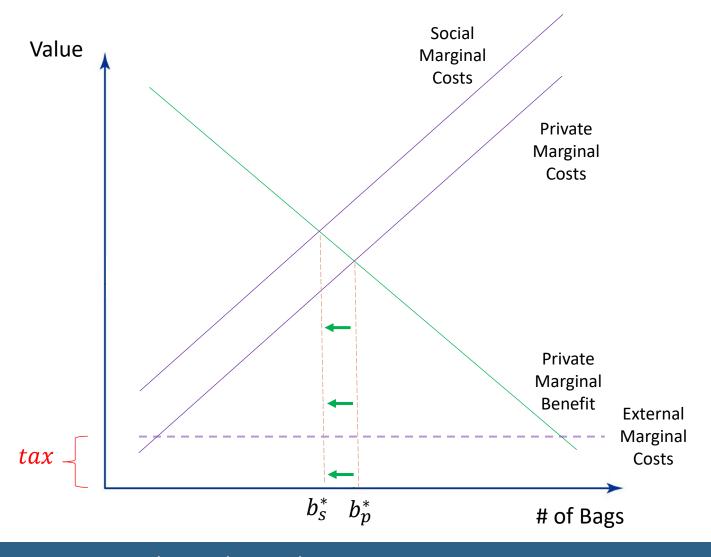
Lecture 3: Market Failures: Examples and in Practice

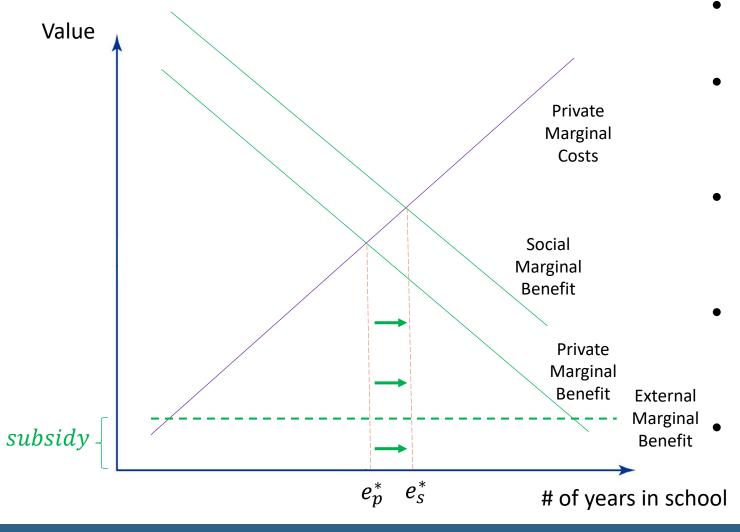
Prof. Parthum Environmental Economics Econ 475

An Example: The Market for Plastic Bags



- What's missing?
 - The external marginal cost
- Adding the external cost to the private recovers the (true) social marginal cost
- The socially optimal number of bags is *fewer* than the private. The private market *overprovides* bags
- What is a possible fix?
 - A tax equal to the external marginal cost

An Example: The Market for College Education



- In education, the externality is a "good"
- The private market would underprovide public education
- A government *subsidy* can help correct this market failure
- In this context, is the external marginal benefit actually fixed (flat)?
 - Try drawing an example of what you think external marginal benefits are

Market Failures

There are many different types or root causes of market failures. Think of some examples under each of these categories.

- 1. Information Asymmetries
 - One party in a transaction knows more than the other
- 2. Market Structure/Power
 - One party can influence the market equilibrium
- 3. Public Goods
 - Nonrival and nonexcludable
- 4. Externalities
 - Private actions have unintended effects

