Public Pressure and Heterogeneous Effects of Voluntary Pollution Abatement

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Motivation

Voluntary pollution abatement programs (VPAs) are widely used complement for the mandatory regulations:

- Flexible: lack quantitative targets, sanctions and punishment.
- Do not require legislative action.
- No direct benefit.

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Research Questions

- How do VPAs create an incentive for firms to participate?
- What factors determine the effectiveness of VPAs in reducing pollution?

Literature

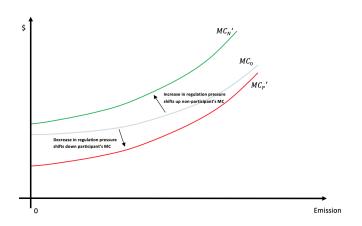
- VPAs are effective. (Khanna and Damon, 1999; Bi and Khanna, 2012; Innes and Sam, 2008).
- VPAs are not effective. (Welch et al., 2000; Gamper-Rabindran, 2006; Vidovic and Khanna, 2007; Brouhle et al., 2009; Carrion-Flores et al., 2013; Vidovic and Khanna, 2012; Vidovic et al., 2019)
- VPAs participants increase emissions compared to non-participants.
 (King and Lenox, 2000; Gamper-Rabindran and Finger, 2013)

Three Factors Affecting Firm's Emission Behavior:

- Firm characteristics related to abatement cost
 - Example: Managerial skill, size, energy type, equipment, ownership...
- Emission cost from mandatory regulation pressure
 - Source: government intervention/regulation
 - Example: environmental taxes, expected inspections and violation penalties.
- Emission cost from public pressure
 - Source: Coasian bargaining in corporation-society conflicts (Heal, 2005)
 - Example: "green" consumers boycott, loss in stock market, negative media coverage, local communities protests and lawsuits...

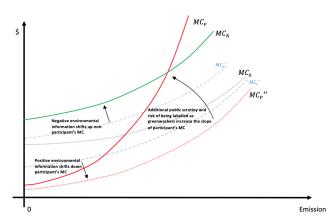
Regulation Pressure

VPA shifts regulation resources from participants to non-participants (Gamper-Rabindran, 2006; Innes and Sam, 2008; Li and Khanna, 2018).



Public Pressure

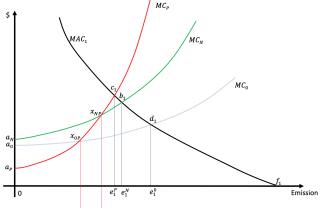
- VPA participation as positive news lowers public pressure, and vice versa (Aerts et al., 2008).
- VPA participation brings public scrutiny and the associated risk of being labeled greenwashers (Lyon and Maxwell, 2011; Kim and Lyon, 2011).



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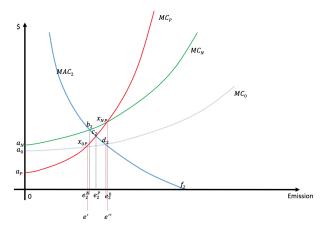
3 Types of Firms: High Pollution Firms

- 1. MAC intersects MC_P and MC_N on the right of x_{NP} ;
- 2. Always reduce emissions, participants $(d_1 \rightarrow c_1)$ reduce more than non-participants $(d_1 \rightarrow b_1)$;
- 3. More polluting firms have smaller participation incentive ($Area(a_Na_Px_{NP}) Area(x_{NP}b_1c_1)$; $Area(x_{NP}b_1c_1) \uparrow$)



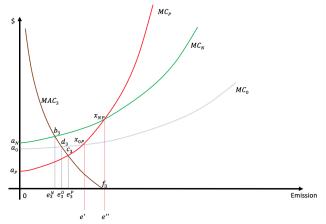
3 Types of Firms: Medium Pollution Firms

- 1. MAC intersects MC_P between x_{NP} and x_{OP} ;
- 2. Always reduce emissions, participants $(d_2 \rightarrow c_2)$ reduce less than non-participants $(d_2 \rightarrow b_2)$;
- 3. More polluting firms have greater participation incentive $(Area(a_na_pc_2b_2)\uparrow)$



3 Types of Firms: Low Pollution Firms

- 1. MAC intercepts MC_P on the left of x_{OP} ;
- 2. Non-participants reduce emissions $(d_3 \rightarrow c_3)$, participants increase emissions $(d_3 \rightarrow b_3)$;
- 3. More polluting firms have greater participation incentive $(Area(a_na_pc_3b_3)\uparrow)$



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Testable Hypothesis

Hypothesis 1. If the lagged emission level $e_{t-1} \leq e''$, then a marginal increase in e_{t-1} increases the probability of participation. If the lagged emission level $e_{t-1} > e''$, a marginal increase in e_{t-1} decreases the probability of participation.

