# Identifying the Most Harmful Sources of Ambient Air Pollution to Better Protect Public Health

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# Air pollution and health

Particulate matter (PM)

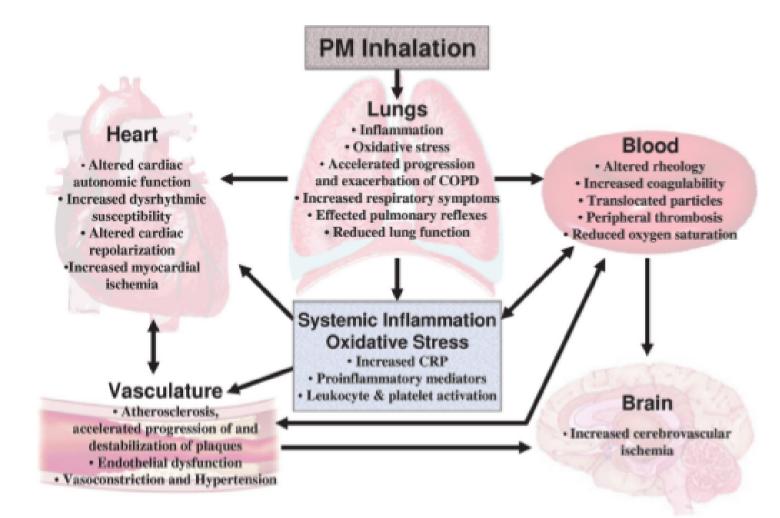


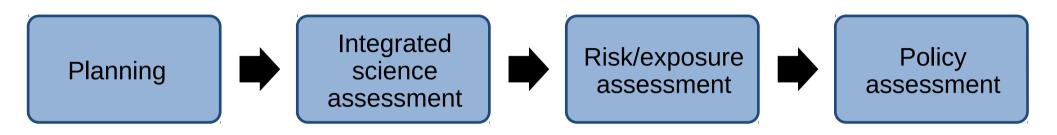
Image: Pope and Dockery (2006)

# National Ambient Air Quality Standards (NAAQS)

Pollutant [links to historical tables of NAAQS reviews]		Primary/ Secondary	Averaging Time	Level	Form
<u>Carbon Monoxide (CO)</u>		primary	8 hours	9 ppm	Not to be exceeded more than once per year
		printary	1 hour	35 ppm	Not to be exceeded more than once per year
Lead (Pb)		primary and secondary	Rolling 3 month average	0.15 µg/m <sup>3 <u>(1)</u></sup>	Not to be exceeded
<u>Nitrogen Dioxide (NO<sub>2</sub>)</u>		primary	1 hour	100 ppb	98th percentile of 1-hour daily maximum concentrations, averaged over 3 years
		primary and secondary	1 year	53 ppb (2)	Annual Mean
<u>Ozone (O<sub>3</sub>)</u>		primary and secondary	8 hours	0.070 ppm <sup>(3)</sup>	Annual fourth-highest daily maximum 8-hour concentration, averaged over 3 years
	PM <sub>2.5</sub>	primary	1 year	12.0 µg/m <sup>3</sup>	annual mean, averaged over 3 years
		secondary	1 year	15.0 µg/m <sup>3</sup>	annual mean, averaged over 3 years
Particle Pollution (PM)		primary and secondary	24 hours	35 µg/m <sup>3</sup>	98th percentile, averaged over 3 years
	PM <sub>10</sub>	primary and secondary	24 hours	150 µg/m <sup>3</sup>	Not to be exceeded more than once per year on average over 3 years
<u>Sulfur Dioxide (SO<sub>2</sub>)</u>		primary	1 hour	75 ppb (4)	99th percentile of 1-hour daily maximum concentrations, averaged over 3 years
		secondary	3 hours	0.5 ppm	Not to be exceeded more than once per year

# NAAQS review

Separately for each criteria pollutant:



- Individual pollutants vs. mixtures
- PM is a heterogeneous mixture
- Mixtures emitted from sources
  - Interpretable
  - Better targets of intervention

