

Air Pollution Social Costs Estimated With Chemical Transport Models

Peter J. Adams
Center for Atmospheric Particle Studies (CAPS)
Carnegie Mellon University

CEDM Annual Meeting
May 21, 2015

Air Pollution Often Dominates



Simultaneously Mitigating Near-Term Climate Change and Improving Human Health and Food Security

Science (2012)

Drew Shindell,^{1*} Johan C. I. Kuylenstierna,² Elisabetta Vignati,³ Rita van Dingenen,³

CH₄

BC Tech

Black carbon mitigation may have some benefits as a short-lived climate forcer
...but definitely has large benefits now for health (25x as big)

Valuation

Climate, billions \$US	331 ± 118	142 (+71/−106)
(\$US per metric ton CH ₄)	(2381 ± 850)	
Crops, billions \$US	4.2 ± 1.2	3.6 ± 2.6
(\$US per metric ton CH ₄)	(29 ± 8)	
Health, billions \$US	148 ± 99	3717 (+3236/−2563)
(\$US per metric ton CH ₄)	(1080 ± 721)	



$$\text{Social Cost} \simeq \left(\frac{\Delta \text{Mortality Rate}}{\text{by } \Delta \text{PM}_{2.5}} \right) \times (\text{Exposed Population}) \times (\text{Value of a Statistical Life})$$

- Requires
 - AQ model (emissions \rightarrow concentrations)
 - epidemiology “concentration-response” (mortality)
 - exposure (population density, spatial resolution)
 - value of a statistical life
- Frequently reported as \$ per ton emissions
- “Marginal emissions” (\$/ton)
 - assumes linear response; appropriate for small Δ

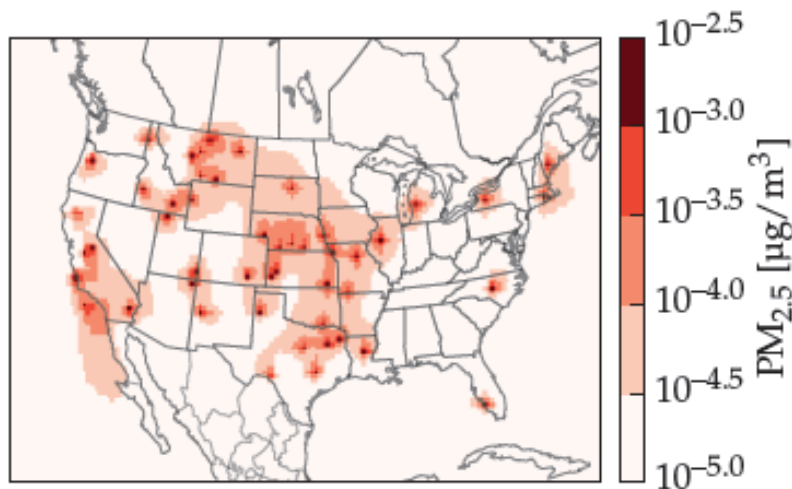


- Dispersion model based estimates
 - ***lack rigorous treatment of secondary (i.e. not directly emitted) PM_{2.5} formation***
 - not recommended for long distance transport
 - e.g. CRDM → APEEP, COBRA
- Chemical transport models (CTMs)
 - ***computationally intensive***
 - one-off simulations of specific policies (e.g CSAPR)
 - Response Surface Model (RSM), papers by Fann
 - simulations tend to aggregate emissions nationally, regionally, and/or by sector
- Reduced-form models inherit problems of AQ model



- Provide marginal social costs (\$/ton) that are
 - simple, easy-to-use
 - computationally efficient
 - **based on state-of-science CTM**
 - **approx county-scale spatial resolution**
 - seasonal variability
- Brute force: ~9,000 CPU-years
- Resulting tool is EASIUR
 - (Estimating Air quality Social Impacts Using Regression)

Partial Solution: “Tagged” Simulations



PSAT simulation with 50 tags for July
(6.6 kg EC/day were added at each location)

CAMx Particulate Source Apportionment Technology (PSAT)