Hypothesis 2. If the lagged emission level $e_{t-1} \leq e''$, then a participant reduces emissions by a smaller magnitude than non-participant (and is a free-rider). If the lagged emission level $e_{t-1} > e''$, then a participant reduces pollution by a larger amount than non-participant.

Data

We use the EPA 33/50 program (1991-1995) data set from Zhou et al. (2020).

The goal of EPA 33/50 program was to reduce total emissions of 17 toxic chemicals by 33% by 1992 and by 50% by 1995, compared to the 1988 baseline.

There are 8,670 plants from 1988 to 1996.

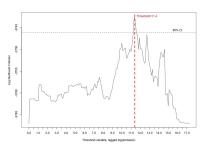
We consider two samples: the whole sample, and a 1991 sub-sample (with all non-participants and only first year (1991) participants).

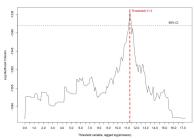
Empirical Model

Two-stage regression to solve the endogeneity from self-selection in participation.

- First stage probit model estimates participation incentives
- Identify e'' by grid search that maximizes likelihood of probit model (Hansen, 1999, 2000).
- Second stage dynamic panel model estimates the participation effect (instrumented by first stage estimates) using Arellano and Bond (1991) GMM.
- ullet Allow heterogeneous participation effect on either side of the threshold $e^{\prime\prime}$.

Empirical Results: First Stage





- 33/50 program participation incentive in 1991-1995
- (a) MLE criterion of threshold parameter, (b) MLE criterion of threshold parameter, 33/50 program participation incentive in 1991

Figure 1: Threshold estimation and 95% confidence interval

Empirical Results: Second Stage

Table 1: Main analysis: 1991-1996 participation effect (benchmark results)

| | Dependent variable: log(33/50 emissions), first difference | | | | | |
|--|--|----------------------|----------------------|----------------------|---------------------|----------------------|
| | Full Sample | Full Sample | | 1991 Sub-Sample | 1991 Sub-Sample | |
| | | Before threshold | After threshold | | Before threshold | After threshold |
| Participation Status | -0.556*** (0.053) | -0.183*** (0.062) | -0.956*** (0.104) | -0.287*** (0.072) | 0.758*** (0.190) | -0.806*** (0.109) |
| Control Variables | Υ | Υ | | Υ | Υ | |
| Industry Specific Time Trends | Υ | Υ | | Υ | Υ | |
| State Specific Time Trends | Υ | Y | | Υ | Y | |
| Threshold log emission | - | 11.4 | | - | 11.3 | |
| Over-identification test (Hansen J Statistics P Value) | 0.2794 | 0.3242 | | 0.2105 | 0.2729 | |
| Weak identification test (Cragg-Donald Wald F Statistics) | 69.611 | 70.083 | | 49.133 | 47.385 | |

Note: *p<0.1; **p<0.05; ***p<0.01 Covariates: lag log(33/50 emissions)

Covariates: lag log(33/50 emissions), HAP/TRI Ratio, LCV Score, county nonattainment status, county income per capita. All covariates are in first differences.

IV's are first stage estimated participation probability and the third-year lag of log 33/50 emissions.

Result Interpretation

$$e_{ijt}^{P}/e_{ij,t-1} = e_{ijt}^{N}/e_{ij,t-1} \times exp\Big(\gamma_{1} \mathbb{1}_{e_{ij,t-1} \le e''} + \gamma_{2} \mathbb{1}_{e_{ij,t-1} > e''}\Big),$$

$$e_{ijt}^{P} = e_{ijt}^{N} \times exp\Big(\gamma_{1} \mathbb{1}_{e_{ij,t-1} \le e''} + \gamma_{2} \mathbb{1}_{e_{ij,t-1} > e''}\Big).$$
(1)

Participation effect estimated by the full sample:

- Below the threshold: -16.7%
- Above the threshold: -61.6%

Participation effect estimated by the 1991 Sub-Sample:

- Below the threshold: +113.4%
- Above the threshold: -55.3%

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Conclusion

- We develop a theoretical model illustrating a firm's incentives to participate in VPAs and outline the firm's emission decisions.
- The scrutiny of participating firms' environmental outcomes by the public is the key factor in determining whether VPAs participants free-ride the program or not.
- Public pressure can be effectively leveraged to complement traditional regulation and keep firms from free-riding in voluntary pollution abatement programs.

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Questions, Comments and Suggestions

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