Air Pollution and Health: The India case

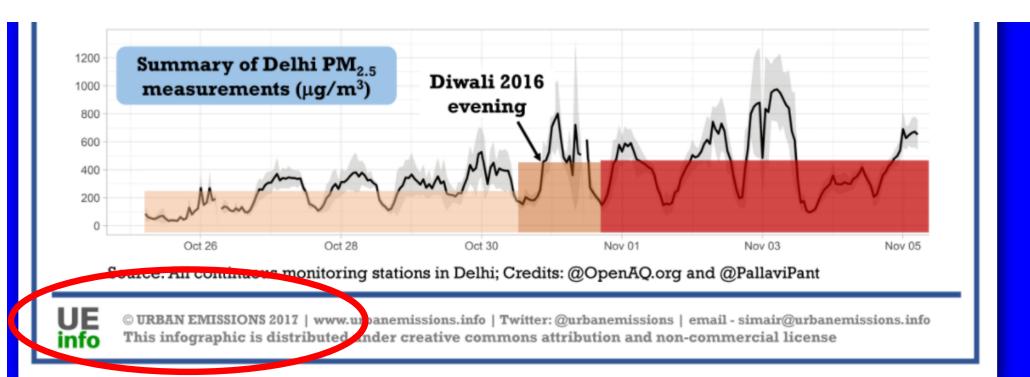
Kirk R. Smith, PhD and John Balmes, MD School of Public Health, UC Berkeley

Collaborative Clean Air Policy Centre

India Habitat Centre, New Delhi

First World NCD Congress

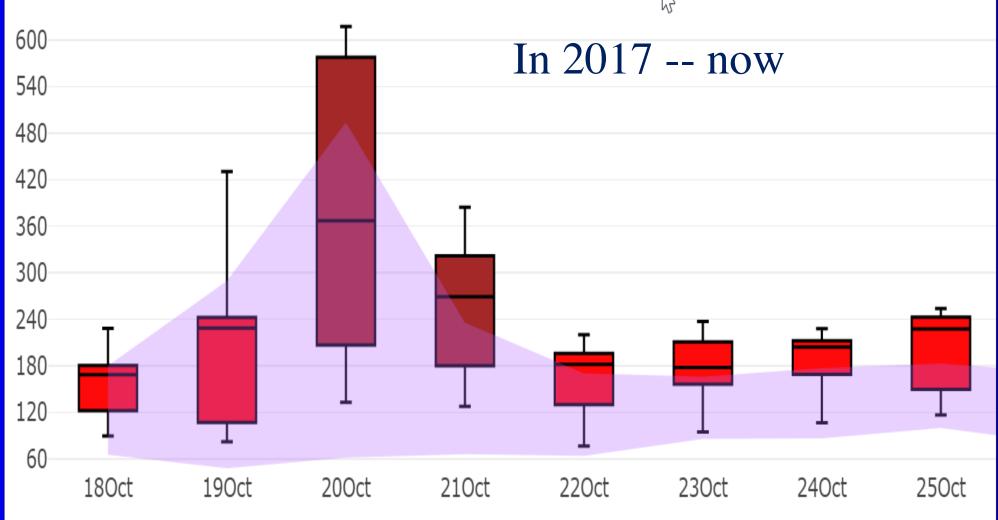
PGIMER, Chandigarh November 5, 2017



Delhi's pollution in 2016 – note effects of 1) the holiday (Diwali) and 2) crop residue burning

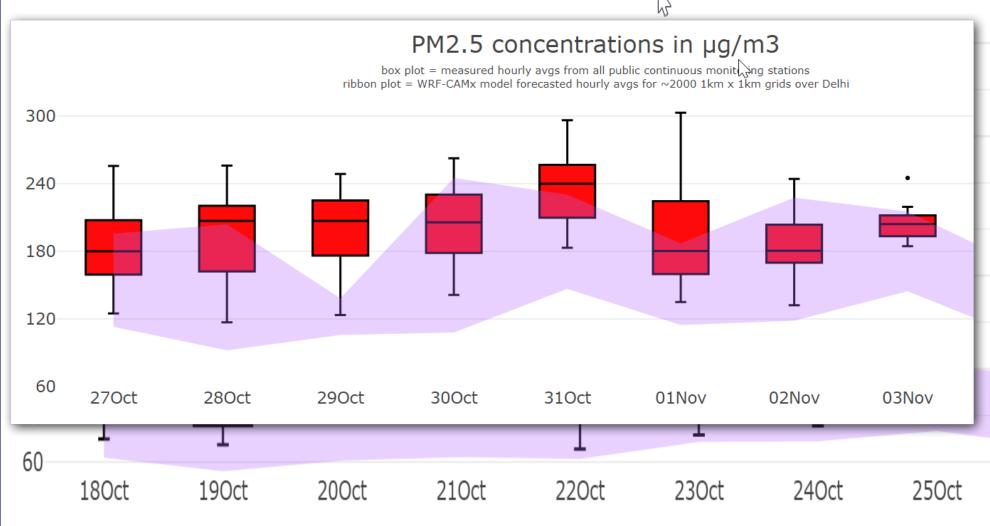
PM2.5 concentrations in µg/m3

box plot = measured hourly avgs from all public continuous monitoring stations ribbon plot = WRF-CAMx model forecasted hourly avgs for \sim 2000 1km x 1km grids over Delhi



PM2.5 concentrations in µg/m3

box plot = measured hourly avgs from all public continuous monitoring stations ribbon plot = WRF-CAMx model forecasted hourly avgs for $\sim\!2000$ 1km x 1km grids over Delhi



Particle deposition

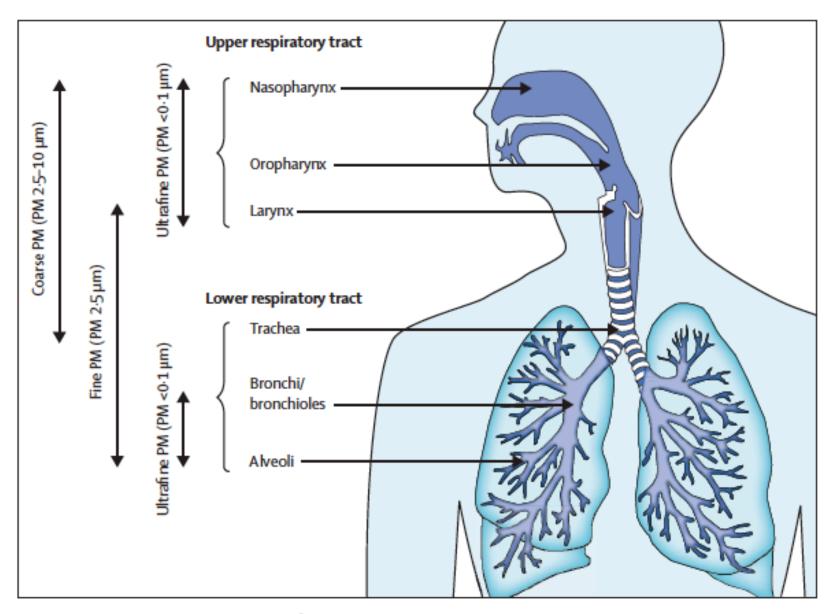
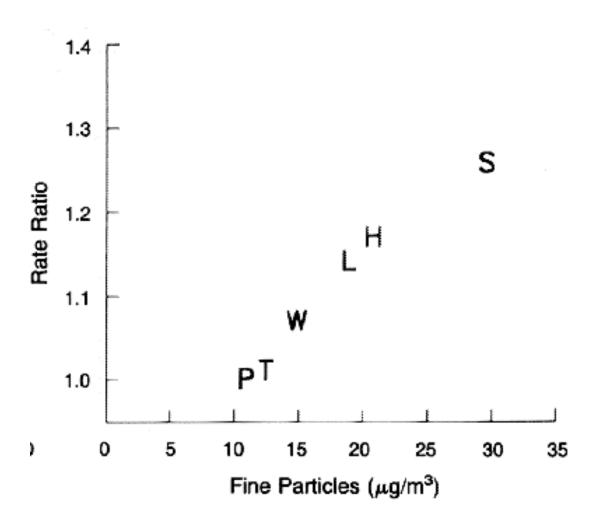


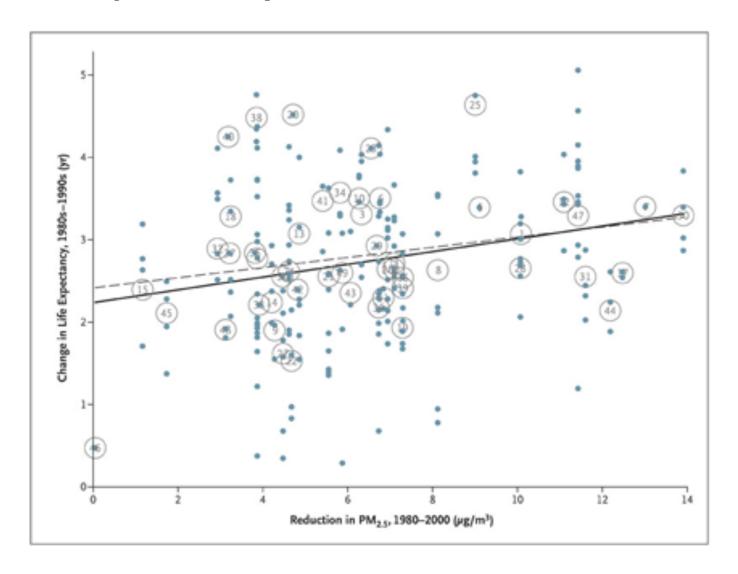
Figure 4: Compartmental deposition of particulate matter

An Association between Air Pollution and Mortality in Six U.S. Cities



Dockery et al. N Engl J Med 1993;329:1753-1759

Fine-Particulate Air Pollution and Life Expectancy in the United States



Pope et al. N Engl J Med 2009;360:376-386.

