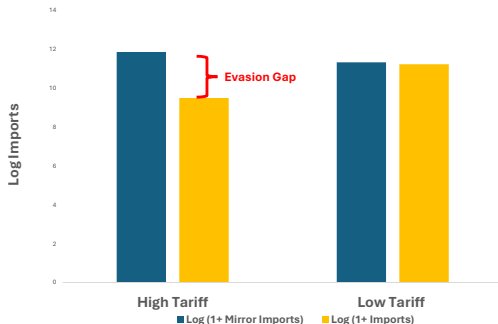
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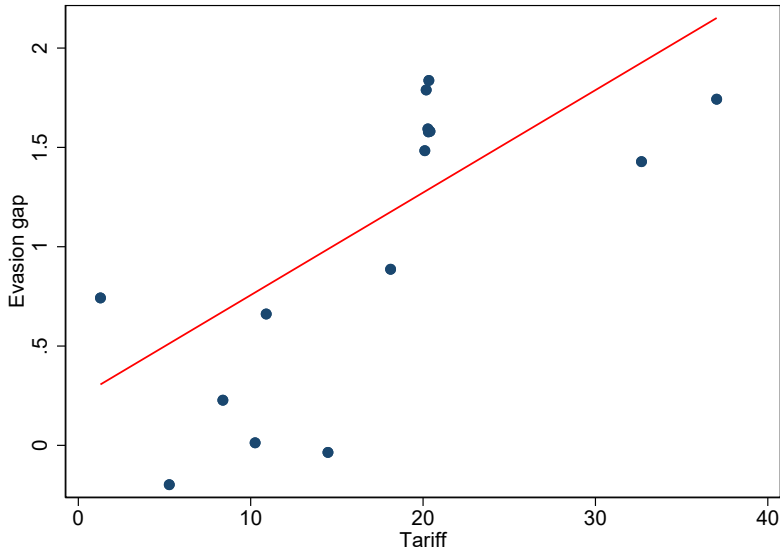
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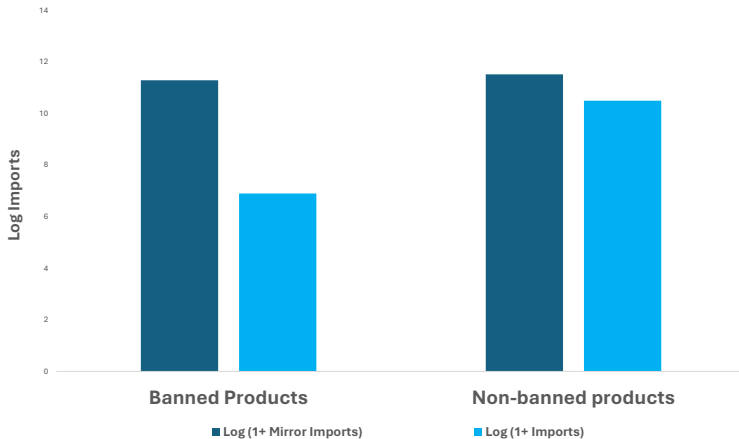
# Evasion Gaps Increase with Tariffs

Total revenue loss: 33%



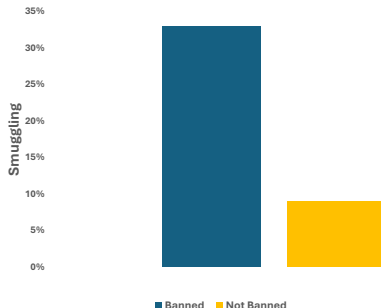
# Were Bans Effectively Enforced?

