The Effects of China's Waste Import Ban on Pollution Relocation in the U.S.

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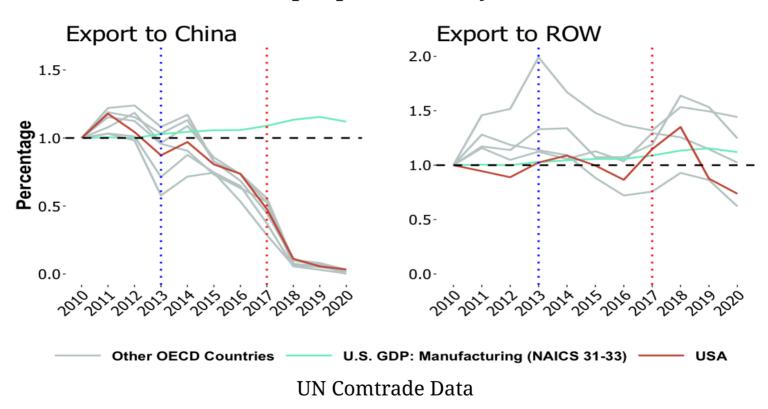
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Introduction

- Recycling rate 7% → 32% from 1960s to present
- For many years, most U.S. recyclables were exported to developing countries
- China was the biggest importer of U.S. recyclables
- In 2017, China implemented its **Green Sword (GS) Policy** banning almost all recyclable waste imports
- Considerable domestic environmental costs
 - air pollution from re-processing of these materials
 - landfill methane (GHG) emissions
 - land and water pollution
 - ocean disposal
- U.S. has no **economical or efficient** recycling infrastructure
 - Recyclables went to landfills.

GS Policy and Trade

Plastic Scrap Export Volume by Countries



- U.S. plastic scrap exports to China dropped by 99%
- U.S. plastic scrap exports to the ROW **increased**, **then decreased** after China's GS policy

Research Questions

- For the U.S.
 - What has been the effect of China's GS policy on **Domestic Emissions** from landfill facilities in the U.S.?

- For the state of **California**
 - What are the **Distributional Effects** of the GS policy on pollution relocation for local communities at census block levels?
 - What are **Environmental Justice (EJ)** implications?

Relevance

Trade and Environment Shapiro (2016), Shapiro (2018), Shapiro (2021)

→ My innovation: study a trade policy that directly restricts externality export and explore the policy's causal effects on local emissions in the U.S.

Environmental Gentrification and Environmental Justice Baden and Coursey (2002), Cameron and McConnaha (2006), Banzhaf and Walsh (2008), Depro et al. (2011), Banzhaf and Walsh (2013), Depro et al. (2015), Banzhaf et al. (2019), Ho (2020), Hernandes and Meng (2020), Shapiro and Walker (2021)

 \rightarrow My innovation: how an exogenous intl. trade policy affects U.S. EJ problems.

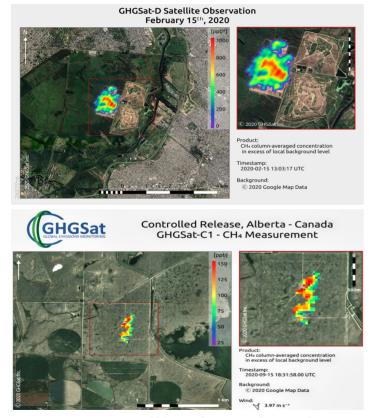
The efficiency of curbside recycling programs Adaland and Caplan (2006), Bohm et al. (2010), Kinnaman (2014), Kinnaman et al. (2014)

→ My innovation: show that in the absence of an overseas market for recyclables, the U.S. recycling system is inefficient even though it has the "efficient" recycling rate.

Behavioral Economics of Curbside Recycling Kurz et al.(2000), Halvorsen (2010), Ashenmiller (2009), Ashenmiller (2011), Best and Kneip (2019), Berck et al. (2020), Berck et al. (2021)

→ My innovation: use this exogenous trade policy as a tool to explore the relationship between the recycling programs and local environmental outcomes in the U.S.

