

Air Pollution and Manufacturing Firm Productivity: Nationwide Estimates for China

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Introduction

- Understanding cost of air pollution key to optimal regulation.
- Emerging literature focuses on labor productivity:
 - Graff Zivin and Neidell (AER, 2012): Fruit pickers in California.
 - Chang *et al.* (AEJ-Policy, 2016): Pear packers in California.
 - Chang *et al.* (AEJ-Applied, 2019): Call center staff in China.
 - He et al. (AEJ-Applied, 2019): Textile workers in China.
- However:
 - Focus on narrow groups of workers in particular occupations; external validity concern for broad-based environmental policies.
 - Only partial equilibrium analysis

Contributions

- Comprehensive, nationwide estimates:
 - All above-scale firms 1998 – 2007.
 - Over 90% of China's manufacturing output.
 - Find much larger impacts than previous studies (elasticity of -0.17)
 - A related paper by Dechezlepretre et al. on European regions
- First general-equilibrium analysis
 - Incorporate feedback from manufacturing output on pollution
 - Similar to Dynamic Integrated Climate-Economy (DICE) model (Nordhaus, Science 1992)
 - Effect is amplified to -0.31 considering the general equilibrium effect

Outline

- Introduction
- **Partial equilibrium: Pollution on output**
- General equilibrium
- Conclusion

Pollution on productivity

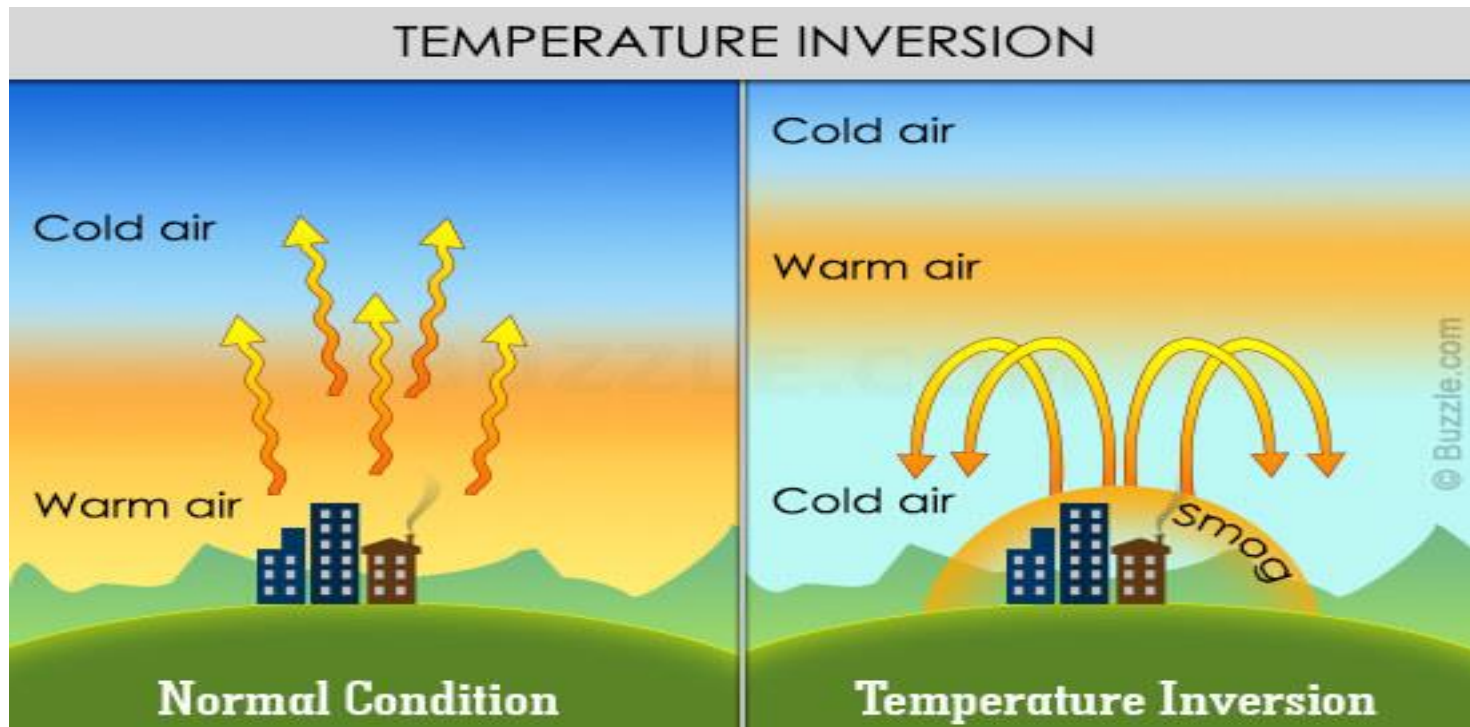
- $Productivity_{it} = \beta_0 + \beta_1\Omega_{it} + \beta_2W_{it} + \alpha_i + \rho_t + \varepsilon_{it}$:
 - i : firm; t : year.
 - Two measures of productivity: output per worker and TFP
 - Ω : PM_{2.5}.
 - W : weather controls.
 - α : firm fixed effects.
 - ρ : year fixed effects.
 - ε : firm-level unobserved productivity factors.