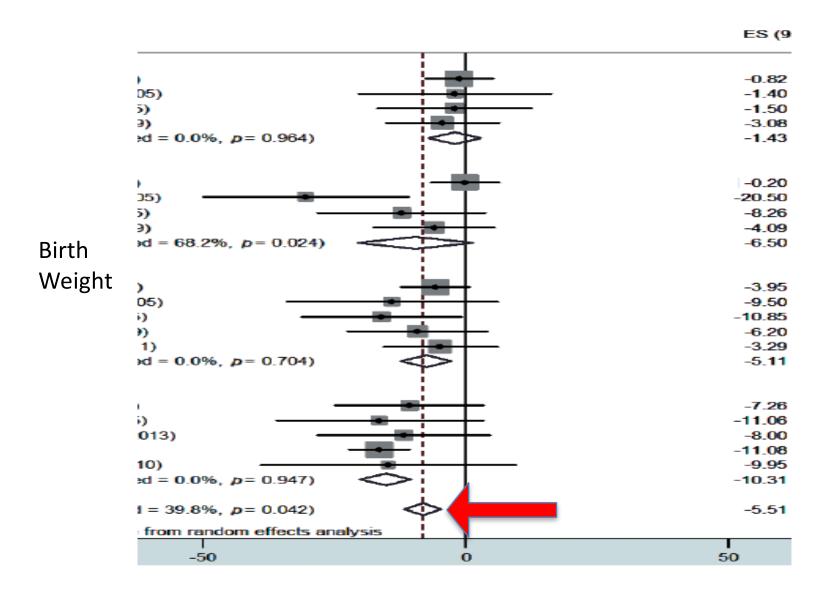
Other Effects - examples

- Low birth weight
- Pre-term birth
- BMI
- Diabetes

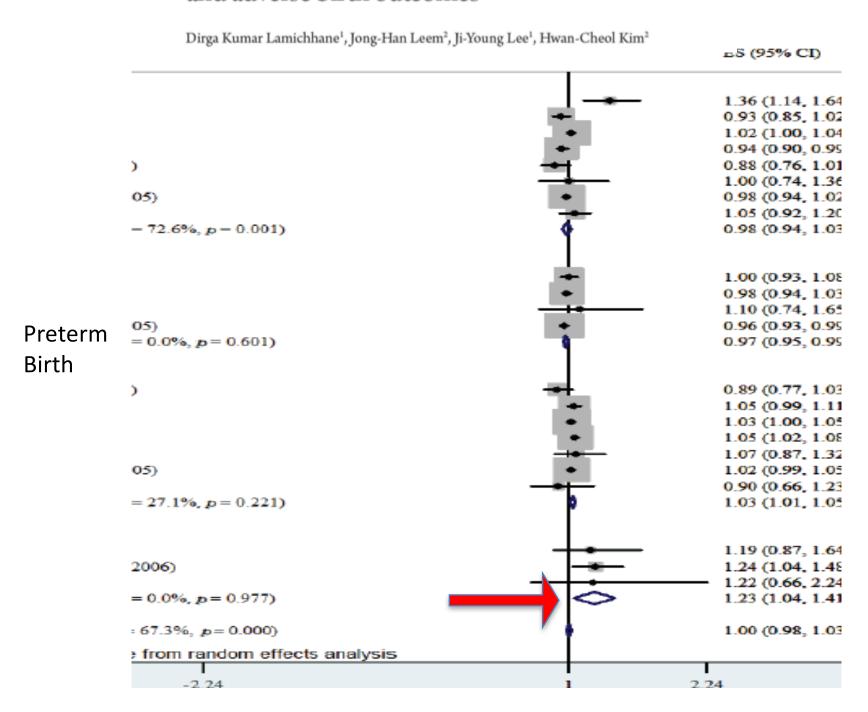
A meta-analysis of exposure to particulate matter and adverse birth outcomes

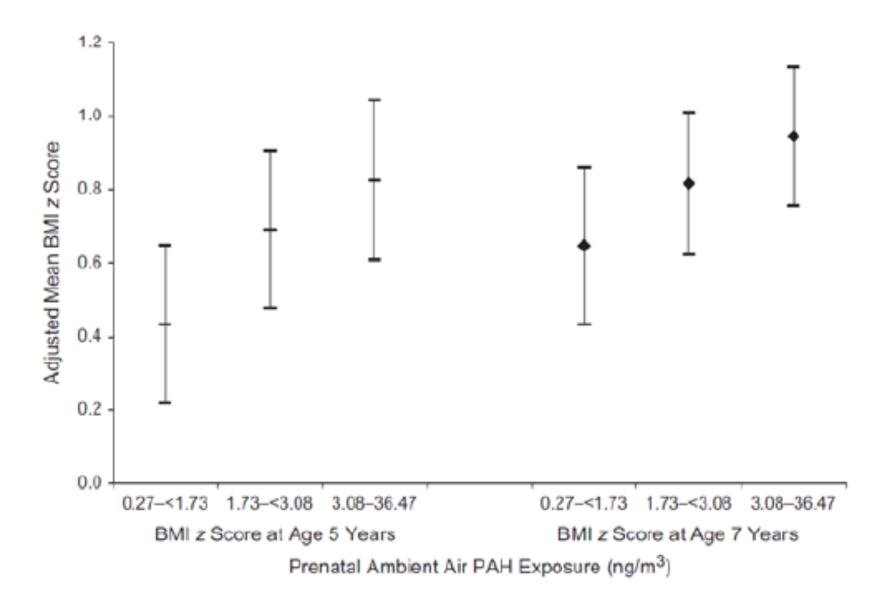
Dirga Kumar Lamichhane¹, Jong-Han Leem², Ji-Young Lee¹, Hwan-Cheol Kim²

Environ Health Toxicol 2015;30:e2015011.



A meta-analysis of exposure to particulate matter and adverse birth outcomes





Rundle et al., Am J Epidemiol 2012

Air Pollution and Diabetes

- Several studies have shown associations between diabetes in adults and exposure to trafficrelated air pollution (TRAP)
- Evidence also building for children



Potential Mechanism

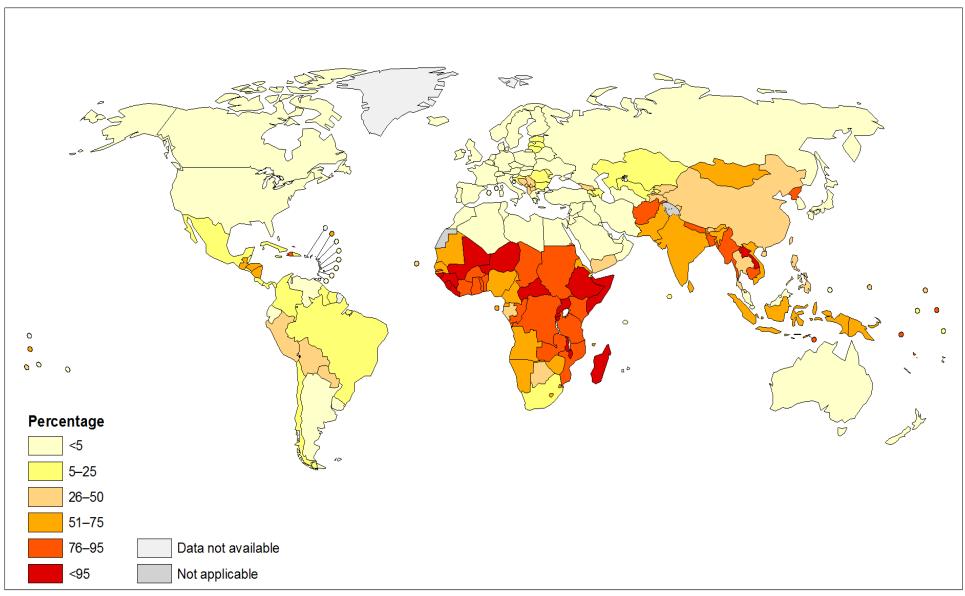
- Air pollution can induce oxidative stress and systemic inflammation
- PM_{2.5} induced adipose tissue inflammation and insulin resistance in a mouse model of diet-induced obesity (Sun et al. Circulation 2009)
- Hypothesis:
 - Exposure to air pollution in utero and in early childhood increases risk of obesity and abnormal glucose metabolism later in childhood







Population Cooking with Solid Fuels in 2010 (%)



The boundaries and names shown and the designations used on this map do not imply the expression of any opinion whatsoever on the part of the World Health Organization concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. Dotted and dashed lines on maps represent approximate border lines for which there may not yet be full agreement.