# Were Import Bans Effectively Enforced?



Evasion gaps are higher for banned products

# Lost Exports and Import Bans



Banned products are more likely to be smuggled

- ▶ Lost exports: Mirror imports  $> 0$  and formal imports  $= 0$

# Identification challenges

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  - ▶ TWFE estimates can be biased due to
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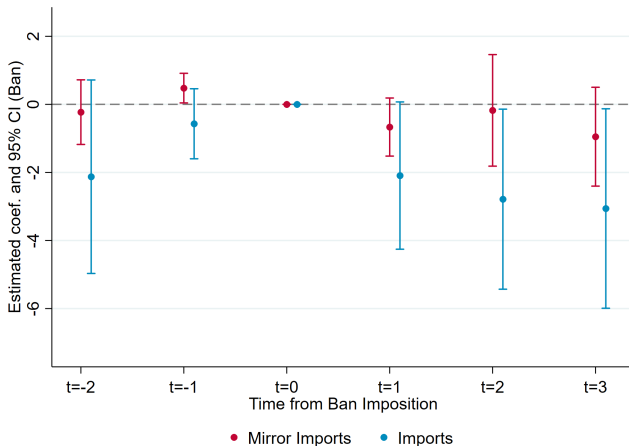
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- ▶ Separately estimate impact of ban imposition and ban removals

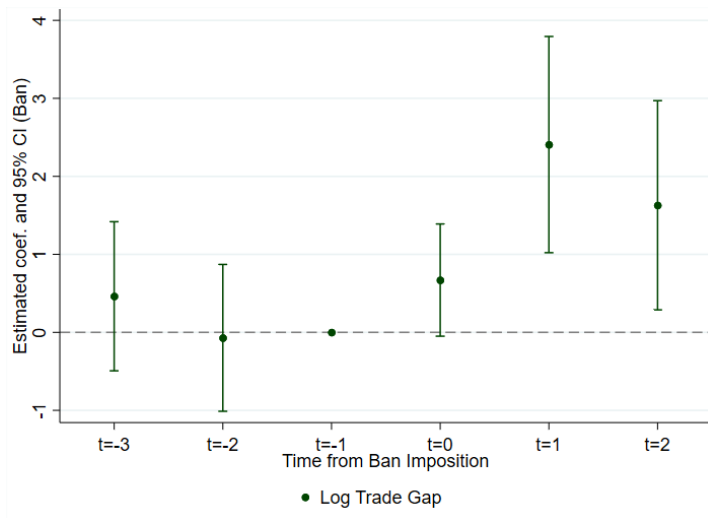
# Impact of ban impositions on trade

Imports fall more precipitously than mirror imports



Notes: Figure shows the results of a staggered PPML estimator (Moreau-Kasltler, 2025) of mirror imports and imports on ban imposition controlling for  $(1 - \text{Ban}) \times \text{Taxes}$ .

# Import bans trigger evasion



Notes: Figure shows the results of a local projection difference-in-difference estimation (Dube et al. 2025) of the log trade gap on ban imposition controlling for  $(1 - \text{Ban}) \times \text{Taxes}$ .

# Did Bans Trigger Price Increases?

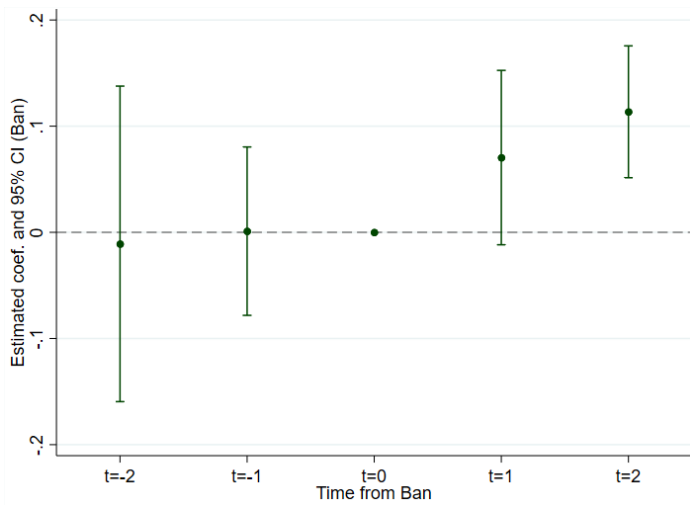
# How did bans impact prices?

## Econometric Strategy

- ▶ Use same strategy as for trade
  - ▶ use local-projections difference-in-difference estimator (Dube et al., 2025)
- ▶ First examine impacts on average prices, then assess heterogeneity by susceptibility to evasion

# Impact of ban imposition on prices

Prices increase by 9.9% when bans are imposed



Notes: Figure shows the results of a local projection difference-in-difference estimation (Dube et al. 2025) controlling for  $(1 - \text{Ban}) \times \text{Taxes}$ .