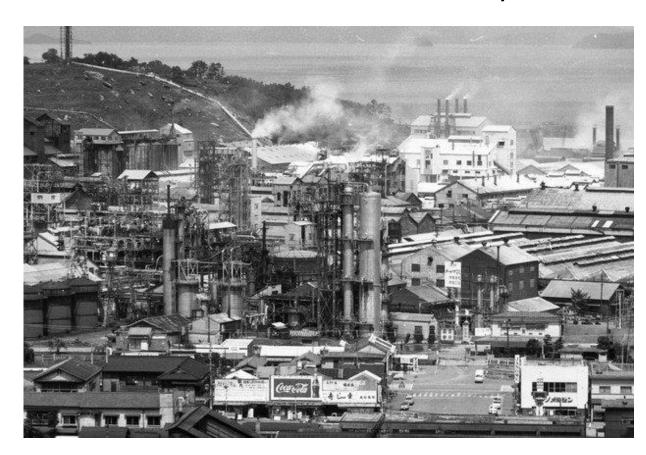
The Cuyahoga River Caught Fire at Least a Dozen Times, but No One Cared Until 1969

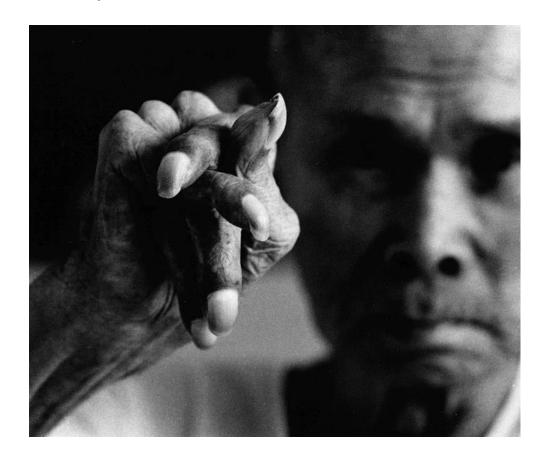




Source: <u>history.com</u>

Minamata disease, 1956: the release of methylmercury in the industrial wastewater from a chemical factory owned by the Chisso Corporation



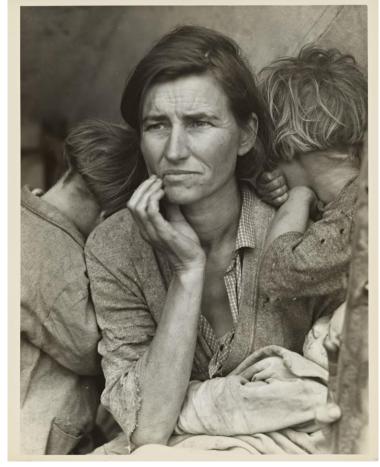


Source: **Blind Magazine**

The Dust Bowl

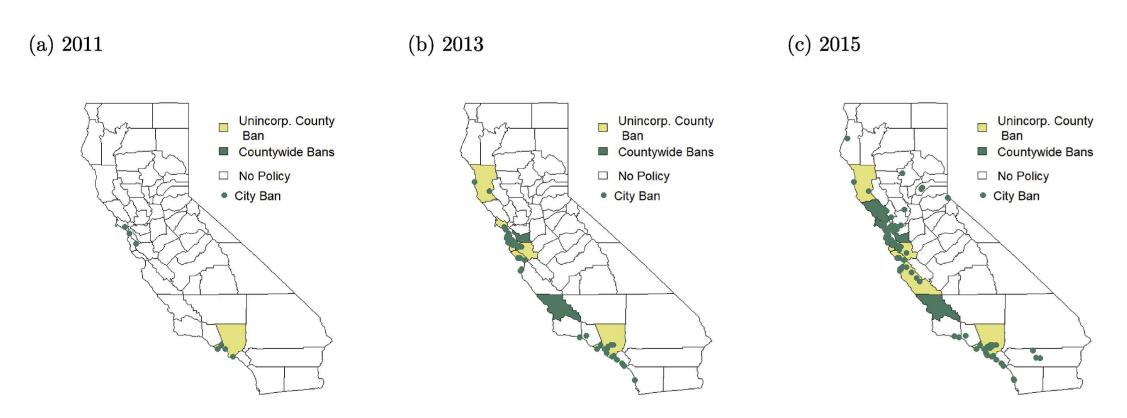
As one "black blizzard" hit after another, harmful dust particles accumulated in people's lungs, causing hundreds of deaths and sickening thousands.