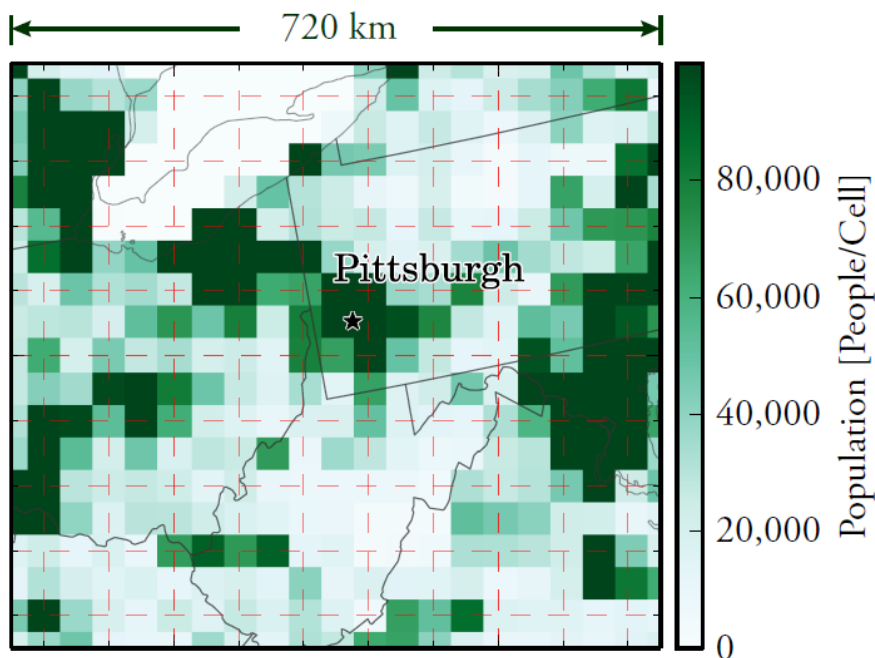
- tracks the contributions of multiple emission sources separately on $\text{PM}_{2.5}$ concentrations in all receptor locations.
(Wagstrom et al, 2008; Koo et al, 2009)
- reduces computational costs substantially by a factor of ten, creating a large dataset for this study.

Provides large data set of locations
... still not every county in US

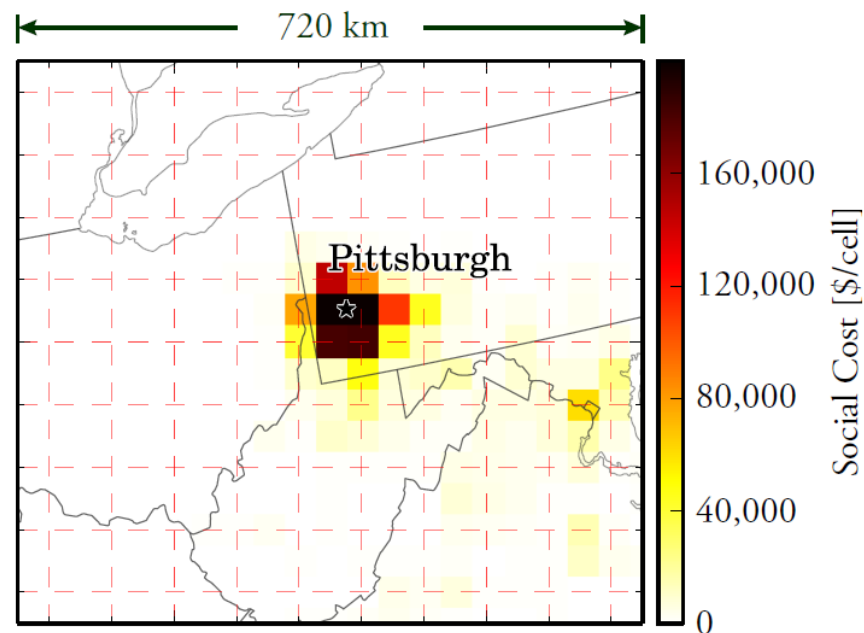


Estimating Air pollution Social Impacts Using Regression (EASIUR)

$$\text{Social Cost} \simeq \left(\frac{\Delta \text{Mortality Rate}}{\text{by } \Delta \text{PM}_{2.5}} \right) \times (\text{Exposed Population}) \times (\text{VSL})$$