### Sources: where does air pollution come from?

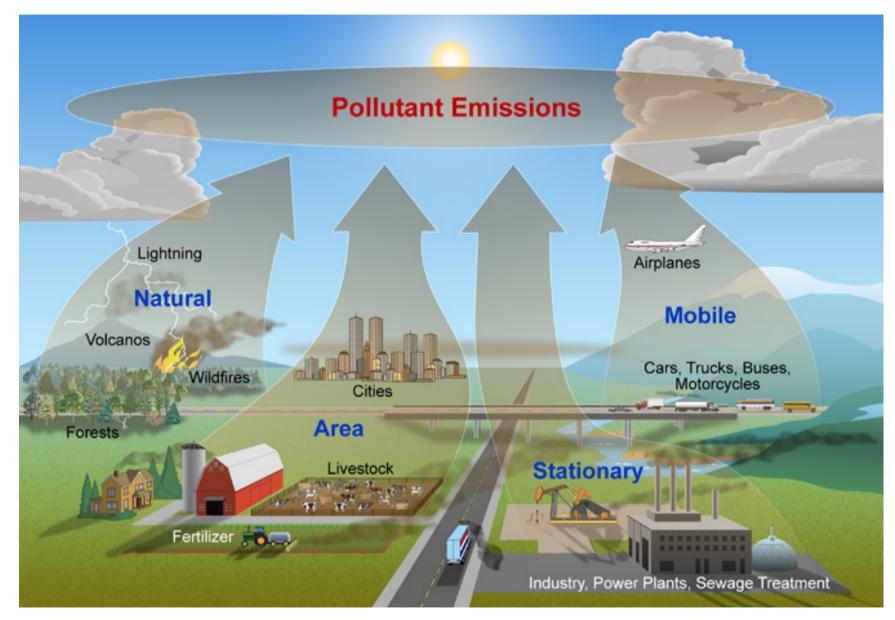


Image: nps.gov

# Air pollution

Exposure-oriented air pollution monitor

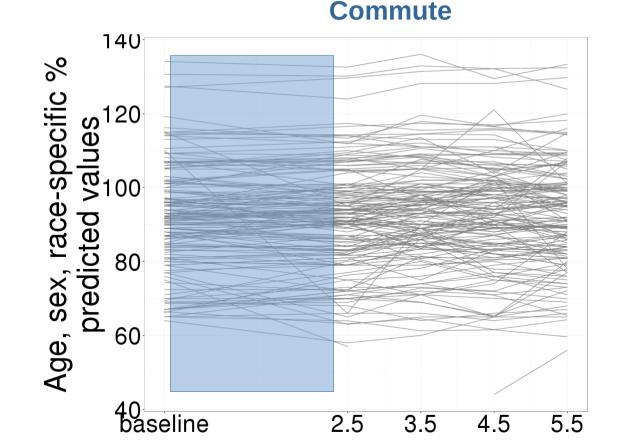
- Ambient monitor
- Relevant exposure monitor (e.g. in-vehicle)



#### Source-oriented air pollution monitor

# Atlanta Commuters Exposure (ACE) Study

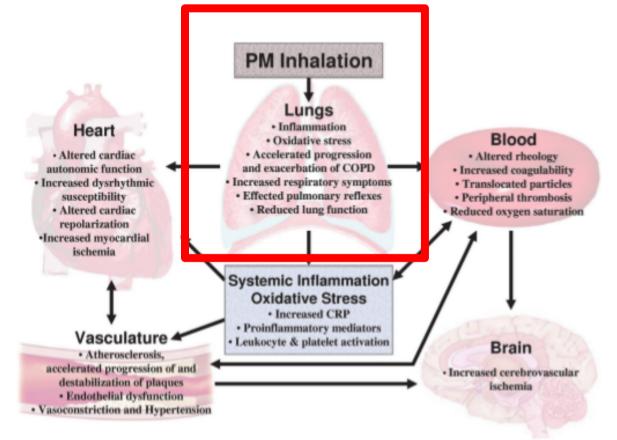
- 100 individuals performed 161 commutes around Atlanta, GA
- Healthy and asthmatic participants
- In-vehicle pollution measurements
- Pre and post-exposure lung function measurements
- Forced expiratory volume (FEV1)



# Atlanta Commuters Exposure (ACE) Study

# Subclinical health outcomes

- Forced expiratory volume in the first second (FEV1)
- Forced vital capacity (FVC)
- Exhaled nitric oxide (eNO)