Source: <u>history.com</u>

By: <u>Dr. Rebecca Taylor</u>



Note: The local governments of unincorporated counties and incorporated cities can pass ordinances to regulate disposable carryout bags. City-level policies are depicted with dark green circles. Unincorporated county policies are shaded in light yellow. Countywide policies—where all unincorporated areas and all cities in a county implement DCB regulations—are shaded in dark green.

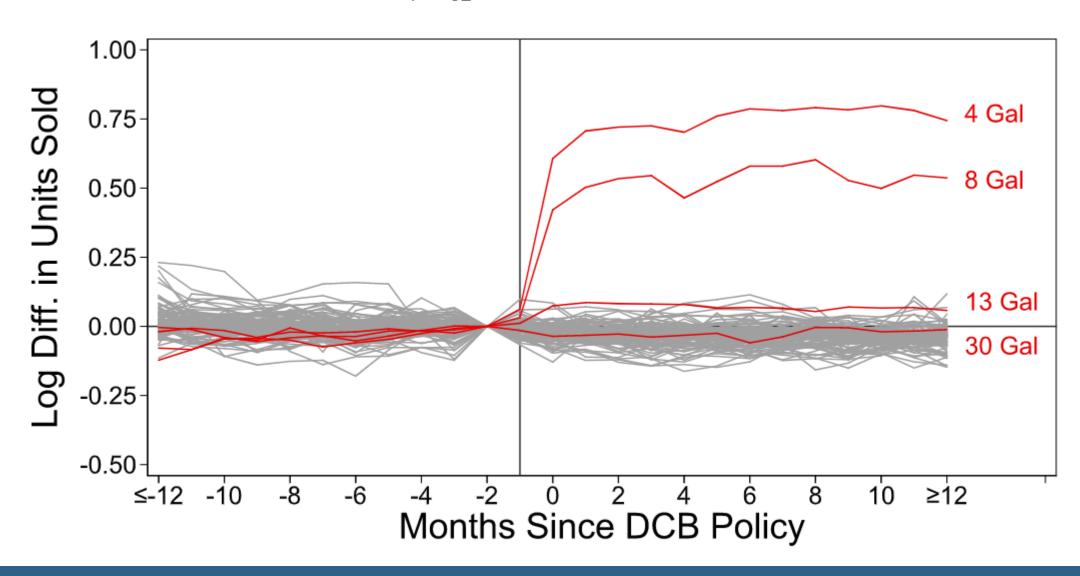
By: <u>Dr. Rebecca Taylor</u>

$$Y_{sjm}^{B} = \sum_{l=-12}^{12} \beta_l D_{l,jm} + \theta_{sj} + \delta_m + \epsilon_{sjm}$$

where Y_{sjm}^B is the outcome variable for store s in jurisdiction j and month-of-sample m with respect to bag product group B, θ_{sj} is a vector of store fixed effects, and δ_m is a vector of month-of-sample fixed effects. $D_{l,jm}$ is a dummy variable equaling one if jurisdiction j in month m implemented a DCB policy l months ago, with l=0 denoting the month of implementation.

The β_l vector is the parameter of interest, as it traces out the differences in outcomes from before the DCB policies to after. I hypothesize that sales of trash bags deemed by customers to be substitutes for plastic carryout bags will increase. Thus, for any product group B that is a substitute for plastic carryout bags, I would expect the β_l coefficients in the post-policy period to be greater than zero.

$$Y_{sjm}^{B} = \sum_{l=-12}^{12} \beta_l D_{l,jm} + \theta_{sj} + \delta_m + \epsilon_{sjm}$$



By: Dr. John Morehouse and Dr. Ed Rubin

Abstract: In federalist systems, local governments can maximize local welfare by exporting locally produced negative externalities. We empirically substantiate this externality-export strategy for air pollution using historical power-plant siting, administrative borders, and prevailing wind directions. Using a simple, nonparametric test, we show that decision-makers disproportionately sited coal-fueled plants to reduce counties'/states' downwind pollution exposure. Natural-gas-fueled plants—lower polluters—did not follow this strategy. We then illustrate the extreme exportability of coal plants' pollution: within 6 hours, 50% of coal plants' emissions leave their source states—and 99% depart source counties. These results highlight how local strategic responses challenge federalist systems.