Population around Pittsburgh



Social Cost of Pittsburgh Elemental Carbon

3 t EC/day emitted in Pittsburgh

Research Design

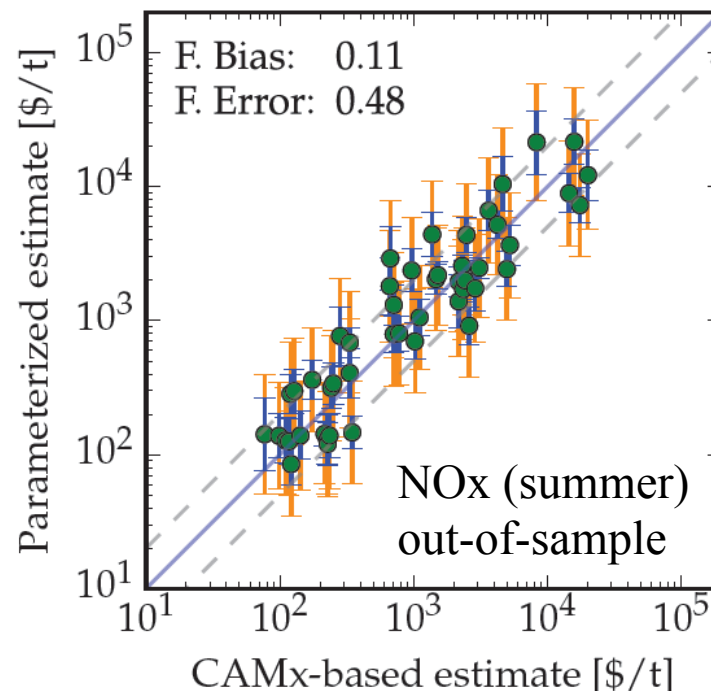
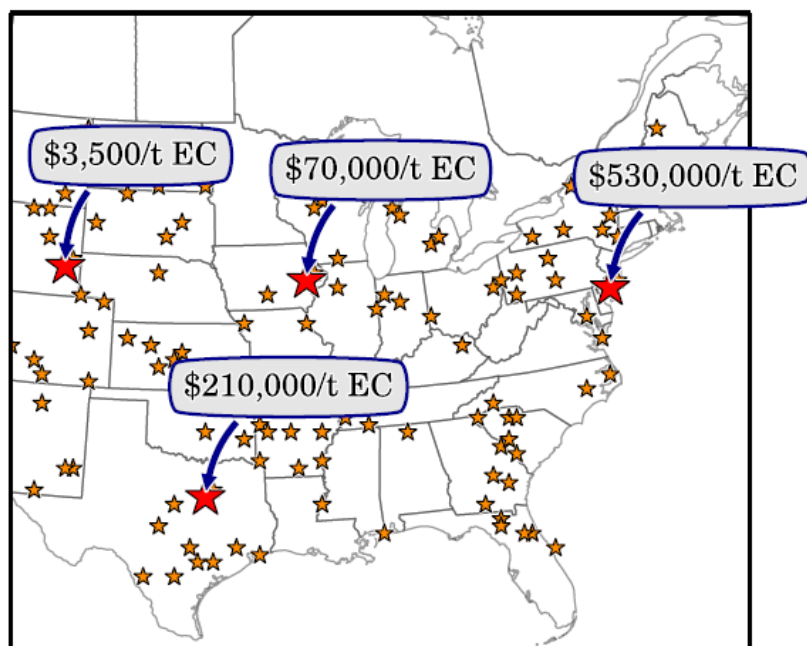


Our goal is to derive:

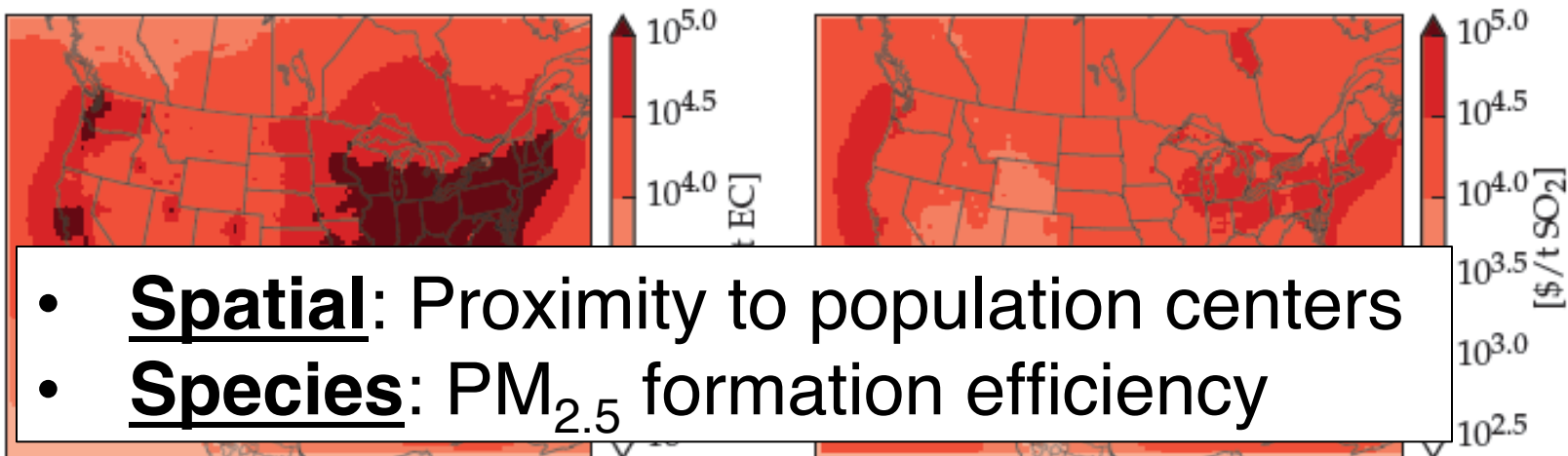
Per-tonne Social Cost [\$/t] = f (Population, Atmospheric Variables)

Intake Fraction [ppm] = g (Population, Atmospheric Variables)

3. Calculate per-tonne social costs and intake fractions for the 100 locations.

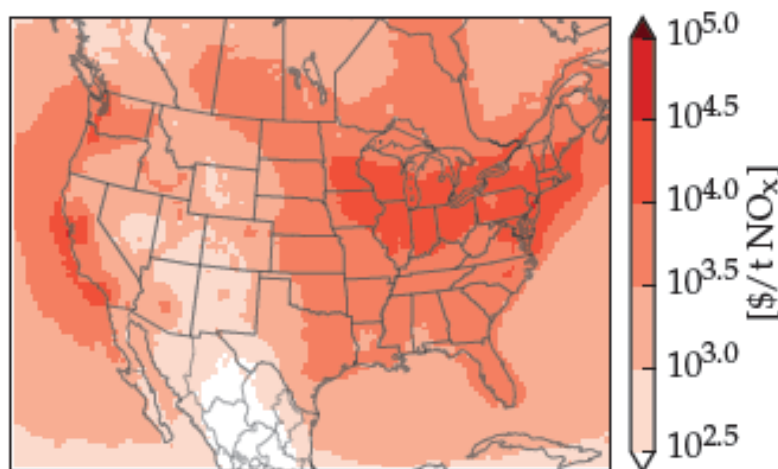


EASIUR Results (per tonne)

