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Regional Heterogeneity in Preferences for Environmental Regulation and the Effect of Pro-environment Voting on Toxic Emissions

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# Regional Heterogeneity in Preferences for Environmental Regulation and the E ect of Pro-environment Voting on Toxic Emissions

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#### **Abstract**

In this paper, I use county-level measures of pro-environment voting from the U.S. House of Representatives as a proxy for regional heterogeneity in preferences of citizens for more or less regulation in order to estimate their e ect on toxic air emissions at a local level. Even though constructing county-level voting scores from congressional district scores requires a degree of approximation in counties that lie partially in multiple districts, the fact that county lines do not change with the decennial Census allows

## 1 Introduction

The structure of the air quality regulatory environment in the United States is such that minimum federal standards are set by the Environmental Protection Agency. Federal standards could include maximum allowable ambient concentration of certain pollutants or requirements of the technology that must be employed by new or existing rms. Over time, the enforcement of federal standards has become the responsibility of local enforcement agen-

from the U.S. House of Representatives or U.S. Senate as a proxy for community attitudes. It seems reasonable to assume that a county-level voting score would be a better proxy for the local regulatory environment than voting scores at the state level, because aggregation at the state level fails to identify which communities in the state are pro-environment. This is important because within each state there are \green" counties and counties that care comparatively little about the environment. An independent organization known as the League of Conservation Voters (LCV) keeps scorecard records on pro-environmental voting behavior of both U.S. Representatives and U.S. Senators. Using these scores provides a measure of how each politician voted and is assumed to proxy how much each community or county values the environment, regardless of how many pro-environment bills are actually passed at the national level.

Several studies have attempted to link measures of citizen attitudes toward pollution to regulatory stringency and its impact on rm behavior. For example, Henderson [16] considers state attitudes toward pollution as measured by the xed-e ect term from a xed-e ects regression with pollution abatement expenditures as the dependent variable. This xed e ect measures the degree to which states either \over spend" or \under spend" on abatement activity with overspending being associated with pro-environment attitudes. He identi es measures of time-invariant attitudes toward pollution and nds that a 1-percent increase in abatement expenditures leads to a 0.04-0.05 percent improvement in air quality measures. Gray and Shadbegian [13] evaluate temporal and cross-sectional variation in state-level aggregates of League of Conservation Voting (LCV) records and nd that the share of a

rm's production arising at the state level is negatively related to LCV scores. Gray [12] also uses state-level aggregates of LCV scores as a measure of attitudes towards pollution and

since within counties there are differences for or against increased regulation that need to be considered by the legislator. Figure 1 summarizes the key features of the model and I have identified important links which I will refer to as links  $A, \ldots, F$ .

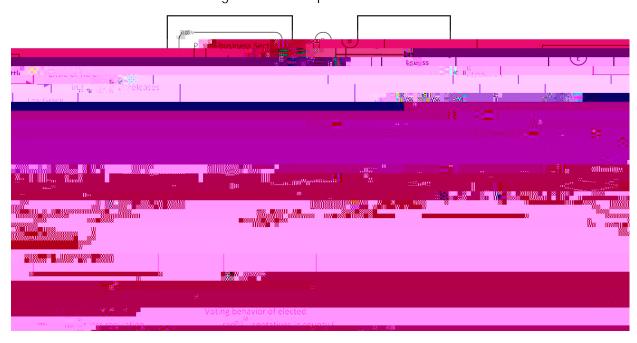


Figure 1: Conceptual Framework

I consider four groups of individuals whose attitudes or preferences for more or less

The model suggests that the adverse e ects of pollution will be primarily present in lower income and minority communities. The designation of high versus low income is somewhat arbitrary and is slightly unclear from the literature what the exact distinction should be regarding who is most a ected by pollution. The poverty line could be chosen as the speci-c means of separating low from high income designation within counties, but it seems at the county level from Figure 2 that the counties most negatively a ected by toxic releases are counties with a per capita income slightly higher than the poverty line. Per capita income is sensitive to high income outliers and income distributions are usually right-skewed, which would suggest that those most a ected by the pollution are those who may be below the poverty line. Link A in Figure 1 shows that pollution a ects lower income and minority populations.