Data Source: World Health Organization
Map Production: Public Health Information
and Geographic Information Systems (GIS)
World Health Organization



Toxic Pollutants in Wood Smoke from Simple (poor) Combustion

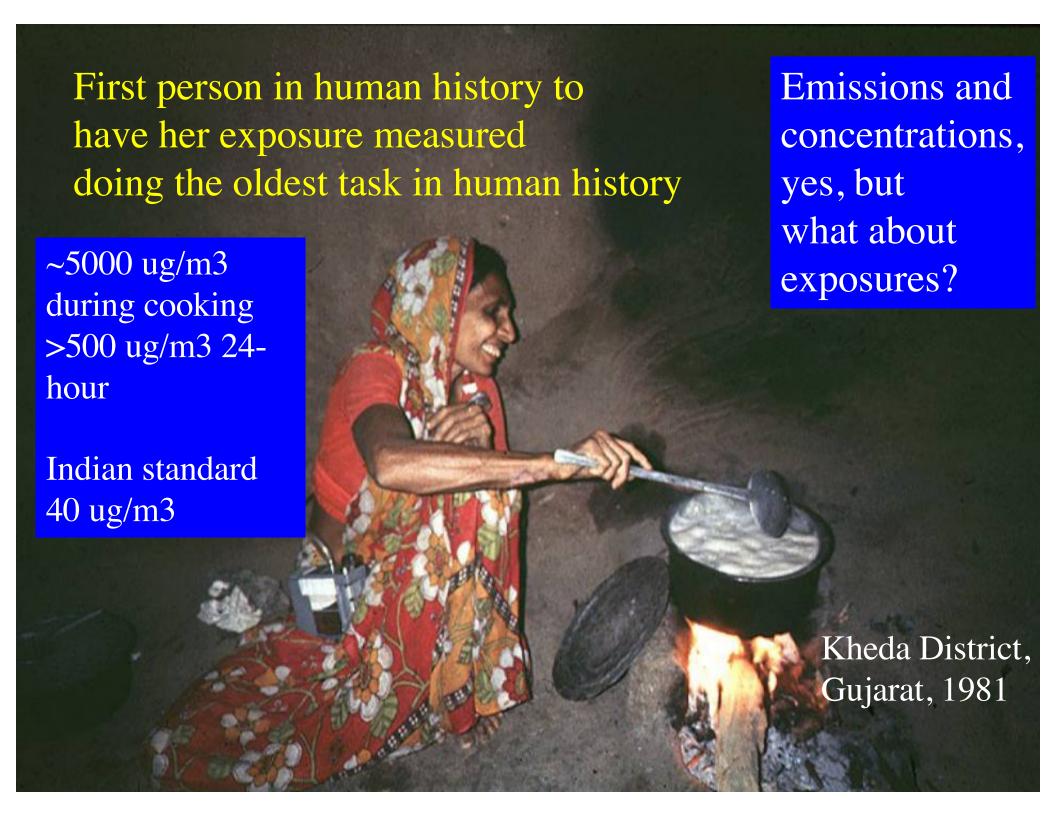
• Small particles, CO, NO₂

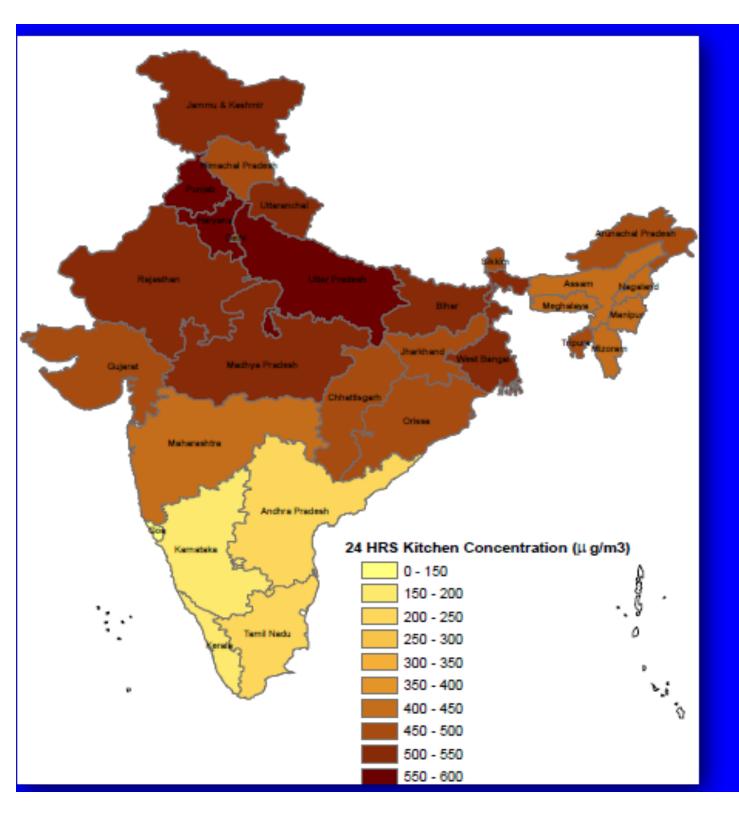


- 33+ phenols such as catechol & cresol
- Many quinones such as hydroquinone
- Semi-quinone-type and other radicals

Source: Naeher et al, *J Inhal Tox*, 2007

• Chlorinated organics such as methylene chloride and dioxin

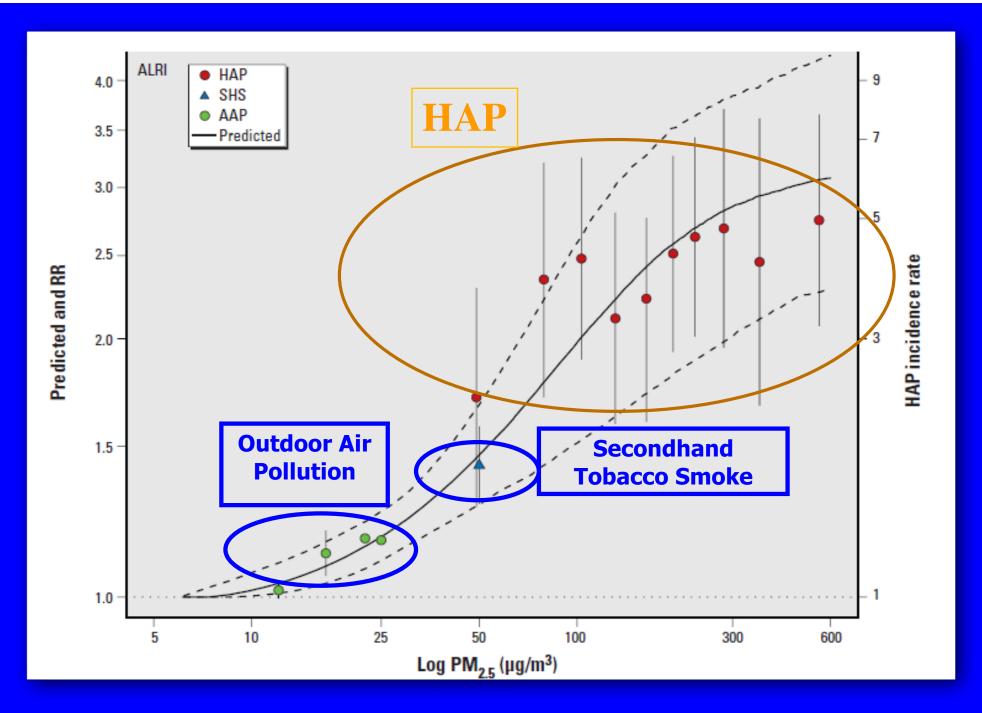




State-wise estimates of 24-h kitchen concentrations of PM2.5 in India

Solid-fuel using households

Balakrishnan et al. 2013 (SRU group)



Burnett et al., EHP. 2014, Integrated Exposure-Response Functions

Table 2. Adjusted relative risk estimates^a for various increments of exposure from cigarette smoking (versus never smokers), second hand cigarette smoke, and ambient air pollution from the present analysis and selected comparison studies.