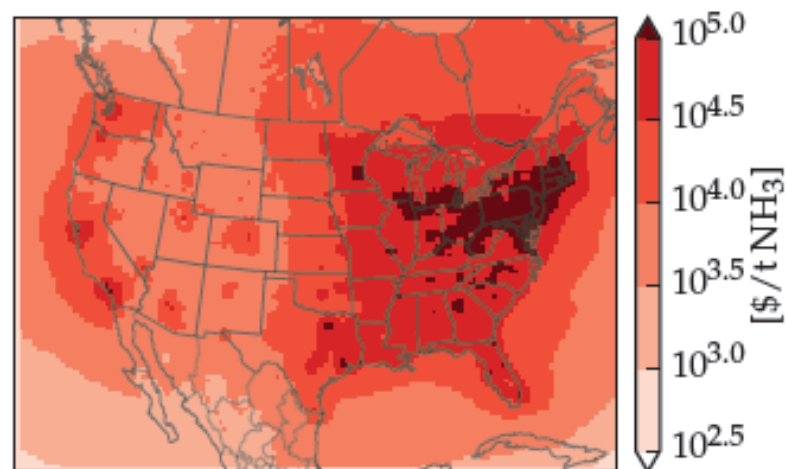


(a) EC: \$170,000/t EC

(b) SO₂: \$27,000/t SO₂



(c) NO_x: \$9,700/t NO_x

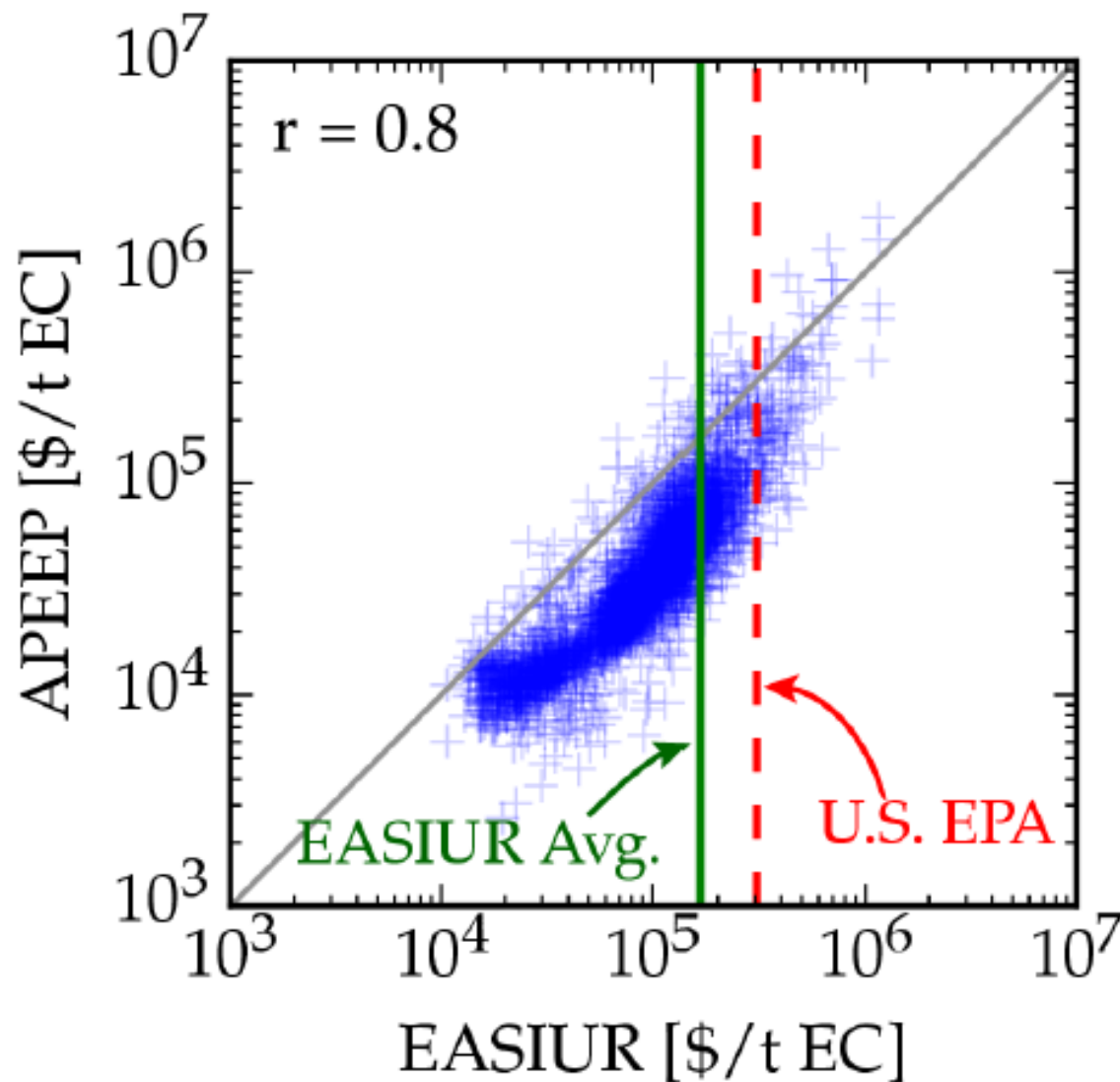


(d) NH₃: \$46,000/t NH₃

EASIUR vs APEEP: Spatial Correlations?