Preferences for more or less regulation vary by group. Those individuals closely associated with business interest will prefer less regulation (link *B*) since more regulation leads

polluting industry. It is assumed that these individuals would prefer job security to more regulation. According to this model, these individuals' preferences would be represented by link *B* in Figure 1. The preferences of the individuals who live in high-income neighborhoods are uncertain. It is reasonable to assume that these individuals place a high value on environmental quality, but it is unclear whether they prefer regulation as a means of obtaining higher environmental quality. The most likely outcome will be that those who can a ord to move to locations with higher environmental quality will self-select into cleaner neighborhoods rather than relying on the government to provide it for them. On the other hand, there may be individuals who prefer a cleaner environment for society as a whole for altruistic reasons and they realize that regulation is one possible means of achieving that objective. These individuals are generally the more educated and realize that better air quality is a public good that is likely to be under produced. Therefore, it is possible that the high-income households could prefer either more or less regulation (link *C* in Figure 1), even though individuals acting in their own self-interest would simply move to cleaner locations.

Three likely objectives of career politicians are re-election (do whatever it takes to keep their job), altruism (place high priority on doing what is in their constituents' interests), and contribute their own ideologies to the decision making pro9 (d) 9rdlot-329(,at)-411e-(o-33)-403(triple).

not all be equally represented. One would expect those groups who are the most organized to communicate their preferences most clearly. Often the most organized are those representing business interests and are frequently found in Washington D.C. lobbying for less regulation. Communities that are less homogeneous, such as minority communities, are less likely to form collective action against polluting industries. The longer the terms of elected representatives, the greater is the likelihood of shirking from the constituents interests, because they are most likely to take into account constituent interests when they are close to re-election. The term length of U.S. Representatives is two years which makes them more accountable to their constituencies than U.S. Senators whose term lengths are six years.

The primary focus of this paper analyzes how voting behavior of U.S. Representatives a ects the level of county-level toxic air releases (links E,F in Figure 1). The theory would predict that the more pro-environment the representatives vote the greater the reduction on toxic releases (link F). The argument is that if there is pro-environment voting at the national level, then there must be overwhelming support for more regulation at the county level, especially since the ones most likely to support more regulation are the ones whose voices are least likely to be heard.

# 2.1 Which groups are most a ected by pollution?

There exists a wide body of literature dealing with the question of how community characteristics in uence environmental outcomes. Generally, all studies have arrived at the conclusion that the two groups most a ected by pollution are low-income communities and minority communities, although most studies argue in favor of either one or the other. The distribu-

tion of county-level toxic emissions in Figure 2 shows that there is a very high concentration of toxic releases in counties in which the per capita income level is below \$25,000, where Figure 3 shows TRI facilities are generally located in counties with a per capita income level of \$30,000 or below.

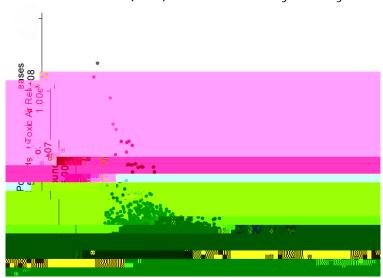


Figure 2: Distribution of Toxic (TRI) Air Emissions by County Per capita Income

A number of studies have analyzed within-county variation in community characteristics to try to identify which groups are the most disproportionally exposed to toxic releases. The following studies have conducted zip code-level analyses and have arrived at varying conclusions: Banzhaf and Walsh [3], Brooks and Sethi [4], Ringquist [25], Arora and Cason [2]. Link *A* is based on the ndings of these papers. For this model, I assume that both low-income and minorities are a ected by toxic releases.

According to Banzhaf and Walsh [3], low-income families are the most negatively affected. They conduct an empirical test of the Tiebout [27] hypothesis that individuals sort into communities with optimal bundles of taxes and public goods. Assuming rm location

be exposed to environmental hazards and are more likely to bene t from the provision of a cleaner environment.