Environmental Outcome Measurement



GHGSat's satellite showed methane from landfills (upper: Buenos Aires, Argentina. lower: Alberta, Canada)

- Why use **methane** emissions?
 - They are a **proxy** for general pollution emissions
 - Methane is far more potent at trapping the sun's heat than carbon emissions
 - Satellite data is available only from 2020
 - Need consistently measured emissions data from 2010 to 2020

Data

- U.S. EPA Greenhouse Gas Reporting Program (GHGRP)
 - Methane emissions from landfill facilities
 - 2010 to 2020 annually
- Approximately 8,000 facilities required to report emissions annually
- Covered industries include power plants, petroleum and natural gas systems, minerals, chemicals, pulp and paper, refineries, waste, etc.
- Data generation process:
 - Facilities in waste industry report the amount of wastes accepted annually
 - Methane Emission is calculated through a complicated model embedded in U.S. EPA

Data

- California Department of Resources Recycling and Recovery (CalRecycle)
 Disposal Flow Data
 - Captures the amount of disposal transported by origin jurisdiction and destination facility
 - 2002 to 2021 quarterly
 - Contains 464 origin jurisdictions and 263 disposal facilities
- Other Data Sources
 - U.S. Trade Census
 - EPA Enforcement and Compliance Historical data
 - Bureau and Labor and Statistics (BLS) Quarterly Employment and Wages at county level
 - o U.S. Census racial mix, median income at census-block level

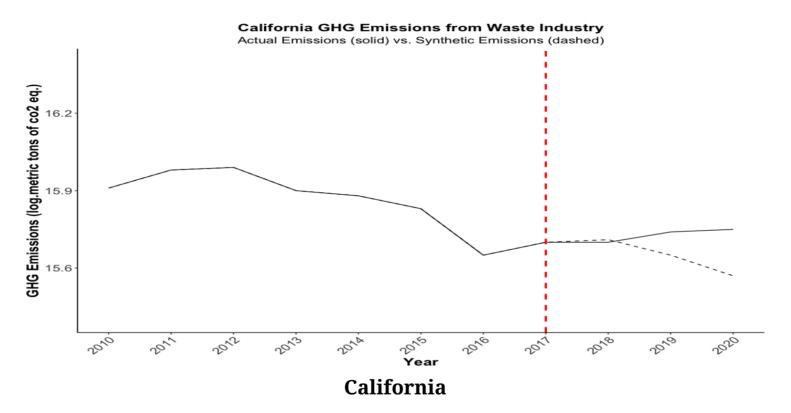
Identification: Synthetic Control Method

- Rely on exogenous variation in methane emissions across all other industries in the EPA GHGRP
 - Power plants, petroleum and natural gas systems, minerals, chemicals, pulp and paper, refineries, etc. (not waste)
- Take advantage of the fact that other industries which also emit GHG were not affected by China's GS policy
- Use other industries(all states) as donor pool for synthetic control group
- Train the model using pre-policy time 2010-2017
 - calculate state-industry pair weights to minimize prediction error

$$\hat{Y_{1t}^N} = \sum_{j=2}^{J+1} w_j Y_{1t} \, .$$

• Predict counterfactual methane emissions in the absence of GS policy using post-policy period (2018-2020)

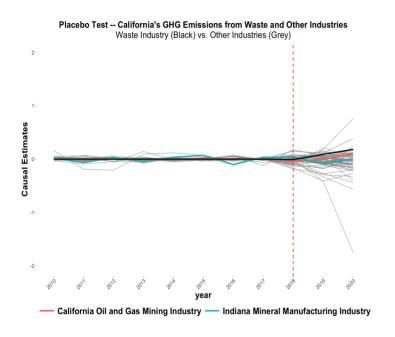
Results

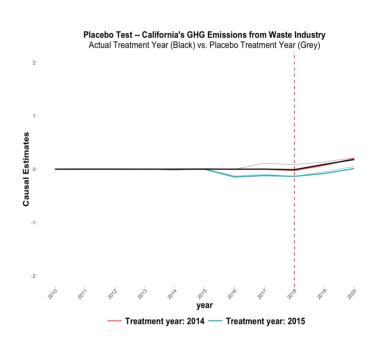


• The difference between actual emissions and synthetic emissions is the causal effect of China's GS policy on U.S. landfill methane emissions