EC
inert
 $r = 0.8$

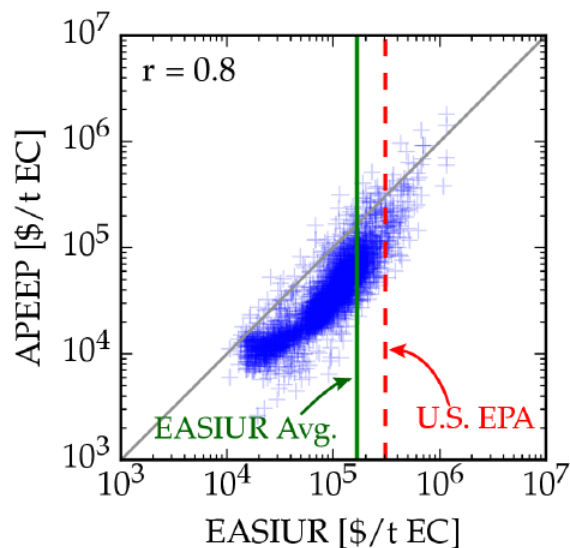


each +
represents
one county

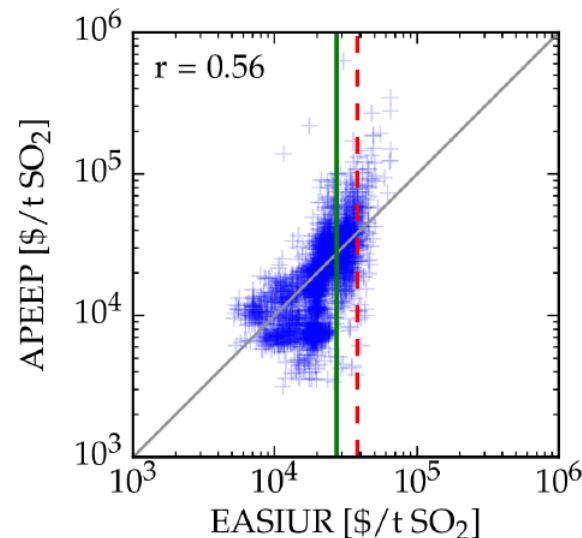
EASIUR vs APEEP: Spatial Correlations?