	Increments of		Estimated Daily			
Source of risk estimate	Exposure	Lung Cancer	IHD	CVD	CPD	Dose PM _{2.5} (mg) ^b
ACS- present analysis	≤3 (1.5) cigs/day	10.44 (7.30-14.94)	1.61 (1.27-2.03)	1.58 (1.32-1.89)	1.72 (1.46-2.03)	18
ACS- present analysis	4-7 (5.5) cigs/day	8.03 (5.89-10.96)	1.64 (1.37-1.96)	1.73 (1.51-1.97)	1.84 (1.63-2.08)	66
ACS- present analysis	8-12 (10) cigs/day	11.63 (9.51-14.24)	2.07 (1.84-2.31)	2.01 (1.84-2.19)	2.10 (1.94-2.28)	120
ACS- present analysis	13-17 (15) cigs/day	13.93 (11.04-17.58)	2.18 (1.89-2.52)	1.99 (1.77-2.23)	2.08 (1.87-2.32)	180
ACS- present analysis	18-22 (20) cigs/day	19.88 (17.14-23.06)	2.36 (2.19-2.55)	2.42 (2.28-2.56)	2.52 (2.39-2.66)	240
ACS- present analysis	23-27 (25) cigs/day	23.82 (18.80-30.18)	2.29 (1.91-2.75)	2.33 (2.02-2.69)	2.33 (2.03-2.67)	300
ACS- present analysis	28-32 (30) cigs/day	26.82 (22.54-31.91)	2.22 (1.97-2.49)	2.17 (1.98-2.38)	2.39 (2.19-2.60)	360
ACS- present analysis	33-37 (35) cigs/day	26.72 (18.58-38.44)	2.58 (1.91-3.47)	2.52 (1.98-3.19)	2.83 (2.28-3.52)	420
ACS- present analysis	38-42 (40) cigs/day	30.63 (25.79-36.38)	2.30 (2.05-2.59)	2.37 (2.16-2.59)	2.61 (2.40-2.84)	480
ACS- present analysis	43+ (45) cigs/day	39.16 (31.13-49.26)	2.00 (1.62-2.48)	2.17 (1.84-2.56)	2.37 (2.04-2.76)	540
ACS-air pol. original	24.5 μg/m³ ambient PM _{2.5}				1.31(1.17-1.46)	0.44
ACS-air pol. extend.	10 μg/m³ ambient PM _{2.5}	1.14(1.04-1.23)	1.18(1.14-1.23)	1.12(1.08-1.15)	1.09(1.03-1.16)	0.18
HSC-air pol. original	18.6 μg/m³ ambient PM _{2.5}				1.37(1.11-1.68)	0.33
HSC-air pol. extend.	10 μg/m³ ambient PM _{2.5}	1.21(0.92-1.69)		1.28(1.13-1.44)		0.18
WHI-air pol.	10 μg/m³ ambient PM _{2.5}			1.24(1.09-1.41) ^c		0.18
SGR-SHS	Low- moderate SHS exp.			1.16(1.03-1.32)		0.36
SGR-SHS	Moderate-high SHS exp			1.26(1.12-1.42)		0.90
SGR-SHS	Live with smoking spouse	1.21(1.13-1.30)				0.54
SGR-SHS	Work with SHS exposure	1.22(1.13-1.33)				0.72
INTERHEART	1-7 hrs/wk SHS exp.		1.24(1.17-1.32) ^d			0.36
INTERHEART	Live with smoking spouse		1.28(1.12-1.47)d			0.54

Pope et al.

Environmental Health
Perspectives
2011, in press

Lung Cancer 35 Adjusted Relative Risk 30 25 20 1.50 10 1.25 5 1.00 Ischemic heart (light gray) Cardiovascular (dark gray) Cardiopulmonary (black) 3.0 Adjusted Relative Risk 2.5 2.0 1.25 1.00 0.0 1.0 120 240 300 360 420 (<3)(4-7)(8-12)(13-17) (18-22) (23-27)(28-32)(33-37)(38-42)Estimated daily exposure, mg of PM_{2.5} and increments of cigs/day

Lung Cancer

Heart Disease

Pope et al.

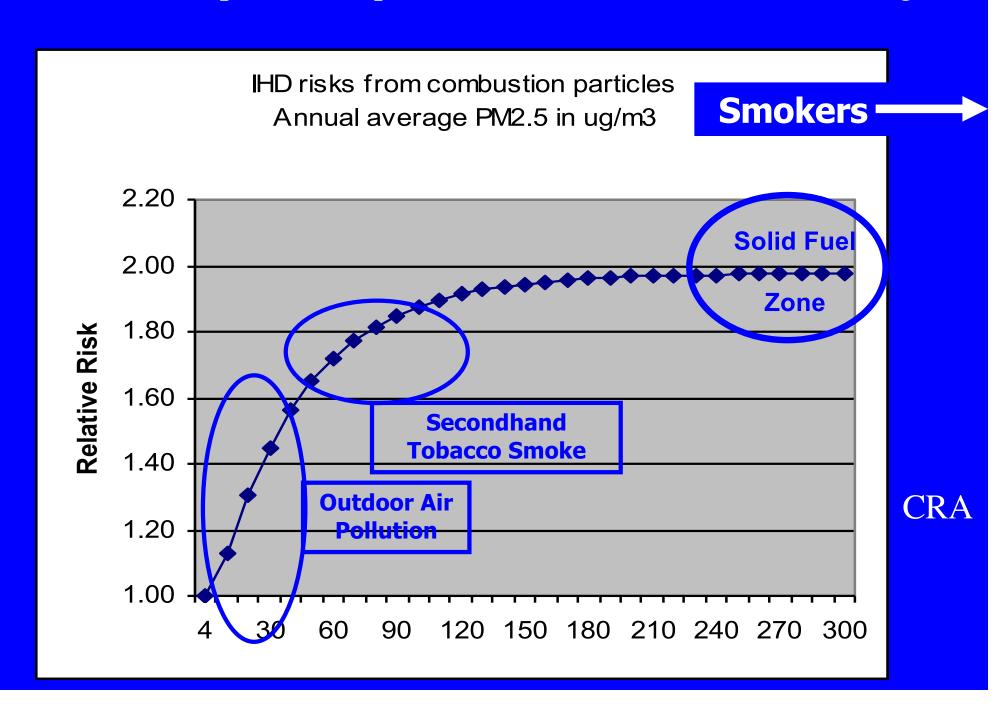
Environmental

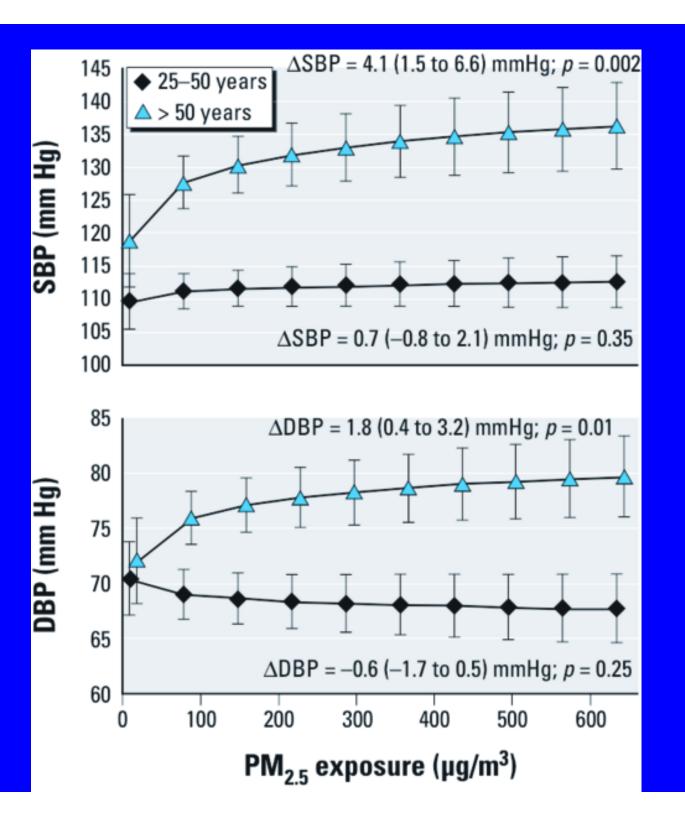
Health

Perspectives

2011, in press

Generalized Exposure-Response: Outdoor Air, SHS, and Smoking





Household
Air
Pollution
and
Blood Pressure

In Yunnan

Baumgartner et al.

