HyADS: A tool for estimating nationwide exposures to emissions from large numbers of sources

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The Team



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Motivation

- We know:
 - The US spends tens of billion \$/yr regulating air quality
 - Regulations manifest as discrete actions on individual point sources
- We want to know:
 - Can we establish <u>direct</u>
 <u>epidemiological evidence</u> that we
 are healthier because of the
 regulations?



Connecting power plants to people with HyADS

- HYSPLIT simulates dispersion of 100 parcels from each stack
 - Parcels tracked for 10 days
 - Omit near-source impacts
 - Omit parcels above planetary boundary layer
 - Parcels not resuspended
- Repeat at 6 hour intervals daily
- NCEP Reanalysis meteorology
- Locations aggregated to 12km 3D grid with monthly boundary layer as height
- Weight by monthly SO₂ emissions

2005 HyADS exposure for facility 6113, Unit 1



- Accessible through hyspdisp R package www.github.com/lhenneman/hyspdisp
- Thanks to hard work by Maja Garbulinska, will soon be updated to disperseR

Connecting people to power plants with HyADS

- Reduced complexity
 - Simplified chemistry/transport
 - Identifies areas impacted, not concentration
- Increased scalability
 - Source receptor matrix from ~1k sources
 - Estimate source impact changes from interventions
 - Develop counterfactual scenarios
 - 1 year run in 1 week (using R!)



HyADS (2005)



Application-specific evaluations



Source impacts on specific geographies

- Geos-CHEM adjoint sensitivities
- State-level, averaged PM_{2.5} from emissions perturbations anywhere in the 3D domain
- Power plant rank correlations
 - High for states near sources (e.g., PA)
 - Lower for far states (e.g., CA)

HyADS reproduces features in more complex models important for health analyses

Ivey et al. 2015 ES&T Dedoussi et al. 2019 ERL Henneman et al. 2019 Atmospheric Environment





Emissions changes and national reductions in HyADS exposure

- 65% reduction in coal power plant SO₂ emissions, 2005-2012
- 69% reduction in HyADS exposure, 2005-2012
- 32% reduction in average PM_{2.5} concentration, 2005-2012
- Questions
 - Did adverse health outcomes decrease with decreasing coal emissions?
 - Are associated decreases different in HyADS and total PM_{2.5}?









Changes in Medicare hospitalization rates associated with coal exposure reductions



- Reduced health outcomes associated with reduced coal emissions and PM_{2.5} exposure
- Regression with Hybrid CMAQ-DDM to convert HyADS to coal PM_{2.5}
- Evidence of larger health reductions for coal exposure reductions than PM_{2.5} reductions

Energy transitions near Louisville, KY

- Identified top four facilities impacting Louisville in 2012 using HyADS
- All units installed SO₂ emissions control or shuttered by 2016
- HyADS exposure decreased over time
- Question did these interventions lead to reduced asthma?

unitless

HyADS,

3e+04

2e+04





5

J. Casey *et al. In review*

Louisville energy transitions natural experiment

- Largest emissions change spring 2015 (Quarter 2)
- Spatial variability across Louisville in who benefited
- ~20% reduction in asthma risk following intervention
- Benefits of transition strongest in areas identified by HyADS

0



Select units' HyADS absolute change



Casey et al. In review

Exposure change and interventions: not all attributable to emissions

- Two reasons for changing exposure:
 - Meteorological variability
 - Emissions change



Exposure = f (Meteorology | Emissions)

 $\Delta \text{Exposure} = f(\text{Met}_{after} | \text{Emiss}_{after}) - f(\text{Met}_{before} | \text{Emiss}_{before})$

 $\Delta Exposure_{met} = f(Met_{before} | Emiss_{before}) - f(Met_{after} | Emiss_{before})$

 $\Delta \text{Exposure}_{\text{emiss}} = f(\text{Met}_{\text{after}} | \text{Emiss}_{\text{before}}) - f(\text{Met}_{\text{after}} | \text{Emiss}_{\text{after}})$

Henneman et al. 2019 Env. Res. Letters

Changes in Louisville exposure

- HyADS change relative to 2012, first quarter
- HyADS exposure changes before 2015 primarily attributable to meteorological variability





Attributing changes national exposure to emissions/meteorology

- Meteorology plays role in who benefits from emissions reductions
 - Δ Meteorology led to <u>smaller</u> Δ Exposure, 2005-2011
 - ΔMeteorology led to larger ΔExposure, 2005-2012
- Attributed to greater recirculation winds around the continent in 2012



Henneman et al. 2019 Env. Res. Letters

In the works – conversion to $\mu g m^{-3}$

- Primarily based on Hybrid CMAQ-DDM
- Accounts for monthly trend, precipitation, temperature
- Annual NMB: 11%
- Annual NME: 22%
- Annual R²: 0.88

The goal: alternative interpretation of HyADS

(Not to reproduce CMAQ-DDM)



HyADS





Conclusions

- HyADS reduced complexity, but...
 - Nimble way to create source-receptor matrix
 - Captures spatial-temporal variability important for environmental health research
- National health benefits achieved through coal emissions reductions
- Asthma reductions in Louisville following multiple interventions
- Meteorology has substantial impacts on calculated benefits
- HyADS currently available as R package

www.github.com/lhenneman/hyspdisp

References

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Extra slides

Large reduction in emissions from United States coal power plants



25 largest facilities

- ~ 1,000 coal electricity generating units (power plants) operating in 2005 and 2012
- SO₂ emissions decreased 65% between 2005 and 2012
- Interventions are costly (\$10's of billions year⁻¹)



https://www.flickr.com/photos/wigwam/2630349031

HyADS Evaluations: Regional comparison with CMAQ-DDM hybrid sensitivities

- Application-specific evaluation important for reduced complexity models
- CMAQ-DDM hybrid PM_{2.5} coal sensitivities seen as gold standard
- High correlation in all regions



Ranking facilities by population-weighted impact in Louisville, KY



- Top 20 facilities that impact Louisville are spread through the Midwest
- Three facilities with large impacts remained in 2012