# Measuring Evadability

# Evadability (“Ease of Evasion”)

- ▶ Use mirror statistics for other countries:
  - ▶ ECOWAS countries not bordering Nigeria nor source of informal flows

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  - ▶  $Tariff_{pist}$ : Bilateral tariff
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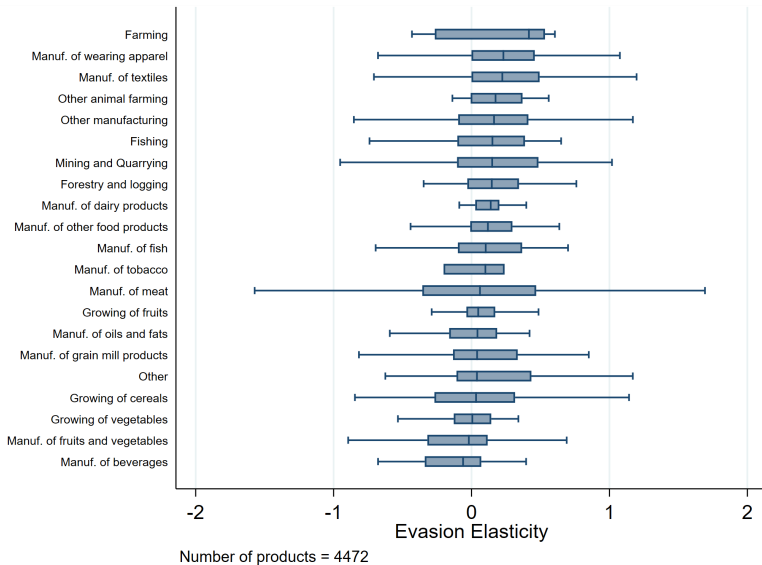
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- ▶  $Ev_p$  measures the “ease of evasion” or “evadability” :
  - ▶ Intuition: how responsive is underreporting to an increase in tariffs?
  - ▶ Arguably exogenous to evasion in Nigeria

# Ease of evasion estimates



# Alternative Proxies for Ease of Evasion

- ▶ **Value relative to transport costs:**

- ▶ Using COMTRADE, we calculate value to transport ratio

$$r = \frac{\text{FOB}}{\text{CIF} - \text{FOB}}$$

sample restricted to trade with the US

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Transport Costs

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Transport Costs

## ▶ **Similarity (ease of misclassification):**

- ▶ Using a pre-trained SBERT model, we compute cosine similarity between product descriptions.
- ▶ Similarity: average similarity among the top quartile of most similar, non-banned products
- ▶ Greater similarity  $\implies$  greater ease of evasion