Ringquist [25] evaluates the claim that TRI facilities are located in poor and minority communities and, after controlling for a variety of background factors, nds that TRI facilities and pollution are concentrated in zip codes with large minority representation. Brooks and Sethi [4] nd that minority (or speci cally more ethnically diverse) communities are more likely to be a ected by pollution due to the lower likelihood of collective action. They also nd that only for the highest income groups with annual incomes exceeding \$67,000 per year does higher median income imply lower exposure to emissions. It could be that speci c groups are targeted when rms emitting hazardous waste make decisions to locate, for instance because of the perception that certain types of communities will be less willing and able to engage in costly collective action against the rms.

# 2.2 Which groups prefer more regulation?

In this section, a distinction must be made between preferences for cleaner environment and preferences for more regulation. It is generally accepted that most people recognize the health bene is of a clean environment and that it contributes positively to the value of outdoor recreation. Those with the means of a ording it will obtain higher environmental quality through such examples as the purchase of homes in the foothills of the mountains, private golf memberships, or eco-tourism. They are likely to prefer less regulation because the bene is do not directly a ect them. They will likely face higher taxes as a result and possibly experience a reduction in home values as previously undesirable areas become more

in demand. Those who cannot a ord a clean environment for themselves will have to rely on the government to regulate and protect their health.

Fischel [8] nds that income, occupation, and education are robust determinants of preferences for environmental quality and that voting on environmental quality is divided along economic and social class lines. Some studies have used referendum data in an attempt to identify how di erent groups within a region di er in their preferences for regulation. Kahn and Matsusaka [18] using data from sixteen California Initiatives nd that environmental goods such as parks appear to be normal goods for people with the mean income level and inferior goods for people with high income. Their indings support the claim that the wealthy can purchase these goods privately and therefore do not prefer public provision of environmental quality which would be provided through higher taxes. Kahn [17] focuses on how changing demographics a ect the perceived bene ts and costs of regulation, and nds that minorities, youths, the more educated, and those who do not work in polluting industries are more likely to support environmental regulation. Elliot et al [7], using aggregate level determinants of support for environmental protection over a span of two decades, and that as real per capita income increases, support for additional spending on environmental policy increases as well. They obtain public opinion data from both the National Opinions Research Center (NORC) and the Roper Surveys that solicit respondents' views on environmental spending

argued that minorities are less likely to form collective action [4] and are therefore less likely to convey their concerns. Because di erent groups are less likely to bond with members of another minority group, this is even more of a concern when the composition of minority communities is heterogeneous. It is also important to consider the opportunity cost of each individual group's time. Lower income families do not have the luxury of much free time for collective action. Lower paying jobs require more hours of labor to earn money necessary for survival. Therefore, the opportunity cost of lobbying politicians is much higher for low-income families than for those with higher incomes and more free time.

#### 2.3 How do legislators decide which way to vote?

To consider which way a legislator will vote, one must—rst identify the incentives facing the individual. The incentives will be very di—erent based upon the position of government under consideration. If many of these public o—cials have chosen this as their career of choice, then it seems reasonable to assume that they would have a strict preference to be re-elected so that they might continue in this line of work. There is also the possibility that certain individuals would like to work their way up to a legislative decision-making position o—ering them a chance to make their own political ideologies heard. Another possible incentive would be to do whatever is best for constituents, making constituent interest a priority.

Peltzman [24] starts with a basic framework in which voting patterns are a function of ideology of the legislator and the interest of the constituents. Fort et al [9] add in a time-path component to the model which addresses the sensitivity of shirking behavior near re-election time. Since ideology (1) and citizen preferences (P) are not directly observable, all studies

that closer scrutiny at re-election time is expected to tighten the principle-agent relationship, so  $\frac{\mathscr{C}}{\mathscr{C}}$  < 0 implies the closer the representative is to a re-election year, the more closely they would be expected to take into account constituent preferences. is a measure of altruism which is on the interval [0;1], where 1 means that the legislator cares a lot about doing what is best for their constituents, regardless of whether they are up for re-election or not, and 0 means they do not care at all except for the purpose of being re-elected. should approach 1 + as the legislator gets closer to an election year.