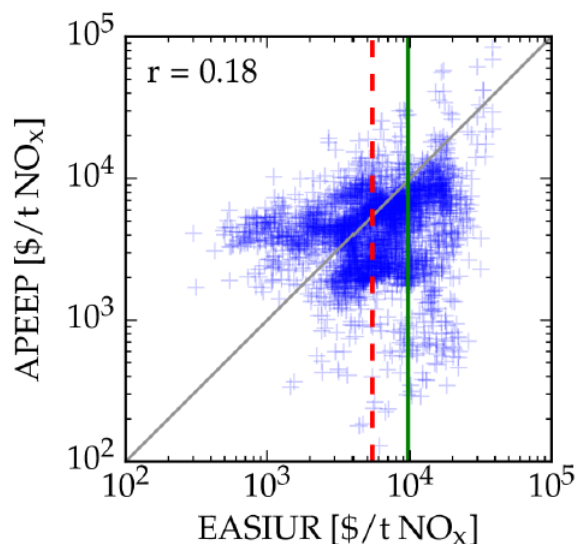
EC
inert
 $r = 0.8$



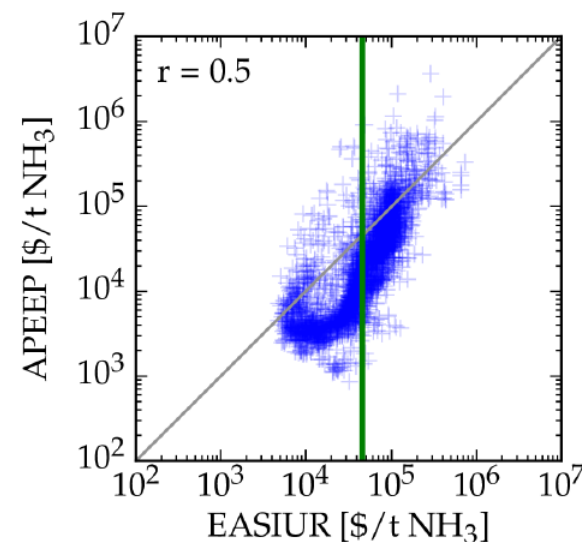
SO2
chem
 $r = 0.56$



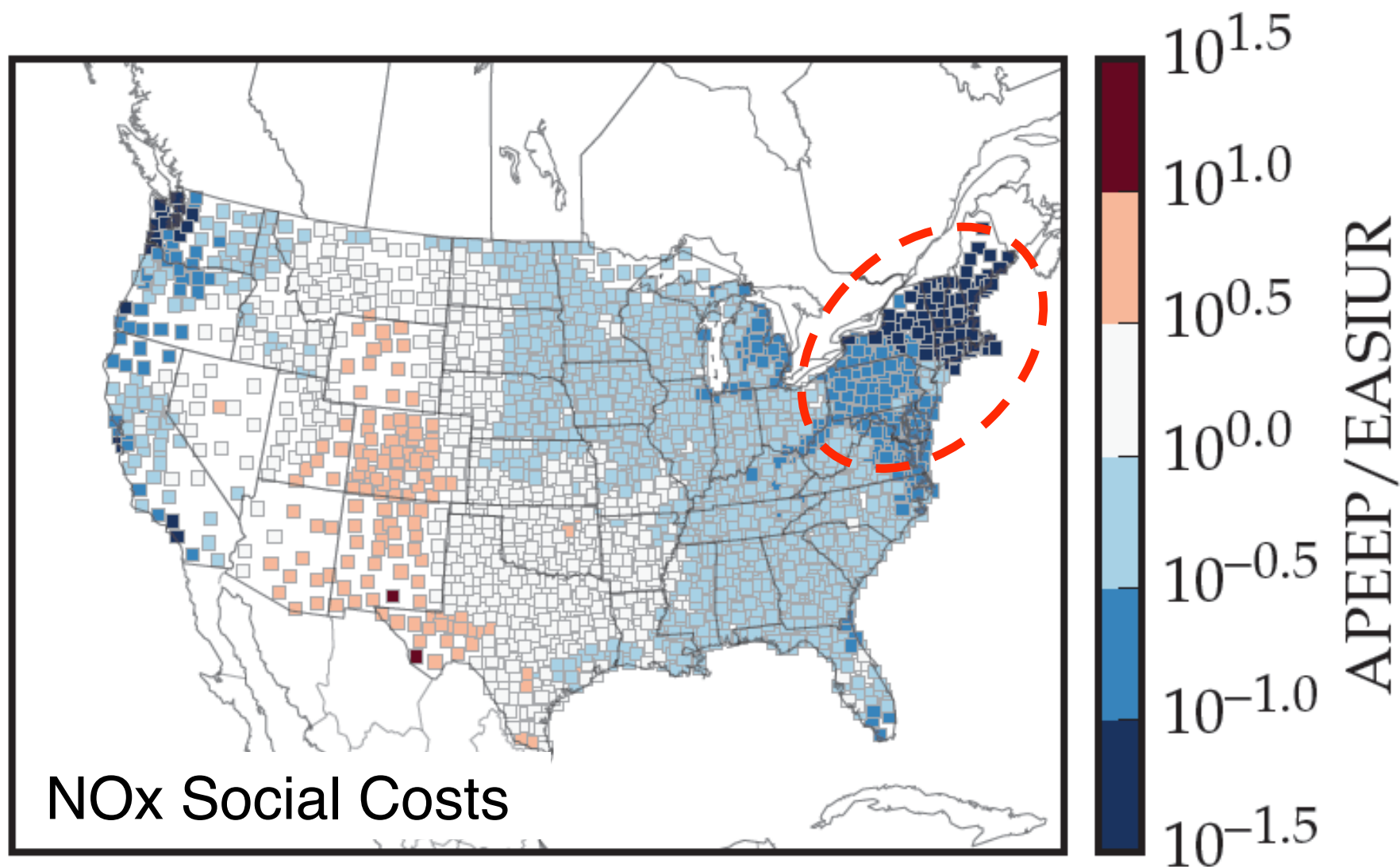
NOx
chem /
thermo
 $r = 0.18$



NH3
thermo
 $r = 0.5$



EASIUR vs APEEP Comparison





- CTMs: slow, for air quality experts
- Previous tools either
 - not based on rigorous model
 - lacked spatial resolution
- Built “EASIUR” to address deficiencies
- Discrepancies/uncertainties between tools
 - factor of ~ 2 , nationally averaged, most species
 - factor of up to 10, difficult species over some regions
 - bias with APEEP seems to be systematic for species with complex atmospheric behaviors
- Extensions: organic $\text{PM}_{2.5}$