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$$p$$
-Value $(n_s) = \mathbf{P}(X \ge n_s; n = N_T, p = 0.5) = \sum_{x=n_s}^{N_T} {N_T \choose x} 0.5^{N_T}$

We operationalize this

test as an implementation of Fisher's Exact Test (Fisher, 1934; Fisher, 1935; Conover, 1971; Imbens and Rubin, 2015). Under a sharp (one-sided) null hypothesis of *no strategic siting to reduce downwind area*, the test statistic n_s (the number of plants for whom downwind area is less than upwind area) is distributed as a binomial distribution with size equal to the number of plants in the sample (N_T) and probability p = 0.5. Under this null, the expected share of plants whose downwind area is less than its upwind area is 50%. Consequently the p-value for a given test statistic is

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Figure 6: HYSPLIT trajectory and dispersion: Two example plants, January and July 2005

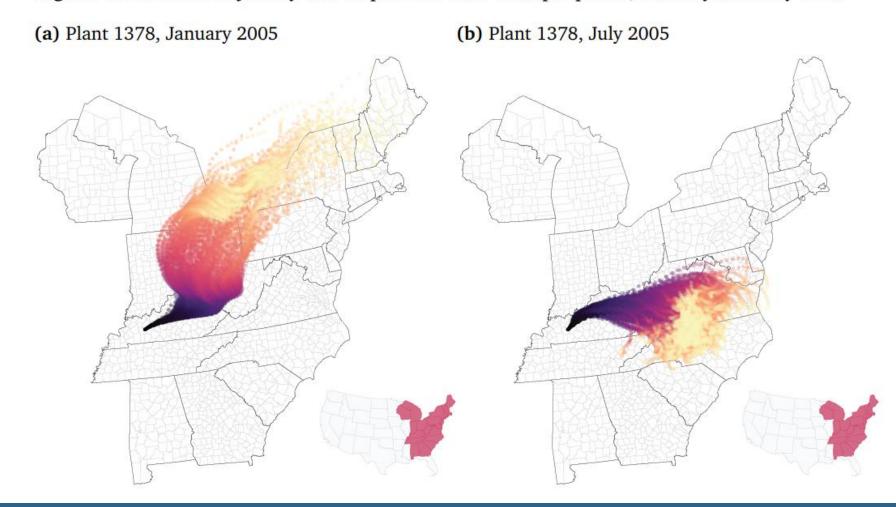


Table 1: Testing strategic siting: Upwind vs. downwind areas for coal and natural gas plants

	(1)	(2)
	Coal-fueled plants	Natural-gas-fueled plants
Panel A: Siting str	ategically within count y	7
Count	514	1,254
Count strategic	292	620
Percent strategic	56.81%	49.44%
Fisher's exact tes	st of H _o : In- county downw	rind area ≤ upwind area
	cent strategic: County] = 5	0%
<i>P</i> -value	0.0012	0.6641
Panel B: Siting str	ategically within state	
Count	514	1,254
Count strategic	277	574
Percent strategic	53.89%	45.77%
Fisher's exact tes	st of Ho: In-state downwir	d area ≤ upwind area
Under H_o : $E[Percent Theorem 1]$	cent strategic: State] = 509	6
<i>P</i> -value	0.0426	0.9987
Panel C: Siting str	ategically within both c	ounty and state
Count	514	1,254
Count strategic	179	314
Percent strategic	34.82%	25.04%
Fisher's exact tes	st of H₀: Downwind area ≤	upwind area in county and state
Under H_o : $E[Pero$	cent strategic: County ∧ Sto	nte] = 25%
P-value	< 0.0001	0.4978