Similarity

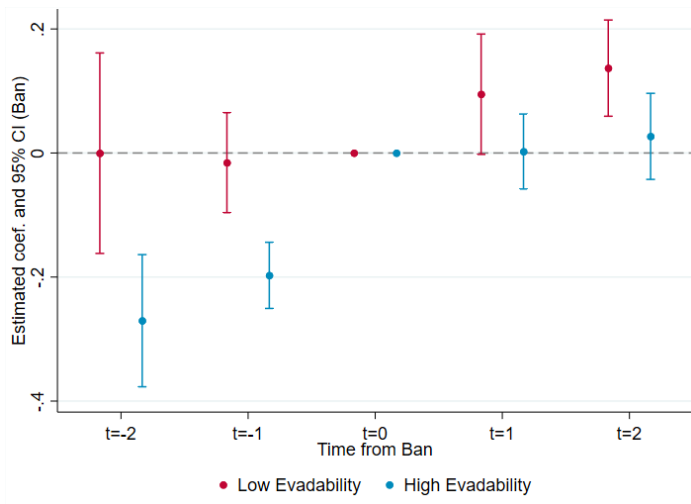
Validation of evadability measures

# Did Price Impact Vary with Susceptibility to Evasion?



# Heterogeneous impact of bans on prices

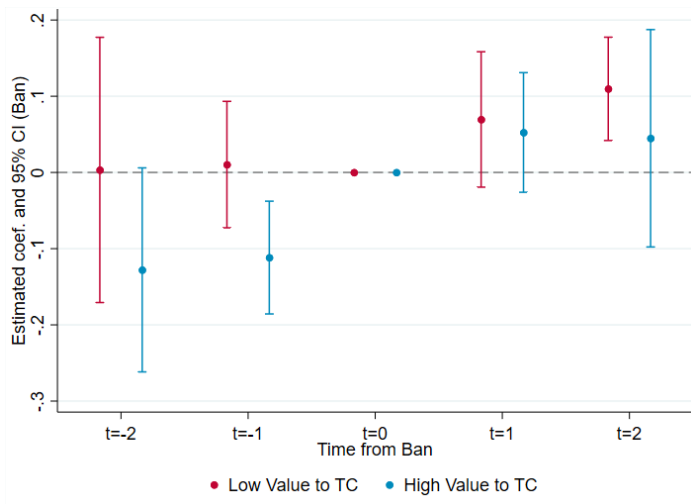
Higher evadability  $\rightarrow$  lower increase in prices



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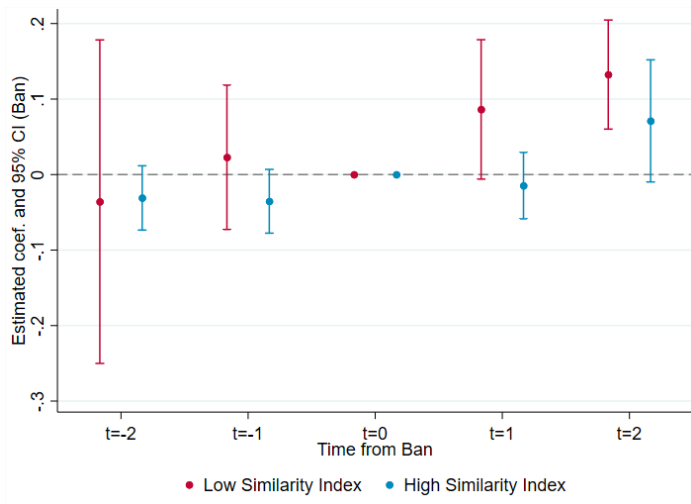
Lower transport costs (=high value to TC)  $\rightarrow$  lower increase in prices



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# Heterogeneous impact of bans on prices

Greater similarity to other products  $\rightarrow$  lower increase in prices



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# Distributional Impacts

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- ▶ Stylized model of households as consumers and producers  
(Deaton, 1989; Artuc Porto and Rijkers, 2017)
  - ▶ Households make consumption and income decisions
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  - ▶ Price changes: estimated
- ▶ Simulations of enforcement changes:
  - ▶ Enforcement conceptualized as changes in evadability

# Welfare impacts of Bans

- First-order welfare effect of a change in the price of  $i$

$$\underbrace{\frac{dV_i^h}{y^h}}_{\text{Welfare effect}} = \left( \underbrace{\phi_i^h}_{\text{Income share of } i} - \underbrace{s_i^h}_{\text{Budget share of } i} \right) \underbrace{d \ln p_i}_{\text{Price change of } i}$$



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- ▶ Limitations
  - ▶ Only captures the direct short-run welfare impacts
  - ▶ No impacts on wages and family enterprise income
  - ▶ No adjustment of consumption and production choices

# Aggregated Effect

- ▶ Aggregate welfare changes across traded goods  $i$  using:

$$\hat{V}^h = \sum_i \left( \phi_i^h - s_i^h \right) d \ln p_i,$$

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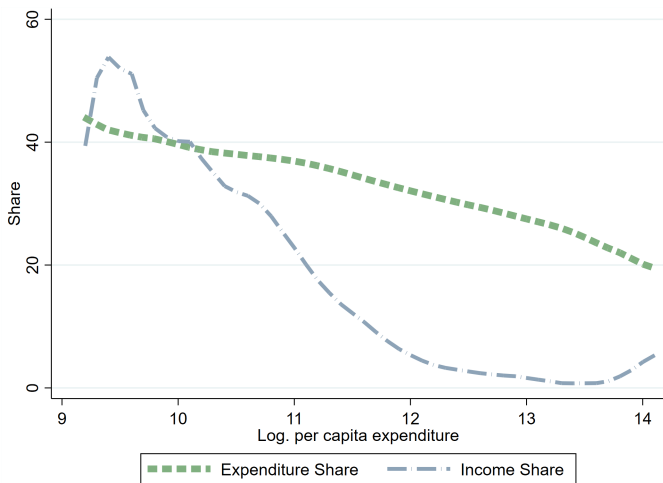
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- ▶ Welfare impacts of import bans and evasion evaluated as:
  1. Baseline: estimated price changes
  2. Counterfactual: reduced ease of evasion ( $Ev$ )
    - ▶ Assume each product falls in the "low evadability" category

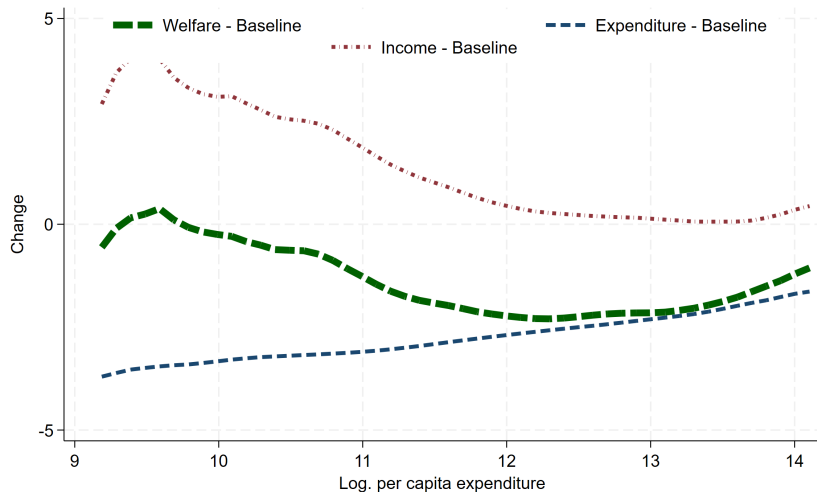
# Income and expenditure shares

Poor household spend more on banned products, but also generate more income from them



# Welfare effects of import bans

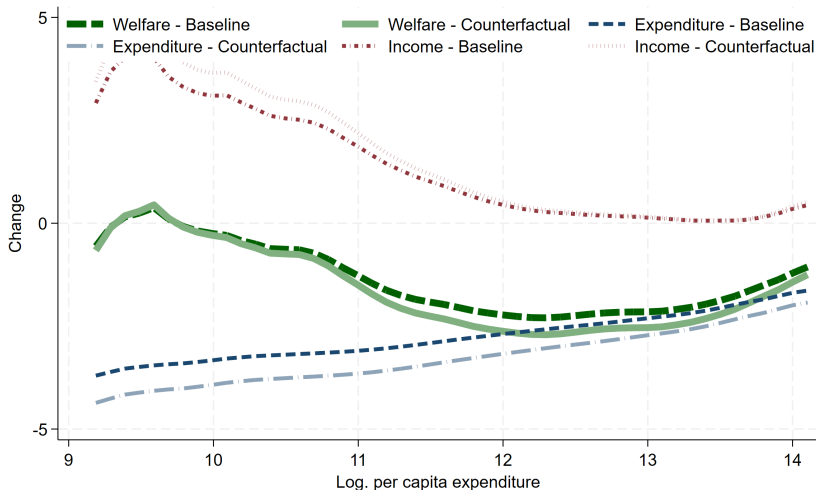
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- ▶ **Bans fuel inflation, the impact of which is softened by evasion**
  - ▶ Average impact of ban imposition 9.9%
  - ▶ Price response attenuated for goods which are easy to evade
- ▶ **Bans repress real income, especially of rich households**
  - ⇒ evasion is welfare enhancing, and pro-rich

# Examples of Banned Products

## **Meat & Poultry**

Live/Dead Birds, Frozen Poultry, Pork, Beef, Bird Eggs (excl. hatching)