The six-year term length of U.S. Senators makes them less accountable to their constituents, at least for the rst three to four years of their term, compared to U.S. Representatives who serve only two-year terms and are more dependent on keeping constituents satis ed for frequent re-elections. Therefore, U.S. Representatives should echo the voices of their constituents much more closely than U.S. senators. The key assumption here about the

Kalt and Zupan (1984) [19] nd that both constituent interests and legislator ideology are important factors. They nd evidence that within a principal-agent relationship legislators operate with enough slack to vote according to their own ideological tastes. Kalt and Zupan (1990) [20] use an ideological residual which is consistent with a liberal-conservative ideological spectrum and that is shown to respond to slack in the principle-agent relationship. Hamilton [15] concludes that the theory of rational political ignorance can help explain legislator preferences for policy instruments to control pollution. Legislators from districts with more toxic emissions face trade-o s in support within their districts, because proposed environmental policies often increase the costs of polluting industries, but reduce the risks to residents from exposure to hazardous chemicals. Gilligan and Matsusaka's [10] ndings provide support for the hypothesis that logrolling leads representatives to spend more than their constituents would like. Durden et al [6] nd that legislators may be viewed as representing strong, well organized interest groups' preferences in exchange for direct and indirect political currency. Go and Grier [11] believe the question of whether legislators fail to represent their constituencies is currently unanswered by the literature, and cannot be answered by models making cross-sectional comparisons of the voting behavior of U.S. Senators.

# 2.4 Do voting outcomes lead to reductions in emissions?

Once the votes in Congress have been passed, the question of what e ect they have on environmental outcomes naturally arises. It should be understood that their e ect is really not a direct e ect, but rather a proxy for increased regulatory stringency at the local level based upon the preferences of the citizens for a tighter regulatory climate. A limited number

| of studies have analyzed the e ect of voting on environmental outcomes, but have only done |  |
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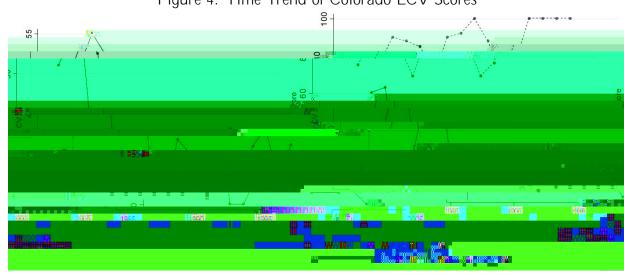


Figure 4: Time Trend of Colorado LCV Scores

are redrawn every decennial Census. Figure 5 shows the LCV scores for two Michigan

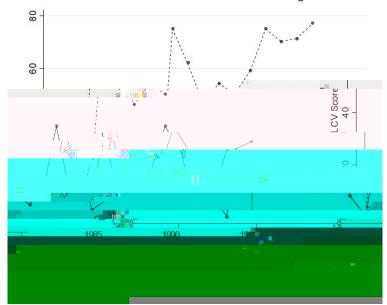


Figure 5: LCV Trends for Leelanau and Muskegon Counties, MI

#### 3.3 Constructing county-level measures

When constructing county-level measures of LCV scores for the U.S. House of Representatives, which are available at the congressional district level, two challenges arise: district lines are redrawn every ten years based on the decennial Census and a number of counties lie partially in multiple districts. The Census lists each congressional district and which counties are represented by that district. Most counties are completely contained within a single district, but there are 454 counties which belong to multiple districts. To illustrate, consider the hypothetical state in Figure 6 which has nine counties (A-I) and three congressional districts (1-3). Counties A, C, D, F, and H all lie within a single district, while counties B, G, and I lie in two districts, and county E lies in all three districts.