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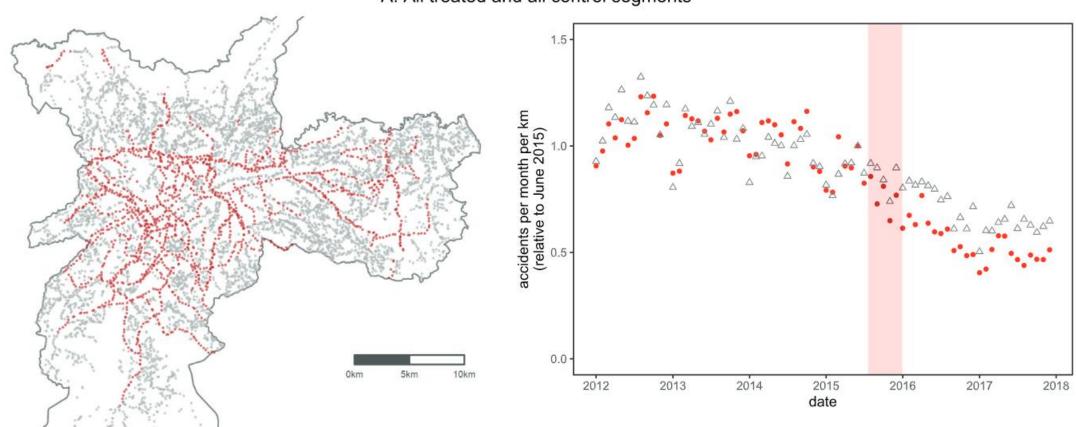
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By: <u>Dr. Amanda Ang et al.</u>

A: All treated and all control segments



By: Dr. Amanda Ang et al.

We use the following Poisson event study model²⁹ of accident counts to estimate of the impact of changing the speed limit:

$$\log (E(y_{it})) = \alpha_i + \beta X_t + \left(\sum_{q=1}^6 \gamma_q D_{it}^q\right) + \zeta C_{it} + \eta C_{it} SLR_{it}$$
 (1)

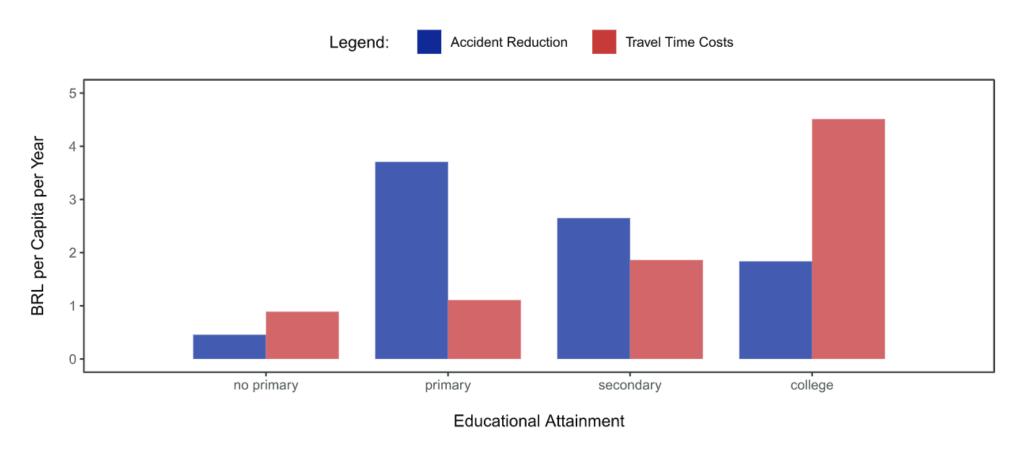
where y_{it} is the number of accidents on segment i during month t, α_i is a segment fixed effect that captures the time-invariant component of accidents on each segment. X_t is a vector of time-varying controls that are measured at the city level. It includes controls for secular changes in driver behavior across the time series with a linear time trend and two covariates that capture aggregate changes in driving behavior during the period: (1) the log of fuel sales in the State of São Paulo and (2) the log of the total number of speed monitoring cameras in São Paulo. The variable D_{it}^q is an indicator for the number of q quarters relative to i's initial treatment (q = 1 is the quarter of initial treatment). We cap observations on all segments at the sixth relative quarter, which ensures that all segments have the same time in treatment. The sixth quarter is our longest-term estimate and reflects our best estimate of the longer-run effect of the policy. Standard errors are clustered at the road level (202 clusters).

