

Final Oral Presentation

M1-Economics (international track), Group 13
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Context & Motivation

Economic Question

- ▶ What is the impact of pollution policies on firms' Total Factor Productivity (TFP) in China?

Context

- ▶ Rapid industrialization in China led to severe pollution
- ▶ Stricter environmental policies introduced in recent years
- ▶ Firms must balance compliance costs with productivity

Context & Motivation

Economic Question

- ▶ What is the impact of pollution policies on firms' Total Factor Productivity (TFP) in China?

Motivation

- ▶ Pollution from firms: a major environmental, health and economic issue
- ▶ It is possible to keep a sustainable growth with severe measures against pollution ?
- ▶ Leads to a trade-off

Literature Review and Contribution

- ▶ Muller and Mendelsohn (2009): Efficient pollution regulations aligning costs and damages improved overall welfare and economic efficiency
- ▶ Alpay, Kerkvliet, and Buccola (2002): Environmental regulations in food manufacturing **positively affected productivity** by fostering innovation (**Porter Hypothesis**)
- ▶ Leiter, Parolini, and Winner (2011): Stricter environmental regulations led to **increased investments** in cleaner technologies, indirectly **enhancing productivity**
- ▶ **How does this study contribute ?**

Literature Review and Contribution

How does this study contribute ?

- ▶ Regional Differences in China: Analyze how variations in environmental regulation across regions affect TFP
- ▶ Industry-Specific approach: Study the overall impact of environmental policies, controlling for industry
- ▶ Causal Evidence with DID: Use the difference-in-differences (DID) method to measure the causal effect of regulation on TFP, comparing regions with different levels of regulatory stringency over time

Literature Review and Contribution

China's Unique Context

- ▶ Focus on China as a rapidly growing economy balancing industrial growth with environmental goals, offering lessons for other developing countries
- ▶ The study will be based on **policies on atmospheric pollution applied in 2017**, such as Hebei Air pollution prevention and control program, aiming to restrain PM2.5 emissions
- ▶ Hebei pollution program: Non complying firms were **sanctionned** or shutted down, **financial incentives** applied...

Dataset Description

The database was found in the study "How air pollution affects corporate total factor productivity?" by Jialiang Yang and Wen Yin

► **Data Source mainly derived from:**

- The Guotaian CSMAR database
- Macroeconomic data from the China Urban Statistical Yearbook
- Patent data from the State Intellectual Property Office on the technical complexity of cities

► **Nature of Data:** Panel Data (2015-2019)

Dataset Description

- ▶ **Collection Method:** Data collected from financial reports of firms and government-published AQI reports

- ▶ **Sample**

An observation corresponds to a company listed on the Shanghai and Shenzhen stock exchanges (A-shares) for a given year

- **Before selection:** 13 738 observations and 29 variables (2015-2019)
- **After selection:** 875 observations and 26 variables (2015-2019)
- **Variables added by ourselves:** "Administrative division", "Treated", "Post" and "ATT"

Main variables Description

- ▶ **Treated:** equal to 1 if the firm is located in **Hebei, Beijing or Tianjing**, 0 otherwise (508 control observations, 367 treated observations)
- ▶ **Post:** equal to 1 after 2017 (introduction of the policy) and 0 for 2015-2016
- ▶ **Dependent variable: TFP**
 - **TFP** determined by the Levinsohn & Petrin (LP) method, it is an **index**
 - In the econometric models of the study, **TFP is log transformed**

Variables Description

► Main variables of interest

- **Number of Employees (NumbEmp)** → In individuals
- **Investment in R&D by listed companies (CRD)** → Annual R&D expenses in Chinese yuan (CNY)
- **Gross Domestic Product (GDP)** → The GDP of the listed company's local city in thousands of Chinese yuan (K CNY)
- **Government Concerns (GS)** → Total annual government subsidies for listed companies in Chinese yuan (CNY)
- **Average treatment effect on treated (ATT)** → 1 if the firm is treated and the year is strictly after 2016, 0 otherwise
- **Free Cash Flow (FCF)** → Cash flow from operating activities divided by balance sheet total at the end of the financial year

Propensity Score Estimation and Firm Matching

- ▶ **Objective:** Estimate the probability of receiving the treatment (policy) using logit regression
- ▶ **Explanatory variables :** Firm characteristics (before policy) such as TFP, R&D expenses, AQI, Size in assets, Free Cash Flow, Cash amount, town GDP, liabilities, Industry...
- ▶ **Matching procedure:** Nearest Neighbor Matching with a caliper of 0.2, matching on year pre-treatment, and **with replacement** (maximum 2)
- ▶ **Outcome:** 875 observations from 180 firms (105 control, 75 treated)

Correlation to treatment (pre-treatment)