<u>Environmental Health</u>

<u>Perspectives</u> 2011, Oct

Intervention to Lower Household Wood Smoke Exposure in Guatemala Reduces ST-Segment Depression on Electrocardiograms

John McCracken, 1,2 Kirk R. Smith, Peter Stone, Anaité Díaz, Byron Arana, and Joel Schwartz 1

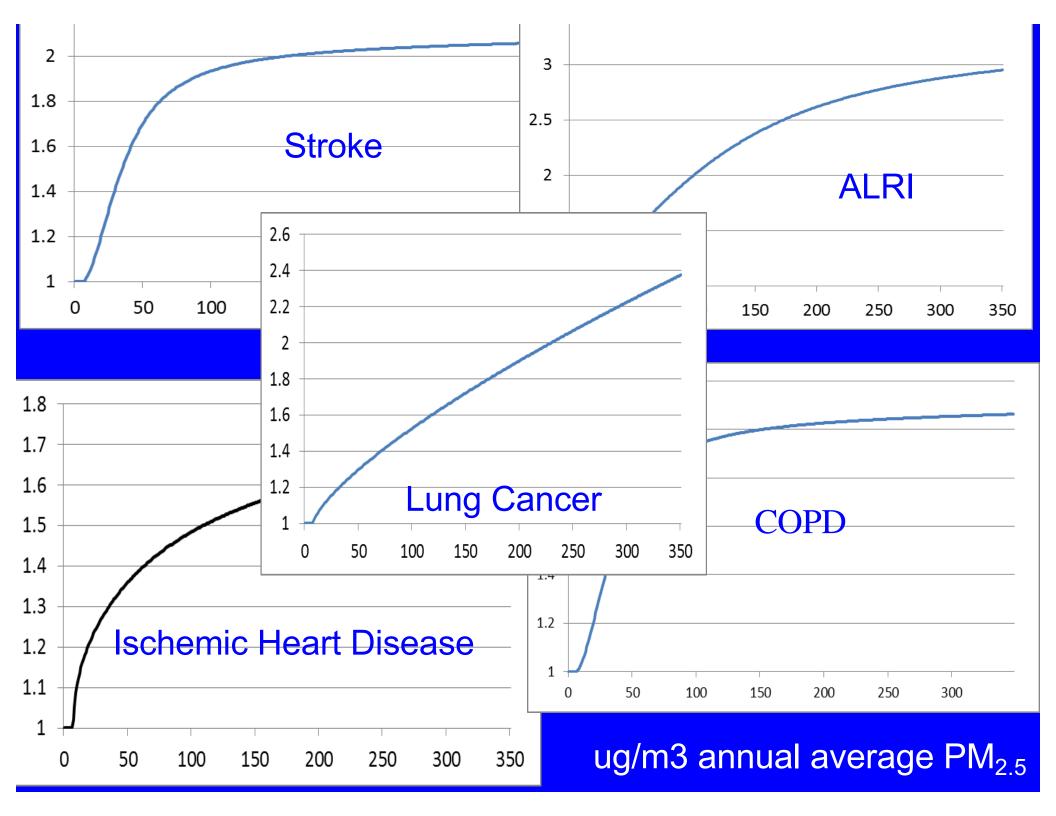
¹Department of Environmental Health, Harvard School of Public Health, Boston, Massachusetts, USA; ²Environmental Sciences Division, University of California, Berkeley, California, USA; ³Brigham and Women's Hospital, Boston, Massachusetts, USA; ⁴Center for Health Studies, Universidad del Valle, Guatemala City, Guatemala

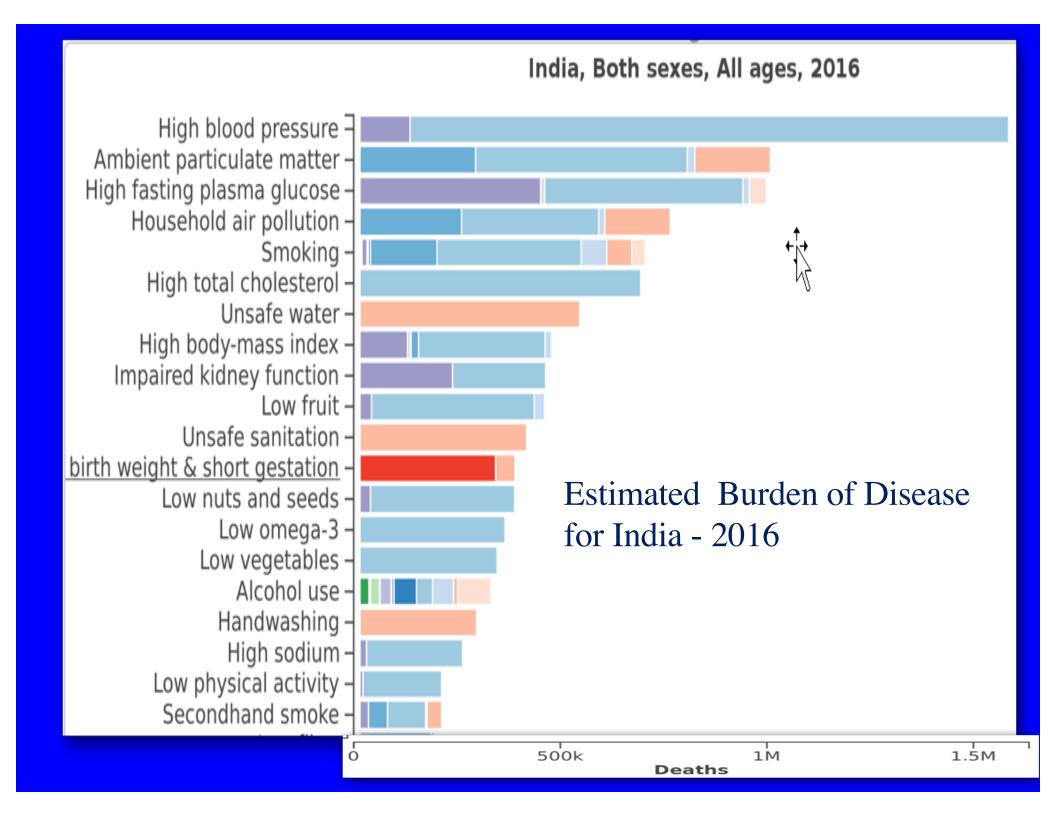
EHP Nov, 2011

Table 3. Odds ratios (ORs) for nonspecific ST-segment depression (30-min average ≤ −1 mm, regardless of slope) associated with chimney-stove intervention compared with open fire from two study designs: between-groups and before-and-after analyses.

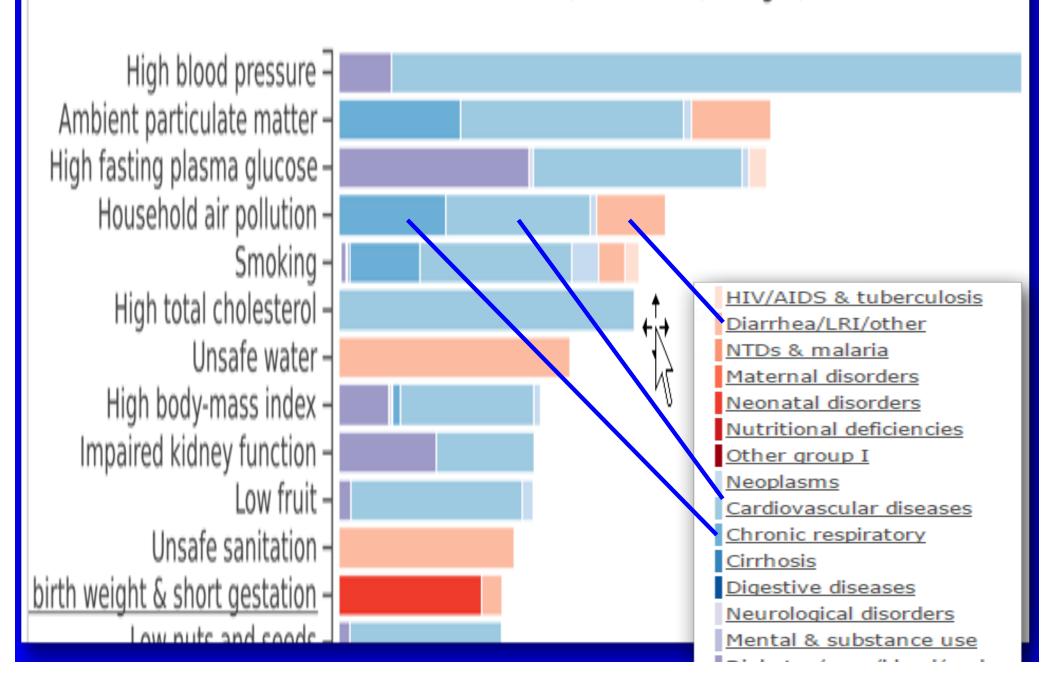
	Crude	Adjusted		
Comparison	OR (95% CI)	p-Value	OR (95% CI)	p-Value
Between-groups	0.34 (0.15, 0.81)	0.015	0.26 (0.08, 0.90)	0.033
Before-and-after (only control group)	0.41 (0.24, 0.70)	0.001	0.28 (0.12, 0.63) ^b	0.002