## **Food & Beverages**

Refined Oils/Fats, Sugar, Cocoa, Spaghetti/Noodles, Juice, Waters, Beer, Stout

## **Household Items**

Bagged Cement, Soaps/Detergents, Mosquito Coils, Paper, Cartons, Toilet Paper, Exercise Books, Footwear, Bags, Carpets/Textiles

## **Pharma & Medications**

Tablets/Syrups (Paracetamol, Aspirin, etc.), Multivitamins, Haematinics, Ointments, IV Fluids, Waste Pharmaceuticals

## **Used Equipment & Vehicles**

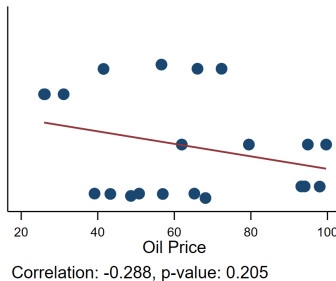
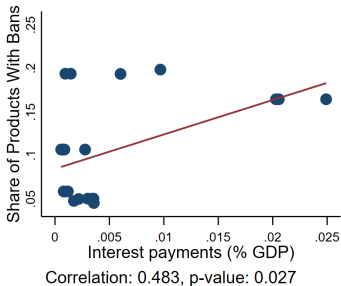
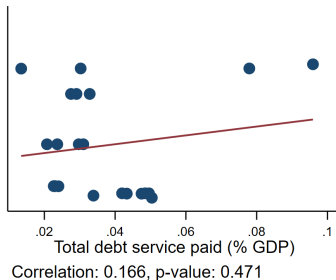
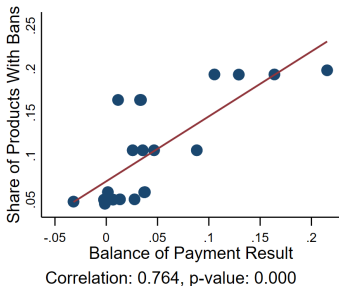
Compressors, AC, Fridges/Freezers, Motor Vehicles > 12 yrs, Rethreaded/Used Tyres

## **Other Items**

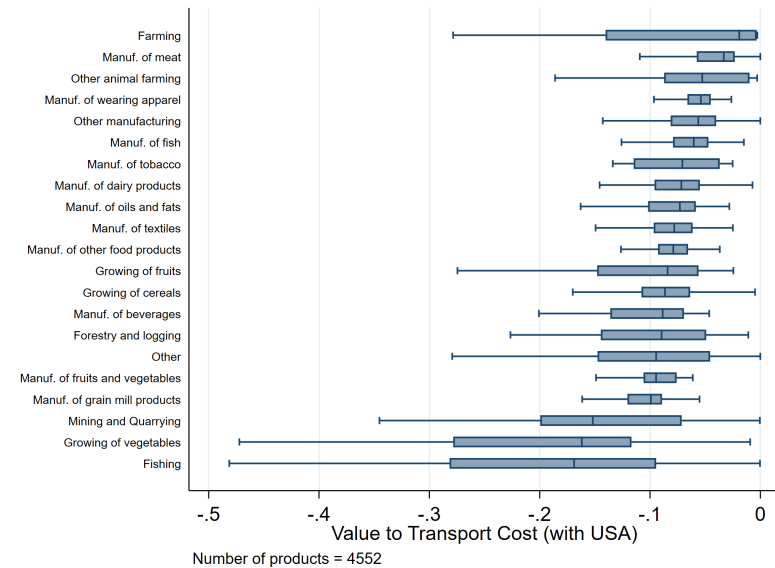
Telephone Recharge Cards, Ballpoint Pens, Parts, Hollow Glass Bottles > 150ml