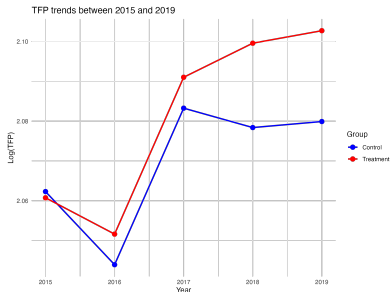


Figure 1: TFP trends: treated vs non-treated

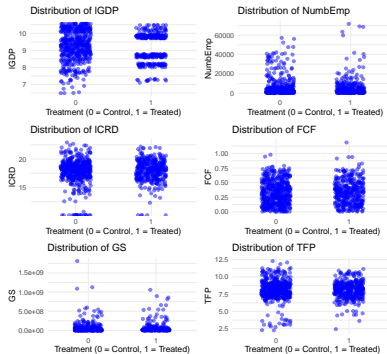


Figure 2: Distribution of key variables with treatment

- ▶ Successive T tests on covariates indicated **no significant mean differences** between groups **before treatment**
- ▶ **Parallel trends assumption** seems respected

Regression Specifications

- ▶ **Main regression equation (controlling for year FE):**

$$\log(\text{TFP})_{it} = \beta_1 \text{att}_{it} + X_{it}\beta + \sum_j \gamma_j \mathbf{1}(\text{year}_i = j) + \varepsilon_{it} \quad (1)$$

- ▶ **Auxiliary equations:**

$$\log(\text{TFP})_{it} = \beta_1 \text{att}_{it} + X_{it}\beta + \text{Industry FE} + \text{Year FE} + \varepsilon_{it} \quad (2)$$

$$\log(\text{TFP})_{it} = \beta_1 \text{att}_{it} + X_{it}\beta + \text{Firm FE} + \varepsilon_{it} \quad (3)$$

- ▶ **Where X_{it} contains the control variables**

Residuals and heteroscedasticity

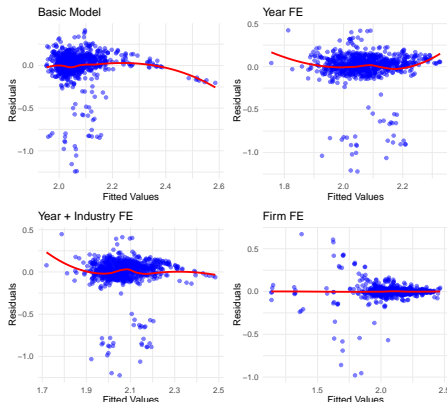


Figure 3: Residuals vs. Fitted values

- ▶ Light heteroscedasticity suspicion for Firm FE, and strong suspicion for the others

Table 1: Breusch-Pagan Test Results

Model	BP Statistic	p-value
Basic Model	18.033	0.00291
Year FE	20.860	0.00753
Y + I FE	28.284	0.02924
Firm FE	7.8624	0.04894

- ▶ P-value < 0.05 , so we reject homoscedasticity
- ▶ We need robust errors to estimate

Regression results

Dependent variable log(TFP)

	No Control	Year FE	Industry & Year FE	Firm FE
ATT	0.024 08* (0.011 91)	0.018 36** (0.002 41)	0.021 24*** (0.001 47)	0.032 19* (0.013 91)
log(NumbEmp)	0.049 82*** (0.007 41)	0.051 54*** (0.003 90)	0.043 56** (0.005 97)	0.037 97** (0.012 00)
log(CRD)	0.017 69** (0.006 25)	0.017 26* (0.005 49)	0.020 10** (0.004 28)	-
log(GDP)	0.012 49 (0.009 27)	0.013 45 (0.007 38)	0.003 25 (0.007 26)	-0.003 82 (0.054 28)
log(FCF)	0.007 55 (0.005 37)	-	-	-
GS	-	-	0.000 01 (0.000 01)	-
Intercept	1.452 73*** (0.098 90)	-	-	-
Num. Obs.	875	875	875	875
R²	0.2881	0.175	0.218	0.760
R² Within Adj.	0.2512	0.166	0.159	0.026

Table 2: Regression results with robust errors

Key results

- ▶ **Interpretation:** The introducing of a policy is associated to a 2.124% increase in TFP on average, when controlling for year and industry FE
- ▶ Robustness check:
 - ▶ **Subsector control:** $ATT(*) = 0.03036$
- ▶ Main takeaway: The relationship is positive and statistically significant. **ATT varies between 1.8% and 3.2%**
- ▶ Policy implications: Encouraging policies that include pollution in production cost can positively impact firm's efficiency

Key results

- ▶ **Conclusion:** According to these results and in alignment with existing literature, environmental regulation has a positive impact on TFP
- ▶ **Limits and suggestions:**
 - Having data on previous year to 2015 could **have provided more significant matching**
 - Having a clearer idea of **how policies were applied**
 - Extend the analysis to **other type of policies** targeting other type of pollutions than gas emission (ex: Food and Catering industry)
 - Extend the analysis of the **impact of policies on emission quotas** implemented in 2021 in China and in recent years