$$\hat{ au_{1t}}=Y_{1t}-\hat{Y_{1t}^N}$$

Placebo Tests

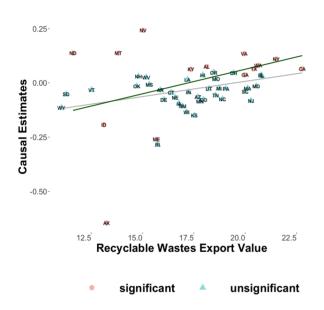




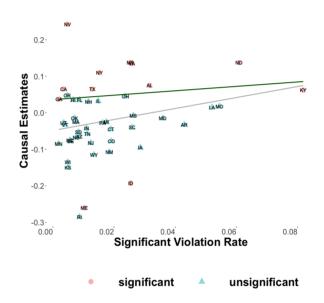
• Use control industries as "fake" treatment industries

• Use other years (2013-2017) as "fake" treatment years

Results



↑ Recyclable wastes a state
 exported → ↑ increase in
 methane emissions.



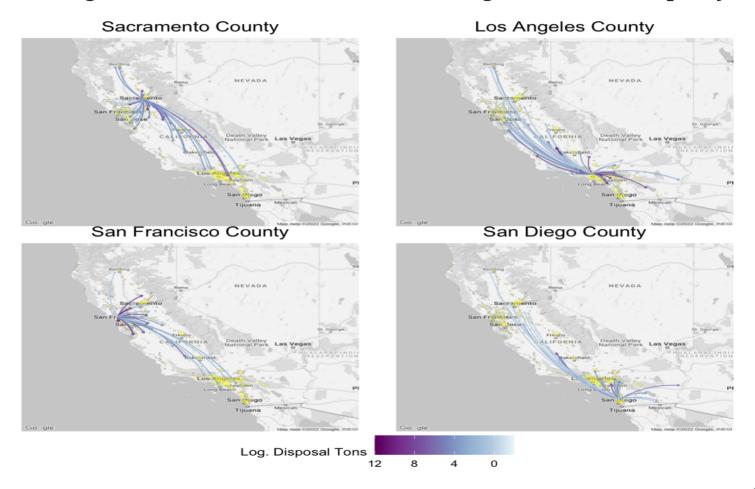
↑ State-level rate of
 "significant" environmental
 violations → ↑ increase in
 methane emissions.

Pollution Relocation

- How does International Trade policy affect pollution relocation?
- Does pollution relocate?
 - **Cap and Trade** Clean Air Act requires new or expanding plants to pay incumbents in the same or neighboring counties to reduce their pollution emissions (Shapiro and Walker (2021)).
 - **US air pollution offset markets** Clean Air Act allows for trading of permanent pollution emissions rights between firms within a metropolitan area (Shapiro and Walker (2020)).
 - Externality-export strategy for air pollution Clean Air Act (Morehouse and Rubin (2022))
 - **Waste flow** Not In my Backyard (NIMBY) regulations limit interstate waste flows (Ho (2020)).