The primary coefficients of interest are the γ_q terms, which measure changes in the number of accidents on a treated segment in

each of the quarters following the treatment of segment (i).³⁰ C_{it} is an indicator for the presence of a speed monitoring camera on segment i during month t, such that ζ estimates the change in accidents on road segments where camera-based enforcement is initiated during the study period.³¹ The interaction term $C_{it} \cdot SLR_{it}$ is an indicator for whether the speed limit reduction policy occurred on a segment that also received camera-based enforcement, such that the coefficient η measures the interaction between the speed limit reduction and the onset of camera-based speed enforcement. Because the policy was partially reversed in January 2017, we restrict our sample to December 2016 to isolate the effects of the speed limit reduction, though we estimate and discuss an extended version of this model that includes the post-reversal period in Appendix B.

By: <u>Dr. Amanda Ang et al.</u>

A: Costs and Benefits of Speed Limit Reduction by Educational Attainment



By: Hemant Pullabhotla and Dr. Mateus Souza

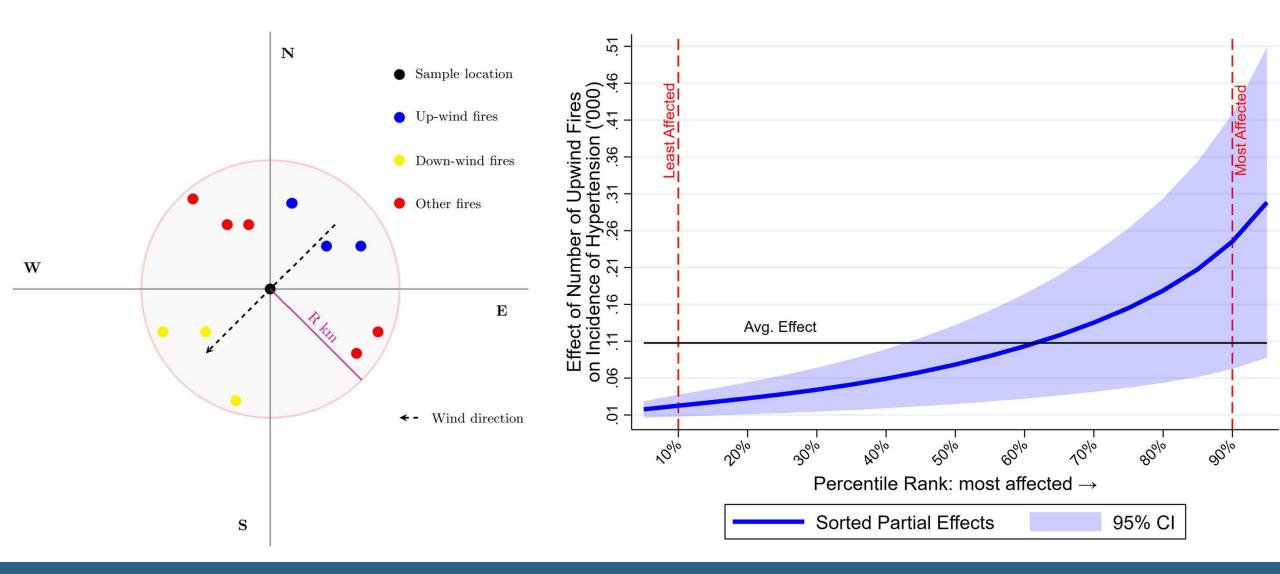
By: <u>Hemant Pullabhotla and Dr. Mateus Souza</u>

By: Hemant Pullabhotla and Dr. Mateus Souza

By: Hemant Pullabhotla and Dr. Mateus Souza

Air pollution from agricultural fires increases hypertension risk

By: Hemant Pullabhotla and Dr. Mateus Souza



Next Week

- Tuesday, Sept 6th is on a "Monday" Schedule
 - Wrapping up Module 1 with a class on positive versus normative notions in applied economics
- Wednesday, Sept 7th is a regularly scheduled class
 - Beginning Module 2 with econometrics and treatment effects
- Saturday, Sept 10th your first reflection post is due