Identification Challenges

- Challenge 1: Simultaneity and omitted variables:
 - More output \rightarrow more pollution (bias \uparrow).
 - Use of alternative inputs (bias \uparrow or \downarrow).
 - Region-specific, time-varying correlation b/w output and pollution (bias \uparrow or \downarrow).
- Challenge 2: spatial sorting:
 - Firms choose regions:
 - With better air quality (\downarrow bias).
 - With worse air quality: “pollution haven” effect (bias \uparrow or \downarrow).
 - High-skilled workers more productive and willing to pay more for air quality (\downarrow bias).

IV Approach

- Use county-level thermal inversions as instrument (Arceo *et al.*, 2016):
 - Meteorological phenomenon that increases pollution.
 - Conditional on weather, not a health risk.



Spatial Sorting

- Approach to firm spatial sorting:
 - Firm fixed effects absorb any initial sorting.
 - Firm sorting during sample period limited (7% change counties and robust to dropping).
 - Pollution not predictive of firm entry or exit.
- Approach to worker spatial sorting:
 - Firm fixed effects absorb any initial sorting.
 - Pollution not predictive of fraction of workers in high-versus low-technology firms.

Data (1)

- Sample period: 1998-2007.
- Productivity data (firm-year level):
 - Annual survey of manufacturers (NBS).
 - All SOEs, non-SOEs with revenues > CNY 5 million.
 - About 350 thousand firms and over 1.5 million observations.
 - Detailed balance-sheet data.
- PM_{2.5} pollution data (county-year level):
 - Satellite-based (NASA).
 - Inferred from Aerosol Optical Depth.
 - Resolution 50 x 60 km, convert to county.

Data (2)

- Thermal inversion data (county-year level):
 - Satellite-based temperature measure at 42 atmospheric layers each 6 hours (NASA).
 - Thermal inversion if temperature at 320 meters (2nd layer) higher than at 110 meters (1st layer).
 - Resolution 50 x 60km, convert to county.
 - Annual days with inversions.
- Merge pollution and inversion with firm data by county-year.

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Output on pollution

- $\Omega(t) = [\lambda(t)Q(t)]^{\mu(t)}$
- $\Omega(t)$ -PM_{2.5}
- $\lambda(t)$ -Fraction of total output produced by manufacturing sector
- $\mu(t)$ -Elasticity of pollution with respect to output
 - Using China joined the WTO as the IV
 - DID (coastal vs inland regions; before and after)

Appendix 13: OLS and 2SLS estimates (effect of output on pollution) using effect of China joining WTO on coastal versus inner regions as an instrument

	(1)	(2)
	OLS	2SLS
Dependent variable:		<u>First stage</u> <u>ln(Value added)</u>
Coast*post 2001		0.0574*** (0.0147)
KP <i>F</i> -statistic		15.3
Dependent variable:		<u>Second stage</u> <u>ln(PM_{2.5})</u>
ln(Value added)	0.0048*** (0.0012)	1.4317*** (0.3666)
County fixed effects	Y	Y
Year fixed effects	Y	Y
Sample size	25,357	25,357

General equilibrium

- $\max_{c(t)} \sum_{t=1}^T P(t) \ln[c(t)] (1 + \rho)^{-t}$
- $Q(t) = \Omega(t)^\theta A(t) K(t)^\gamma P(t)^{1-\gamma},$
- $\Omega(t)^\theta$ -how pollution affects output
- $\Omega(t) = [\lambda(t) Q(t)]^{\mu(t)}$ -how output affects pollution
- $Q(t) = C(t) + I(t)$
- $K(t) = (1 - \delta)K(t - 1) + I(t)$

General equilibrium

- A 1% decrease in $PM_{2.5}$ increases manufacturing output by 0.31%, or 0.043% of China's GDP
- Partial equilibrium is 0.17%
- Intuition: decreased output results in somewhat less pollution and thus somewhat more output
- Important for policy evaluations

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Contact

- Working paper
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