The Census provides the population of each county in a congressional district. Making a list of all counties, I record which districts are in each county and calculate what percentage

weight and population estimates. The temporal coverage of this data ranges from 1988 to 2002 and is available at the facility level. For the purpose of this paper I use pounds of stack air emissions aggregated to the county level. The number of TRI reporting facilities is provided by the RSEI program used to obtain data on emissions.

#### 3.5 Additional Data

Per capita income data were obtained from the Bureau of Economic Analysis [23] and population density data were obtained jointly from the U.S. Census Bureau [5] and the Risk-Screening Environmental Indicators [1]. Both were available at the county level annually from 1988 to 2006.

## 4 Empirical Speci cations

The primary objective of the empirical model is to examine what e ect congressional voting on environmental policies has on toxic emissions at a local level. With that focus in mind, if toxic releases are to decrease, the second objective of the model is then to identify whether it is due to facilities leaving the county or shutting down because of increased regulatory stringency (extensive margin) or whether rms reduce their emissions by decreas [(P)27etetomcpuculatiit

The most similar empirical specication to this study is the one used by Terry and Yandle (T-Y) [26] in an attempt to identify a relationship between LCV scores and toxic releases (TRI). However, there are key differences between the two studies<sup>4</sup>. T-Y conduct their study at the state level while this study is conducted at the county level. T-Y use the average voting records of the two U.S. Senators in each state and this study uses voting records from U.S. Representatives constructed at the county level. The T-Y study is a cross-sectional analysis and this study takes advantage of panel data. While T-Y have a larger number of control variables than I do in this study, it is necessary when conducting a cross-sectional analysis to include as many relevant variables (time-variant and time-invariant) as possible, otherwise the estimation will sufer from omitted variable bias.

The advantage of panel data is that time-invariant variables can be differenced out using a rst-differences model or time demeaned using a fixed-effects model, which greatly reduces the required number of variables for estimation, while not leading to omitted variable bias. That being said, there are still time-variant variables that I feel would be relevant to this study, but I was unable to obtain at this time. Assuming no serial correlation in the error terms, I use a fixed-effects estimation strategy. In the absence of serial correlation, fixed effects are more efficient than first-difference estimation. Generally, first-differences would be employed when a lagged value of the dependent variable is used as a regressor.

One concern with estimation is the potential endogeneity between LCV and TRI emissions. While it is possible that a higher LCV score will lead to a reduction in emissions, it

<sup>&</sup>lt;sup>4</sup>Terry and Yandle use LCV scores as one of a number of key explanatory variables. Their study does not place the primary focus on LCV scores

also seems reasonable to assume that higher emissions levels could cause greater concerns about pollution and therefore higher LCV scores. T-Y also recognize this potential identi - cation issue and they use the 1988 average of Senators LCV scores to explain TRI in 1992 (a four-year lag). In an attempt to identify the causal relationship between LCV scores (or more precisely the standards for which they proxy) and pollution, I treat previous years' LCV scores as the independent variable to test whether there is an e ect on the current level of pollution, since current pollution should not have any causal e ect on LCV scores in years prior to the current time period. Following that line of reasoning, I construct one- to ve-year lagged LCV scores for at least 10 years in order to explain the e ect of these scores on TRI emissions as well as how long before these policies would be e ective. I construct a 15-year panel data set which includes the years 1988-2002 and includes the top fty percent of TRI emitting counties, due to the large number of counties with zero emissions (743 counties) over the fteen year period. The dependent variable is total pounds of stack air emissions from the TRI. The key explanatory variable is the county-level measure of LCV scores, which has been constructed as previously described.

To estimate the e ect of pro-environment voting on overall toxic releases using an ordinary least squares xed-e ects framework, I estimate the parameters of the following equation

$$TRI_{it} = {}_{0} + {}_{1}LCV_{it} {}_{I} + \mathbf{X}_{it} + {}_{1}d1989_{t} + ::: + {}_{14}d2002_{t} + {}_{i} + {}_{it}$$
(4.1)

where  $TRI_{it}$  represents the measure of total pounds of TRI stack air emissions in county i in year t.  $LCV_{it-1}$  is the pro-environment voting score for county i in year t / where  $1 \ 2 \ f1; :::; 5g$  denotes the year lag.  $\mathbf{X}_{it}$  is a matrix of control variables which includes

population density and per capita income. To control for year e ects that a ect all counties, I include  $d1989_t, \ldots, d2002_t$  as dummy variables for years 1989-2002. The term  $_i$  is the county xed e ects, containing all factors within a given county that do not vary over time. To remove  $_i$ , I use time demeaning which is the xed e ects transformation model.  $_{it}$  is the idiosyncratic error term.