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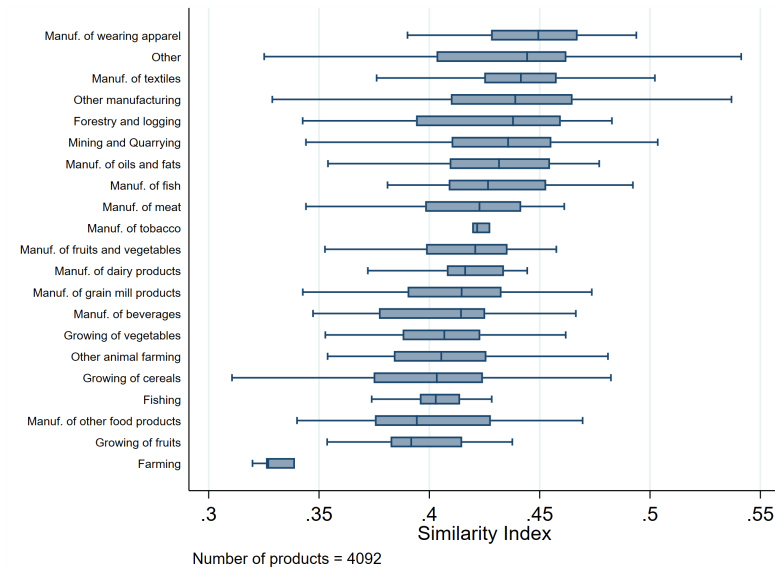
# Prevalence of Bans and Macroeconomic Conditions



# Transport Costs



# Similarity scores





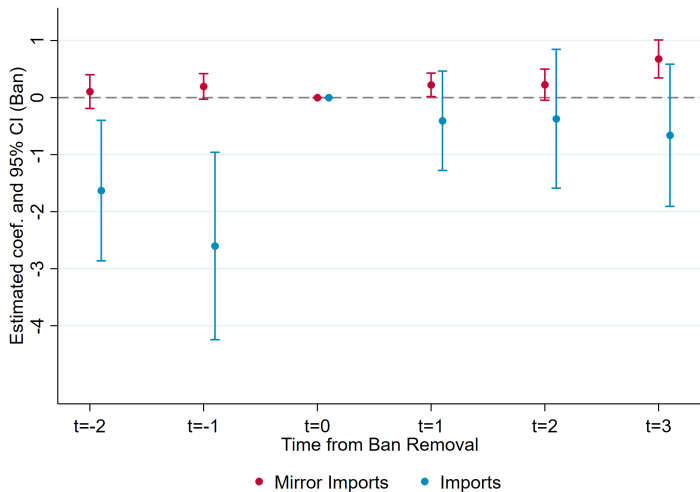
# Validation of Evasion Measures

Explaining evasion gaps in other countries

	Log(Trade Gap + 1)					
	(1)	(2)	(3)	(4)	(5)	(6)
Log(Tariffs +1)		0.065*** (0.011)		0.067*** (0.011)		0.067*** (0.011)
Evadability	0.016** (0.008)	0.015** (0.008)				
Value/Transport			0.143 (0.208)	0.399* (0.211)		
Similarity					0.592** (0.248)	0.813*** (0.258)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes
Source FE	Yes	Yes	Yes	Yes	Yes	Yes
Importer FE	Yes	Yes	Yes	Yes	Yes	Yes
Obs	1,274,080	963,507	1,256,133	945,746	1,134,676	854,904
R <sup>2</sup>	0.066	0.069	0.066	0.069	0.066	0.068

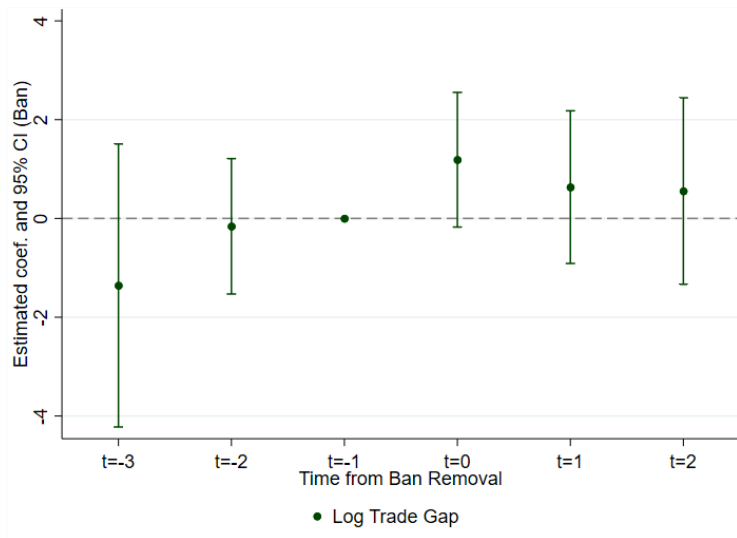
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# Impact of Ban Removal on Trade



Notes: Figure shows the results of a staggered PPML difference-in-difference estimation on mirror imports and imports (Moreau-Kastler et al. 2025) controlling for  $(1 - \text{Ban}) \times \text{Taxes}$ .