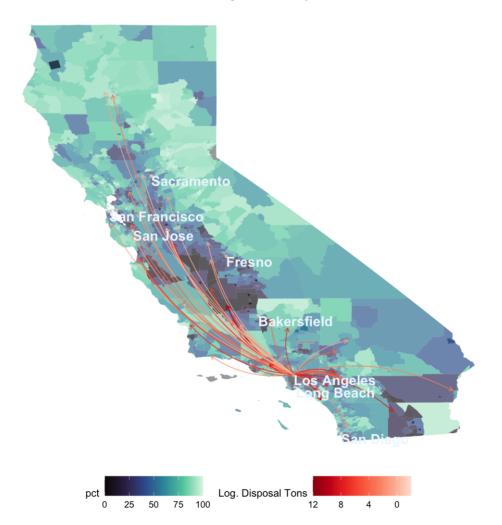
Pollution Relocation

Average net increase in waste flows across regions after the GS policy



Pollution Relocation by Racial Composition



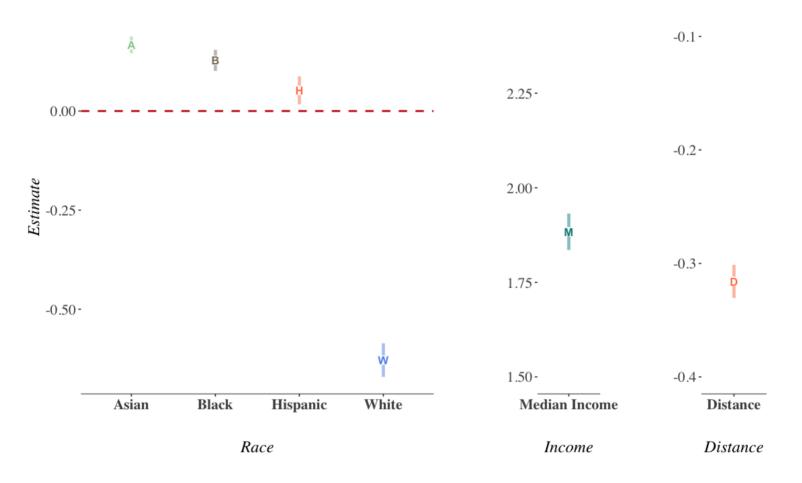


Gravity-type Model

$$egin{aligned} log(Y_{ijt}) &= lpha + eta_1 log(Dist_{ij}) + eta_2 log(R_j) + eta_3 log(X_{jt}) \ &+ eta_5 GS_{post} * log(R_j) + eta_6 GS_{post} * log(X_{jt}) \ &\epsilon_o + heta_d + \mu_{od} + \eta_t + \lambda_{odt} \end{aligned}$$

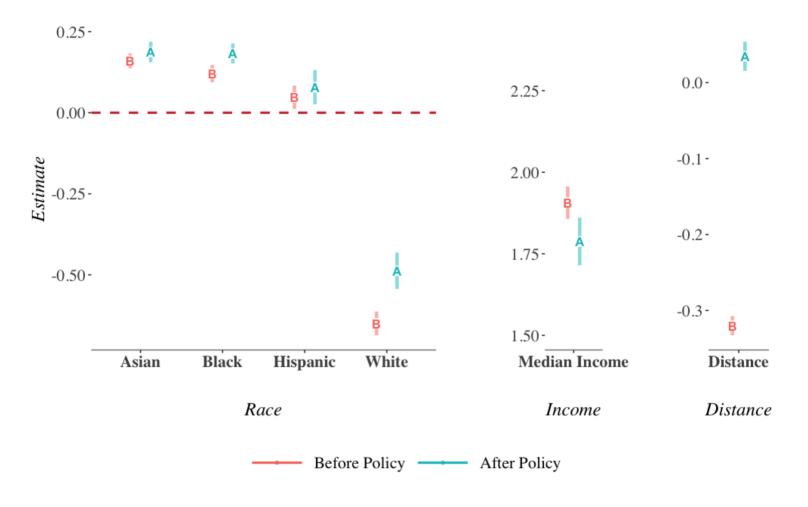
- ullet i origin jurisdiction of California; o origin county
- j area that is a 3km buffer within the destination facility; d destination county
- t year-quarter.
- R_{jt} racial compositions of destination j
- ullet Y_{ijt} tons of the disposal transported from i to j in year quarter t
- ullet GS_{post} dummy variable for the GS policy
- ullet $Dist_{ij}$ distance between origin i and destination j
- X_{jt} median income, regulation of environmental stringency, and economies of scale of waste industry of destination j
- Fixed-effects: ϵ_o , θ_d , μ_{od} , η_t , λ_{odt}

Results prior to the GS Policy



Gravity Model Estimates at Census Block level

Results after the GS Policy



Conclusion Preliminary Findings

• U.S. State-level Methane Emissions

- Many states show statistically significant increases in methane emission
- Relate to historical trade volume, stringency of envir. regulations

• California Pollution Relocation

Before China's GS policy

Waste tend to relocate to minority communities

After China's GS policy

Inflows increased more for remote low-income white communities