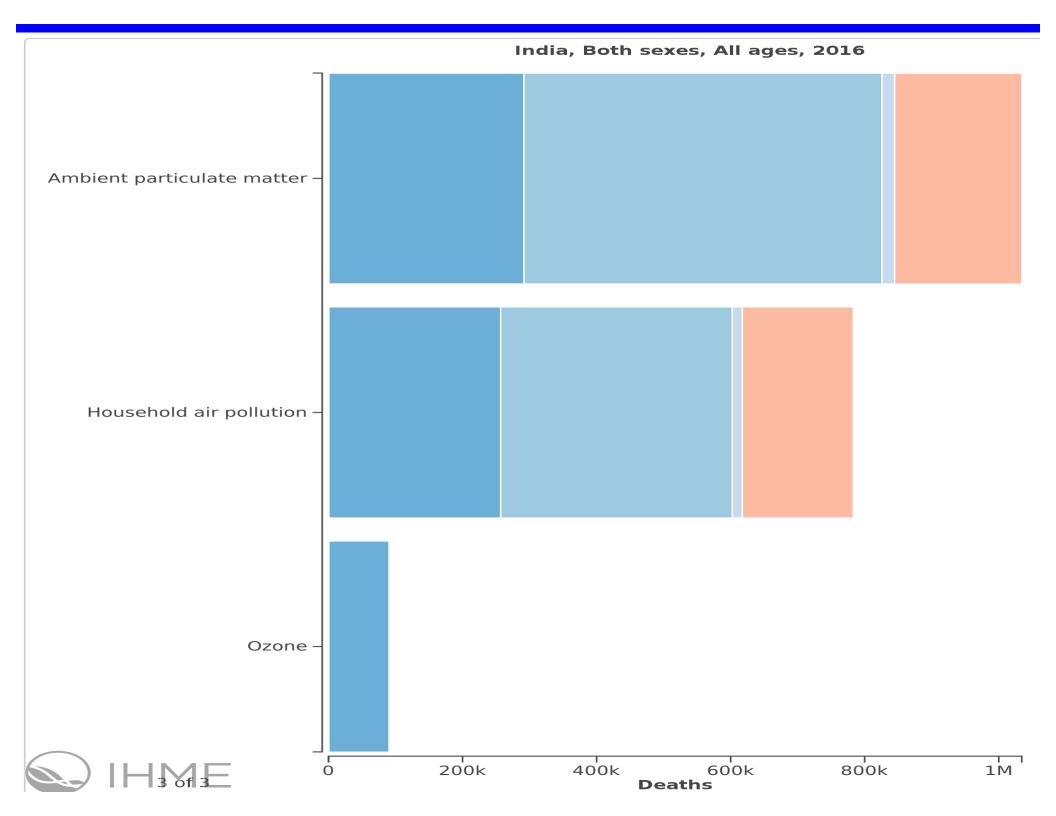
[&]quot;Adjusted for age (quadratic), BMI (quadratic), asset index category, ever smoking, SHS, owning a wood-fired sauna, recent use of wood-fired sauna, and time of day (natural spline with 5 degrees of freedom). "Adjusted for age (quadratic), day of week, season (wet/dry), daily average temperature and relative humidity, daily rainfall, interactions of weather variables with season, recent use of wood-fired sauna, and time of day (natural spline with 5 degrees of freedom).