If toxic releases are decreasing as a result of higher LCV scores, the second objective of the empirical model is to identify whether this decrease is due to facilities leaving the county or shutting down because of increased regulatory stringency (extensive margin) or whether rms reduce their emissions by decreasing production or installing or upgrading abatement technology because of increased regulatory stringency (intensive margin). The second part of the model combines two speci cations to analyze the e ect of pro-environment voting on the number of TRI reporting facilities per county as well as per facility emissions. The panel data set is the same as above using years 1988-2002 and the top fty percent of TRI emitting counties, however, in these speci cations the dependent variables are the number of TRI reporting facilities per county and pounds of TRI stack air emissions per facility per county. To nd out whether toxic releases are decreasing due to fewer facilities (extensive margin) or lower per-facility emissions (intensive margin), I estimate the parameters of the following two equations

$$Facilities_{it} = {}_{0} + {}_{1}LCV_{it} {}_{I} + \mathbf{X}_{it} + {}_{1}d1989_{t} + ::: + {}_{14}d2002_{t} + {}_{i} + {}_{it}$$
(4.2)

Emissions=Facility<sub>it</sub> = 
$$_{0} + _{1}LCV_{it} + X_{it} + _{1}d1989_{t} + \dots + _{14}d2002_{t} + _{i} + _{it} (4.3)$$

using an ordinary least squares xed-e ects framework where  $Facilities_{it}$  represents the

rm exodus (extensive margin) or a reduction in per-facility emissions (intensive margin), and 3.) to compare the results from county-level analysis and state-level analysis. The estimation results of Equation 4.1 for both county- and state-level measures are summarized in Table 1 for the one to three year lags and Table 2 for the four to ve year lags.

Table 1: First Speci cation - Aggregate TRI Stack Air Emissions

|                       | Total Pounds<br>(County)   | Total Pounds<br>(State)   | Total Pounds<br>(County)   | Total Pounds<br>(State)    | Total Pounds<br>(County)  | Total Pounds<br>(State)  |
|-----------------------|----------------------------|---------------------------|----------------------------|----------------------------|---------------------------|--------------------------|
| LCV <sub>t</sub> 1    | -1,251.759*<br>[557.0398]  | -31,085.81<br>[42,501.11] |                            |                            |                           |                          |
| LCV <sub>t 2</sub>    |                            |                           | -1,280.403*<br>[549.6646]  | -19,852.57<br>[44,346.65]  |                           |                          |
| LCV <sub>t</sub> 3    |                            |                           |                            |                            | -1,157.845*<br>[553.8702] | -52,967.85<br>[46,287.1] |
| Population<br>Density | -1,187.114**<br>[364.5939] | 770,888.9**<br>[152557.4] | -1,088.286**<br>[349.8771] | 775,656.2**<br>[160,789.6] | -932.9411**<br>[348.6618] | 799,375.8**<br>[174,099] |

score from 0 to 100 would be expected to decrease toxic releases by 120,000 pounds within one to three years.

