Using satellite-derived air quality datasets for health applications: Disease burdens and environmental justice Susan Anenberg, PhD

NOAA GEO-XO Town Hall

April 29, 2021

Many collaborators, some named throughout talk

Support from: NASA, Health Effects Institute, Environmental Defense Fund

Milken Institute School of Public Health

THE GEORGE WASHINGTON UNIVERSITY



Air pollution continues to be a leading health risk factor in nearly all countries



>90% of people worldwide live with PM_{2.5} concentrations above the World Health Organization guideline



1 High systolic blood pressure	Metabo
2 Tobacco	Environ risks Behavio
3 Dietary risks	
4 Air pollution	
5 High fasting plasma glucose	
5 High body-mass index	
7 High LDL cholesterol	
3 Kidney dysfunction	
9 Child and maternal malnutrition	
10 Alcohol use	
11 Non-optimal temperature	
12 Unsafe water, sanitation, and handwashing	
13 Occupational risks	
14 Other environmental risks	
15 Unsafe sex	
16 Low physical activity	
17 Drug use	ĺ
18 Low bone mineral density	
19 Intimate partner violence	
20 Childhood sexual abuse and bullying	

2019 rank

Health Effects Institute, State of Global Air 2019 Report (2019)

GBD 2019 Study https://vizhub.healthdata.org/gbd-compare/ ²

Metabolic risks Environmental/occupationa risks Behavioral risks

Satellite remote sensing has transformed our ability to understand air pollution disease burdens globally





2004: **Surface air quality monitors**, 800,000 premature deaths associated with urban PM_{2.5} (Cohen et al. 2004) 2010: **Global chemical transport model**, 3.7 million PM_{2.5} and 700,000 ozone deaths globally (Anenberg et al. 2010) 2012: Satellite observations, global chemical transport model, and ground observations combined,
3.2 million PM_{2.5} and 152,000 ozone deaths (Lim et al. 2012)

2016-2020: **methods refined,** ~4 million PM_{2.5} and 200,000 ozone deaths (Forouzanfar et al. 2016, etc.)

Future: geostationary satellites, lowcost sensors, mobile monitoring, ???

2010: Global PM_{2.5} concentrations from satellite AOD (van Donkelaar et al. 2010)



PM_{2.5} mortality in cities worldwide





NO₂ pollution is an important risk factor for pediatric asthma incidence







Ploy Achakulwisut Arash Mohegh

- % of new pediatric asthma cases attributable to NO₂ >20% in cities in both developed and developing countries.
- Despite substantial declines, NO₂attributable pediatric asthma incidence increased from 2000 to 2019 in many areas of the U.S.



Achakulwisut et al., 2019, *Lancet Planetary Health (2019)* Anenberg, Mohegh, et al. in prep, https://www.essoar.org/pdfjs/10.1002/essoar.10506660.1 Identifying inequities in air pollution health risks challenged by coarsely resolved data inputs for exposure estimation



Maria Castillo



Within-city heterogeneity in air pollution health risks: Importance of picking up traffic



- 38-fold variation in mortality attributable to NO₂ across Bay Area, CA
- Mobile monitoring led to more spatial variation in air pollution health risks
- Hourly measurements from geostationary satellites should improve estimation of traffic-related air pollution health risks



Southerland et al. EHP, 2021

COVID-19 pandemic reveals persistent disparities in NO₂ pollution



In many cities, the post-lockdown NO_2 amounts in the least white communities are still ~50% larger than the prelockdown NO_2 amounts in the most white communities



Kerr, Goldberg, and Anenberg, submitted https://www.essoar.org/doi/pdf/10.1002/essoar.10504561.3



Concluding thoughts

- Satellite remote sensing has transformed environmental health surveillance capabilities
- Limitations of satellite data for health applications
 - Temporal coverage/flyover time
 - Spatial resolution
 - Ability to discern components/mixtures
 - There is still disagreement between surface concentration estimates from different methods
- Some thoughts for future directions
 - Important to have continuous record of satellite datasets
 - Use satellite data to identify areas for locating ground monitors
 - Integrate multiple concentration datasets to leverage strengths of each