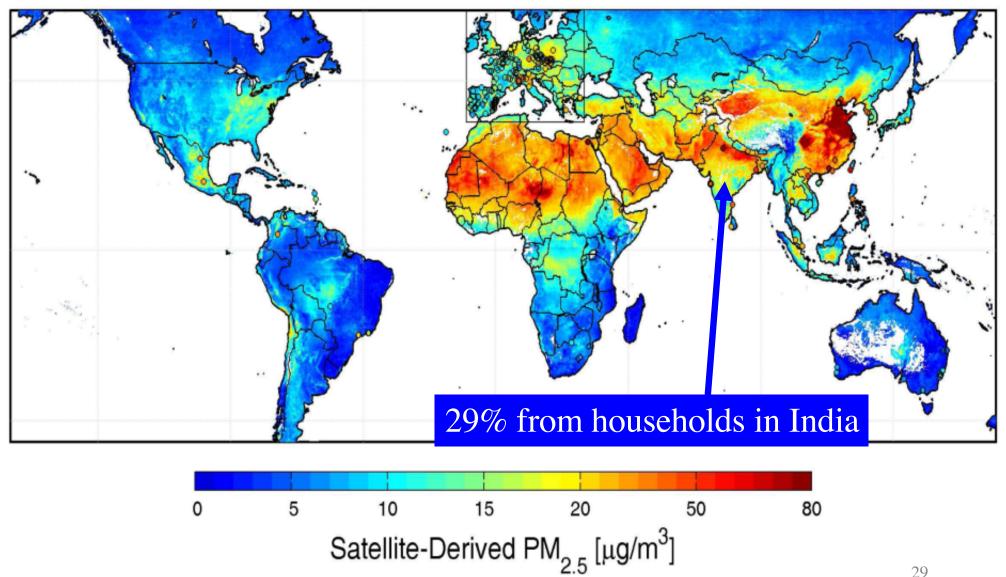


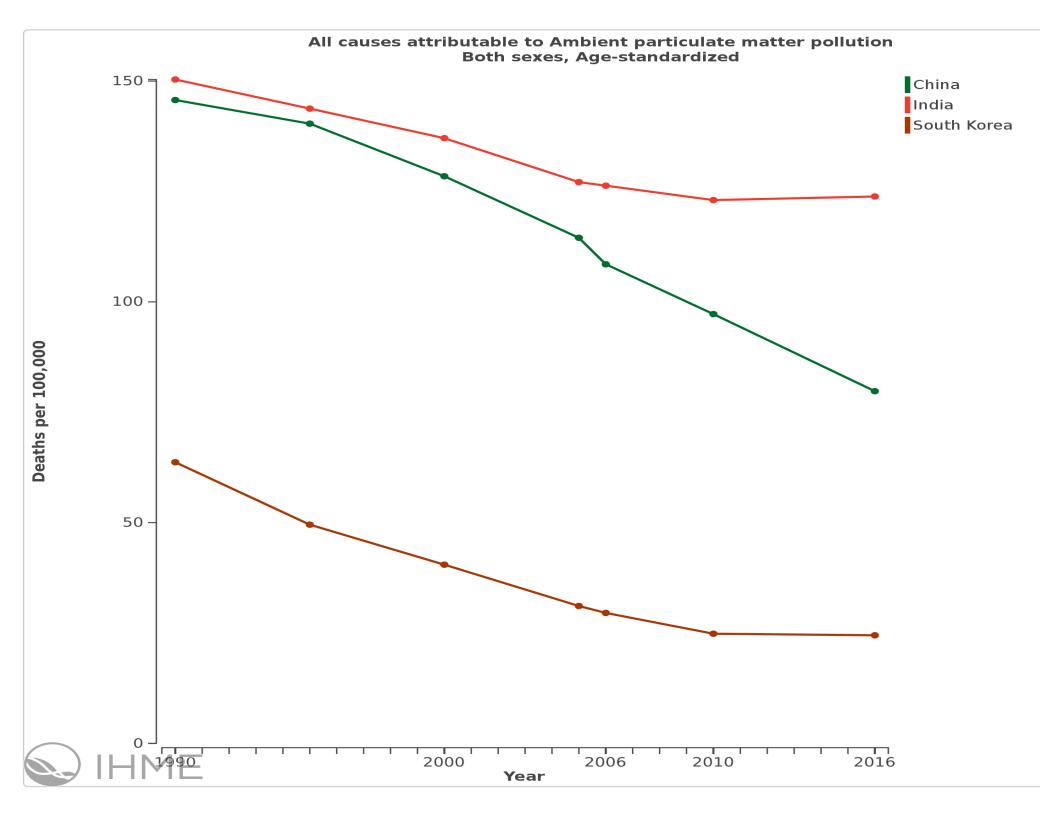
India, Both sexes, All ages, 2016

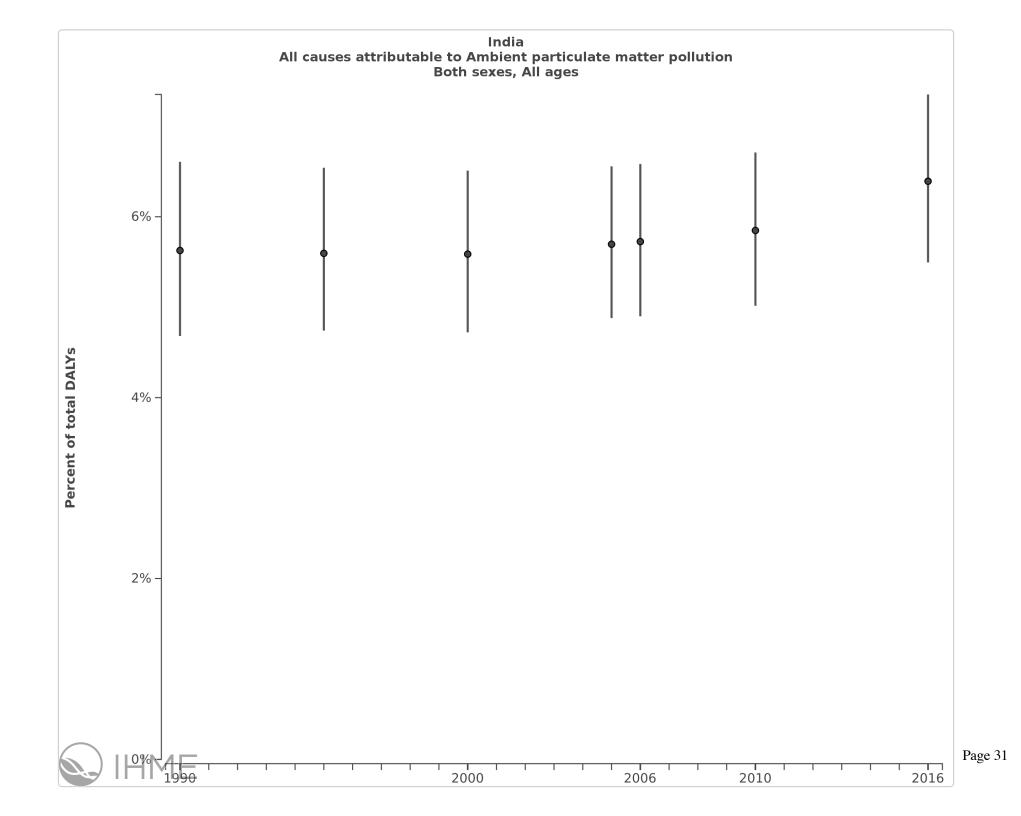




Satellite-based ambient PM_{2.5}







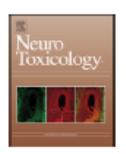
Remember

- GBD numbers are highly uncertain these are central estimates
- And change with new models and databases
- Some health outcomes not included
 - Low birth weight/prematurity
 - TB/asthma
 - Other cancers: cervical, etc.
 - Diabetes, arthritis, low IQ, BMI



Contents lists available at SciVerse ScienceDirect

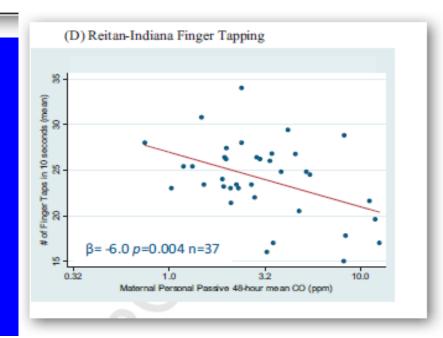
NeuroToxicology

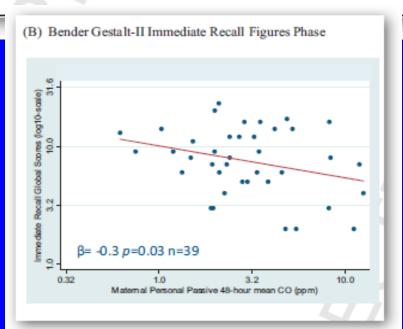


Neurodevelopmental performance among school age children in rural Guatemala is associated with prenatal and postnatal exposure to carbon monoxide, a marker for exposure to woodsmoke

Linda Dix-Cooper, a, Brenda Eskenazi, b, Carolina Romero, c, John Balmes, a,d, Kirk R. Smith, a,*

^d Division of Occupational and Environmental Medicine, Department of Medicine, University of California, San Francisco, CA, USA

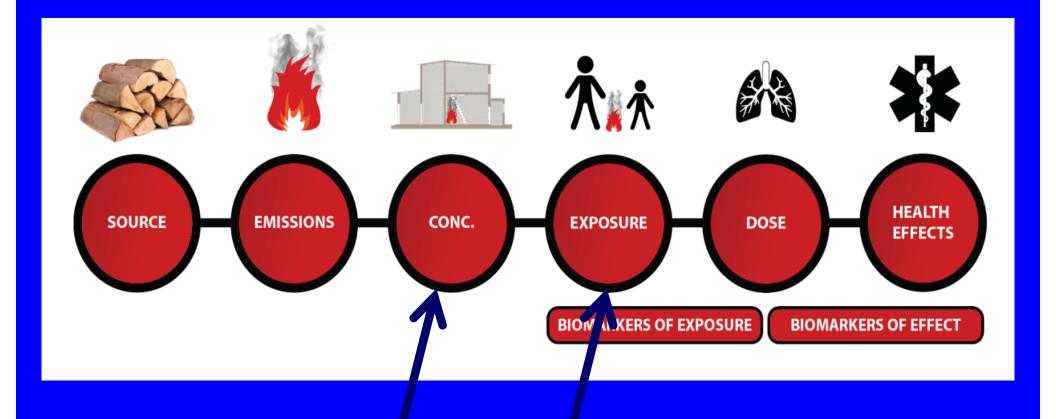




Division of Environmental Health Sciences, School of Public Health, University of California, Berkeley, CA 94720-7360, USA

b Center for Environmental Research and Children's Health (CERCH), School of Public Health, University of California, Berkeley, CA, USA

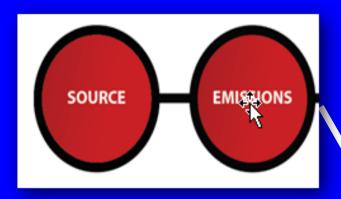
Centro de Estudios en Salud Universidad Del Valle, Guatemala



Classic air politytionrecintarchftodubestonm etric concentrationerin fidereltarchitodubestonm etric concentrationerin fidereltarchitodubestonm etric

Source – Exposure Relationships

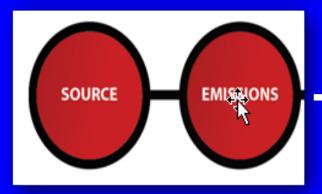
Vehicles

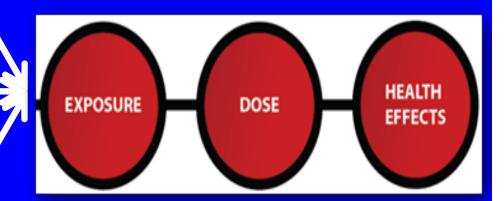


How different?

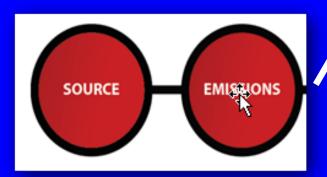
Does it matter?