With a closer look at the data, I identify counties that experience an increase of at least 50 LCV points to see if the model's prediction would hold true. Three di erent Michigan counties that the criteria of er some veri cation. A look at the emissions levels of Alpena County shows that it is one of the high-emission counties in the state with an average of 1,222,064 pounds of toxic emissions per year. Figure 7 shows two time trends for Alpena County: LCV scores and the level of TRI emissions over the fiteen-year period. Alpena

Figure 7: LCV and TRI Trends for Alpena County, MI

The second part of the model decomposes the extensive and intensive margins. From the parameter estimation of Equation 4.2, the number of TRI reporting facilities is predicted to decline as a result of higher LCV scores. From Table 3, the coe cients of  $LCV_{it}$ , for the county-level data are negative and statistically signicant at the 1% level for 12 f1,2g which would suggest that rms are exiting the counties or shutting down because of increased regulatory stringency. However, the magnitude of the coe cients suggests that LCV is not enough of a factor to cause facilities to exit or shut down at the county level. A one-point increase in LCV score leads to 0.006 fewer facilities at the county level and 0.27 fewer facilities at the state level. This does not seem to have a signicant elect at the county level since the maximum increase in LCV score from 0 to 100 would only lead to a 0.6 facility decrease. This is not too surprising given that LCV is an indirect measure of regulatory stringency. Also, the average number of facilities in a county is about 8 and the average change in facilities is -0.006. At the state level there seems to be a small meaningful e ect on facility numbers since the coe cient on  $LCV_{jt}$  , is negative and statistically signicant at the 5% level for I= 1. The model predicts that the maximum increase in LCV score from 0 to 100 in state jwould lead to a decrease of 27 facilities.

From the parameter estimation of Equation 4.3, the lack of statistical signicance with the exception of the two-year lag on LCV suggests that votes have an election per facility TRI emissions and that it takes about two years for these to take elect (Table 5). It appears that

at the state level is consistent with many of the studies on rm location decisions which nd that strict environmental regulation induces rms to locate in or shift in production to less stringent counties. Given the limitations on rm data it is not possible to identify whether the facilities simply shut down or whether they relocated since only the number of TRI reporting facilities is used. At the county level it may be an indication that there is actual reduction of emissions taking place and not simply a redistribution.

#### 6 Conclusion

The primary objective of this paper is to examine what e ect congressional voting on environmental policies has on toxic emissions at a local level. If toxic releases are decreasing, the second objective of the model is then to identify whether it is due to facilities leaving the county or shutting down because of increased regulatory stringency (extensive margin) or whether rms reduce their emissions by decreasing production or installing or upgrading abatement technology because of increased regulatory stringency (intensive margin). The third objective of the model is to run the same empirical analysis using both county-level and state-level data to compare the results in order to see if anything is to be gained from taking advantage of within-state variation.

I use county-level measures of pro-environment voting from the U.S. House of Representatives as a proxy for regional heterogeneity in preferences of citizens for more or less regulation. U.S. Representatives are more accountable to their constituents because of the frequency of re-election and because they represent a smaller geographical region. Even though constructing county-level measures of voting scores requires a degree of approxima-

tion in counties that lie partially in multiple districts, the fact that county lines do not change with the decennial Census allows for measures of emissions activity in speci c locations over time using panel data spanning more than ten years.

People living in low-income and minority communities are the most directly a ected by toxic releases and prefer more regulation since they cannot a ord to self-select into cleaner neighborhoods. They are also the groups that are least likely to engage in collective action against polluters or to lobby politicians to make their voices heard. Assuming that legislators take di erent groups preferences into account when deciding how to vote on di erent policies, if they are voting more pro-environment at the national level, this indicates that there is overwhelming pressure from those groups at the local as well.

The results show that pro-environment voting scores at the county level are associated with a reduction in TRI emissions within one to three years after the voting has occurred. Signi cance at the county level but not at the state level would suggest that changes are taking place in emissions across counties within states rather than across states because LCV scores represent local preferences and not preferences for the state as a whole. It appears that rm exodus is the cause of the reduced emissions at the state level, but at the county level very few rms are exiting as a result of the voting pattern. This conclusion of rm exodus at the state level is consistent with many of the studies on rm location decisions which no that strict environmental regulation induces rms to locate in or shift in production to less stringent counties. At the county level it may be an indication that there is actual reduction of emissions taking place and not simply a redistribution. To the best of my knowledge, this

paper is the rst to construct county-level measures of pro-environmental voting from the U.S. House of Representatives and use them as a proxy for citizen preferences for regulation to determine their e ect on toxic releases at a local level.