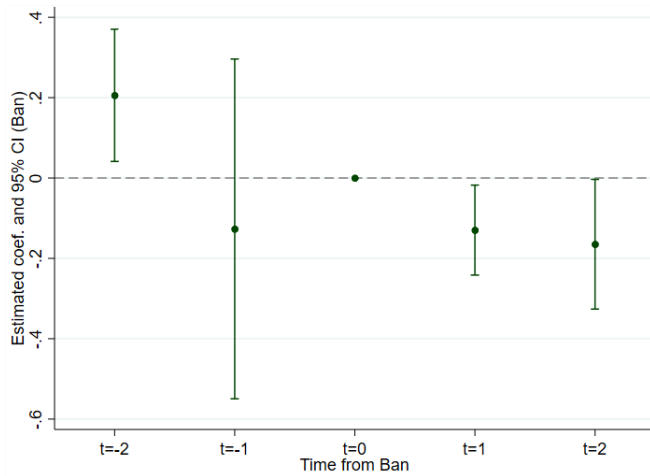
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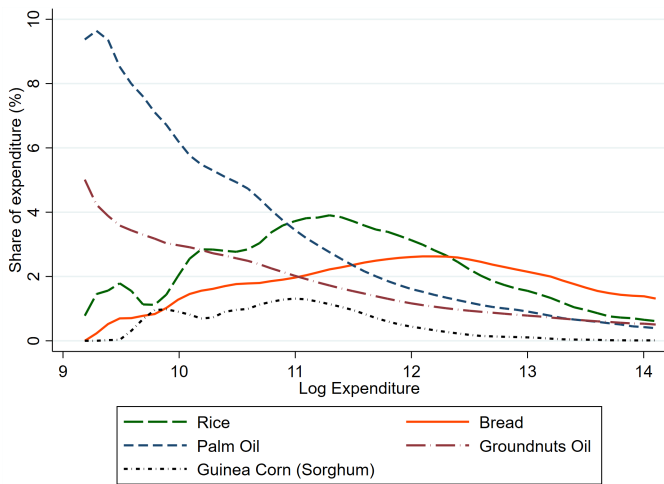
# Impact of ban removal on prices

Prices drop when bans are removed - but there are pre-trends



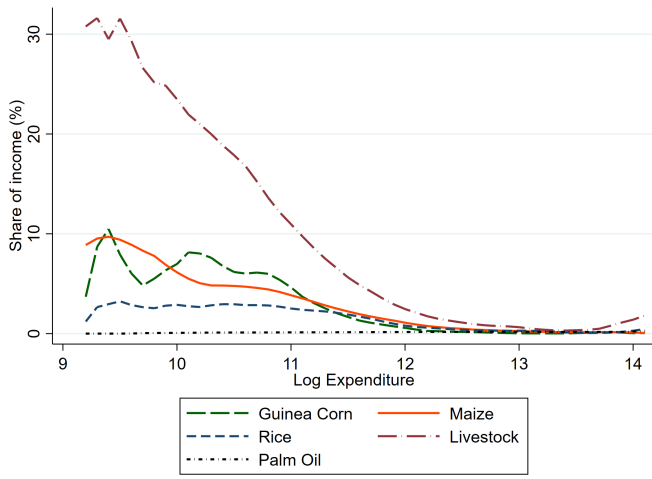
Notes: Figure shows the results of a local projection difference-in-difference estimation (Dube et al. 2025) controlling for  $(1 - \text{Ban}) \times \text{Taxes}$ .

# Expenditure shares specific banned products



[Go back](#)

# Income shares specific banned products



[Go back](#)