Power plants





Stoves

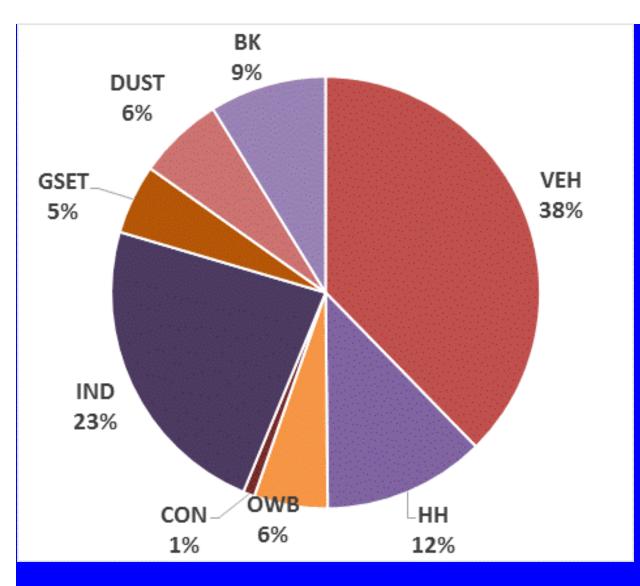


Ministry of Health and Family Welfare Air Pollution Task Force - 2015

- First Ministry of Health in world to treat AP as one of its major priorities and consider along with other risk factors in its mission
- First government agency in the world not to address AP by location, but by total exposure a true health focus
- Thus, not indoor/household, not outdoor, but by what will give the most health benefit

MoHFW AP Task Force

- Total exposure approach requires utilizing estimates of relation between emissions of each source category and exposure.
- Emissions weighted essentially by proximity to population
- Goal is to change source apportionment to exposure apportionment
- Several analytic approaches now available new research agenda to make viable for policy



Hyderabad-2012

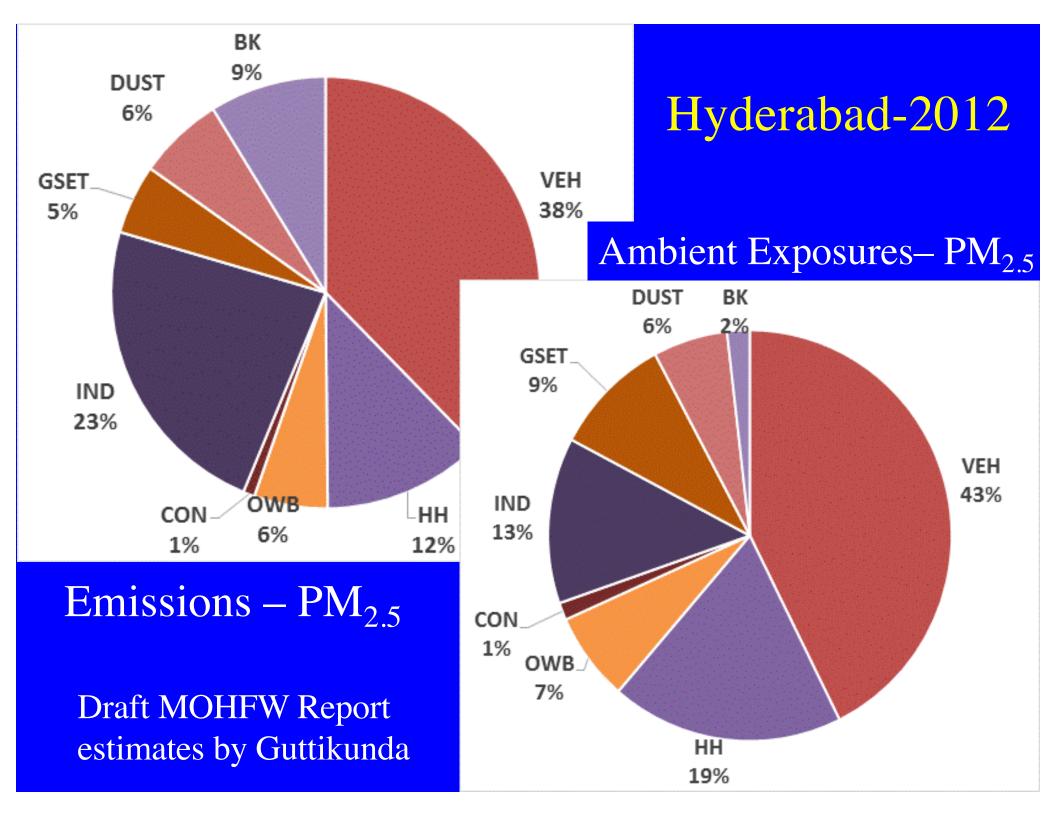
Emissions – $PM_{2.5}$

MOHFW Report, 2015 estimates by Guttikunda

Ambient Intake Fractions in Hyderabad

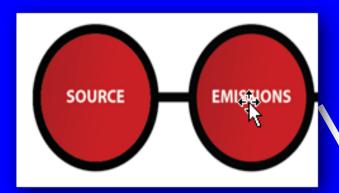
ppm – grams inhaled per tonne emitted

	Average	SD
Households	175	97
Construction	175	93
Waste.burn	140	74
Veh.exhaust	130	64
Gen.sets	123	53
Industries	65	17
Dust	18	4
Power plants	7.4	7.0
Brick.kilns	6.8	1.9



Source – Exposure Relationships

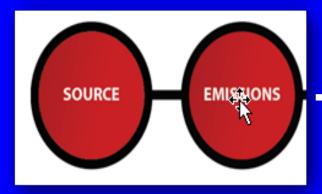
Vehicles

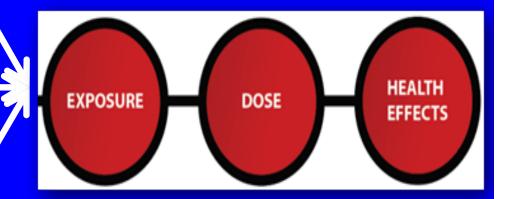


How different?

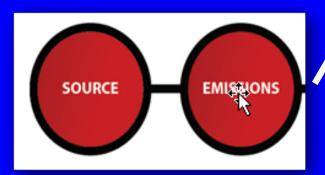
Does it matter?

Power plants

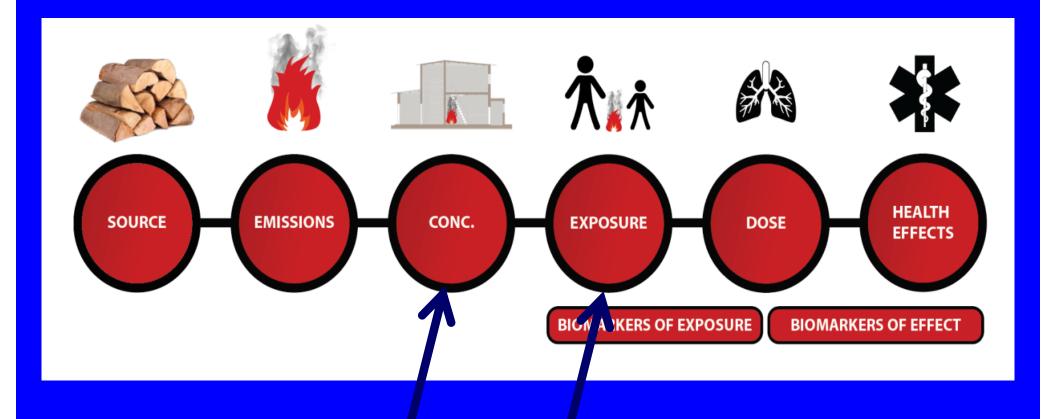




Stoves



Yes, a factor of 100 different!



Classic air mally in the control of that weights concentration of the control of

Summary

- Eventually, we wish to control all sources of air pollution, all the time, everywhere.
- But we cannot afford to do so immediately
- What metric gives the optimal pathway such that the most health protection is occurring at each stage of investment?
- Metrics of exposure are the way to do so

India

- Still with ~two-thirds of households using solid cookfuels
- Most polluted cities in the world, but also major amounts of ambient pollution in rural areas
- Highest burden of disease from air pollution in the world
- Highest total air pollution burden/capita of all middle-income countries 2x China















Mission

The CCAPC

- evaluates, and compares policy options for dealing with India's health-damaging air pollution of all types,
- provides a platform for institutions to work together to solve problems and recommend policy, and
- works to develop capacity to address the policy implications of air pollution in the country.

NIMBY versus MIMBY

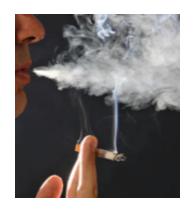
- Not in my backyard (NIMBY) is a well known issue in environmental health
- But MIMBY (Must be in my backyard) is a more fundamental problem
- Local data are obviously valuable
- But we cannot repeat every study in every part of the world
- When do we have enough information for policy?
- In particular for HAP risk for CVD outcomes; these now come from ambient and other data only?

Risk Modifiers

- Age
- Sex
- Race/ethnicity
- Socioeconomic status
- Neighborhood
- Stress
- Diet
- Obesity
- Diabetes
- Other exposures (e.g., tobacco smoke, biomass smoke, occupational vapors, dusts, fumes)







Many thanks

For publications & presentations: Just "Google" Kirk R. Smith

For the new Centre http://www.ccapc.org.in/