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Table 2: First Speci cation (continued) - Aggregate TRI Stack Air Emissions

|                    | Total Pounds<br>(County) | Total Pounds<br>(State)   | Total Pounds<br>(County) | Total Pounds<br>(State)   |
|--------------------|--------------------------|---------------------------|--------------------------|---------------------------|
| LCV <sub>t</sub> 4 | -292.1132<br>[584.833]   | -39,842.36<br>[50,135.78] |                          |                           |
| LCV <sub>t</sub> 5 |                          |                           | 351.4845<br>[608.3921]   | -24,848.22<br>[53,974.59] |

Table 4: Second Speci cation (continued) - Number of TRI Reporting Facilities

|                       | Facilities             | Facilities             | Facilities              | Facilities              |
|-----------------------|------------------------|------------------------|-------------------------|-------------------------|
|                       | (County)               | (State)                | (County)                | (State)                 |
| LCV <sub>t</sub> 4    | .0014742<br>[.0012465] | .1270258<br>[.1224663] |                         |                         |
| LCV <sub>t</sub> 5    |                        |                        | .0014491<br>[.0011525]  | .1236795<br>[.1124741]  |
| Population<br>Density | 0006295<br>[.0007678]  | 4719102<br>[.464494]   | .0002882<br>[.0006999]  | 2117155<br>[.4272333]   |
| Per Capita<br>Income  | 000256**<br>[.0000202] | 0069915<br>[.0017508]  | 0001721**<br>[.0000191] | 0050948**<br>[.0016289] |
| Constant              | 14.70237**             | 513.7995**             | 12.44256**              | 406.6307**              |



Table 7: Summary Statistics

Top 50% of Emitting Counties (1988-2002)

| <u> </u>                 |        |            |           |             |            |
|--------------------------|--------|------------|-----------|-------------|------------|
| Variable                 | Obs    | Mean       | Std. Dev. | Min         | Max        |
| LCV score                | 23,444 | 36.89369   | 29.06418  | 0           | 100        |
| TRI pounds (stack air)   | 23,505 | 897,266    | 2,731,773 | 0           | 1.19e + 08 |
| TRI reporting facilities | 23,505 | 8.5612     | 18.1613   | 0           | 486        |
| Per-facility emissions   | 23,505 | 182,606.3  | 945,129.4 | 0           | 6.50e + 07 |
| Per-capita income        | 23,505 | 19,923.31  | 5,601.505 | 7,380       | 61,759     |
| Population density       | 23,505 | 132.4999   | 555.7981  | 0           | 13,582     |
| LCV score                | 21,866 | 9736862    | 17.65127  | -92         | 92         |
| TRI pounds               | 21,938 | -15,957.49 | 892,607.9 | -3.39e + 07 | 2.35e + 07 |
| TRI facilities           | 21,938 | 0062905    | 1.828708  | -53         | 39         |

# States (1988-2002)

| Variable                      | Obs | Mean       | Std. Dev.  | Min         | Max        |
|-------------------------------|-----|------------|------------|-------------|------------|
| LCV score (U.S House average) | 750 | 46.30506   | 24.27527   | 0           | 100        |
| TRI pounds (stack air)        | 750 | 2.83e + 07 | 2.90e + 07 | 37,296      | 1.44e + 08 |
| TRI reporting facilities      | 750 | 284.128    | 272.5141   | 3           | 1252       |
| Per-facility emissions        | 750 | 118,420.4  | 152,150    | 2,491.004   | 1,691,254  |
| Per-capita income             | 750 | 22,809.8   | 5,385.068  | 11,561.27   | 42,920.69  |
| Population density            | 750 | 66.80337   | 92.02104   | .5229201    | 446.4016   |
| LCV Score                     | 700 | 637406     | 10.86083   | -56         | 43         |
| TRI pounds                    | 700 | -501241.2  | 8,516,750  | -4.12e + 07 | 9.38e + 07 |
| TRI facilities                | 700 | .1542857   | 22.13909   | -119        | 131        |