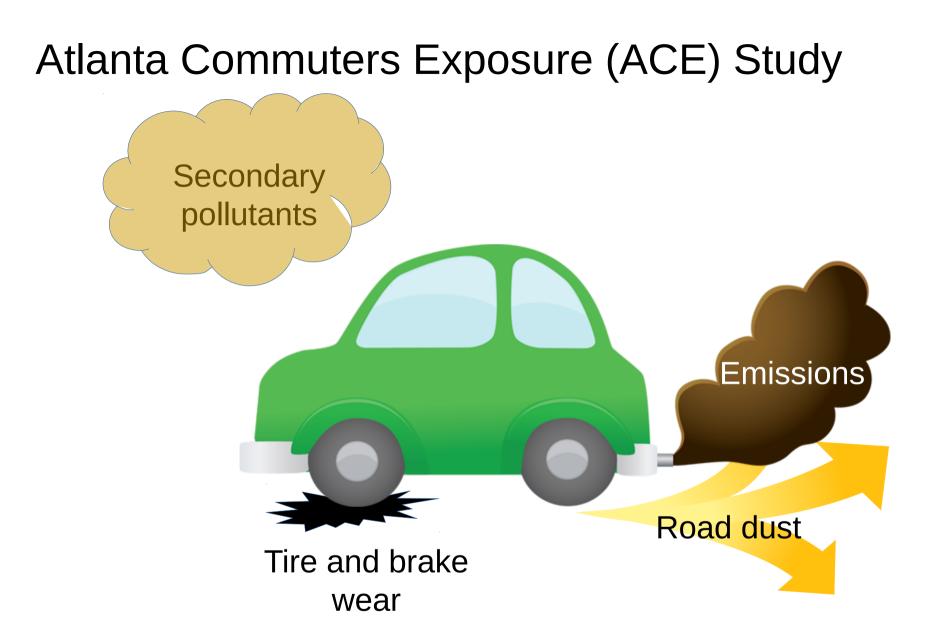


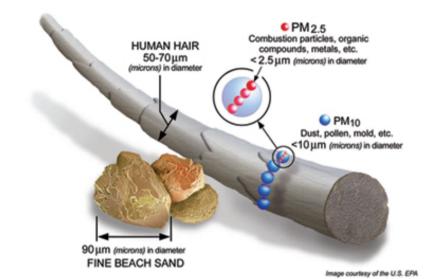
Image: LibreOffice

# Atlanta Commuters Exposure (ACE) Study

PM pollutants

- Particulate matter (PM) less than
  2.5 µm (PM2.5)
- Particle number concentration (PNC)
- Particle-bound polycyclic aromatic hydrocarbons (pbPAH)
- Chemical elements, including metals (Zn, Cd, Ni)
- Carbon-containing constituents
  - black carbon (BC)
  - Water-soluble organic carbon (OC)

#### Noise

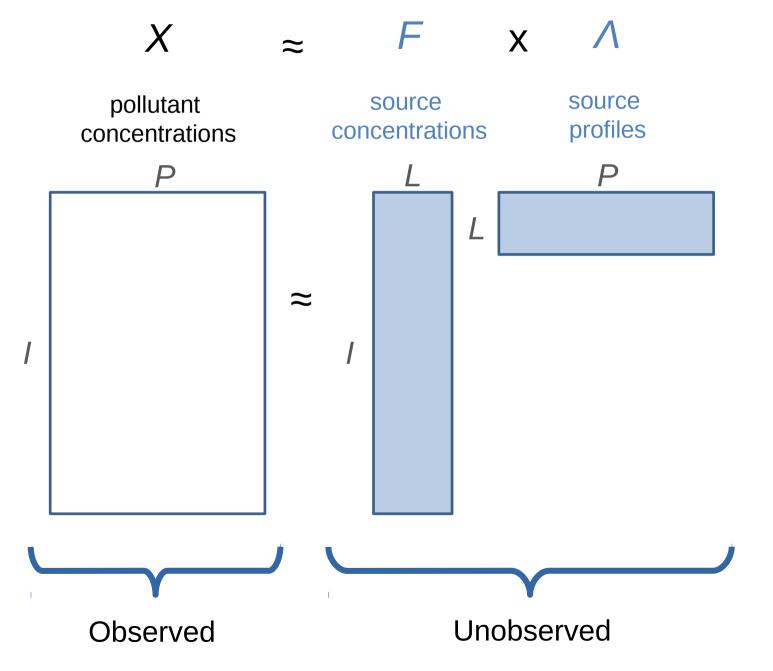


# Pollution data

#### Concentrations in $\mu$ g/m<sup>3</sup>

	Obs.	PM <sub>2.5</sub>	Organic carbon	Elemental carbon	Nitrate	Sulfate			
	1	12.6	2.3	0.8	2.3	2.1			
	2	15.7	3.8	1.2	2.7	2.6			
	3	13.9	3.1	1.2	2.2	3.3			
	4	14.5	4	1.7	2.7	3.3			
	5	16.6	5.4	1.9	2.1	2.8			
	6	8.6	1.7	0.5	0.5	2.9			
	7	8.4	1.4	0.6	0.9	3			
	8	10	2	1.1	1.1	2.2			
			Polluta	ant Emissions					
Image: nps.g									

### Source apportionment models



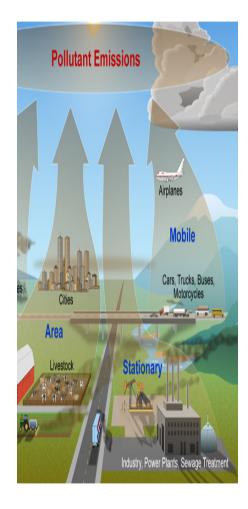
# Challenges in source apportionment (SA)

Need to estimate:

- F source concentrations
- A source profiles

Challenges

- Choosing number of sources L
- Deciding on prior information
  - SA models are not identifiable
  - $X = F\Lambda = F RR^{T} \Lambda = F^{*}\Lambda^{*}$
- Naming sources
  - Understanding major contributors
  - Attributing health effects



# Positive matrix factorization (PMF)

Find F and  $\Lambda$  that minimize

$$Q = \sum_{i=1}^{I} \sum_{p=1}^{P} \frac{\left(x_{ip} - \sum_{l=1}^{L} f_{il} \lambda_{lp}\right)^{2}}{u_{ip}^{2}}$$

such that  $f_{il} > 0$  and  $\lambda_{lp} > 0$  for all *i*,*l*,*p* where

- $x_{ip}$  Amount of pollutant p for observation i
- $f_{il}$  Amount of source *l* for observation *i*
- $\lambda_{lp}$  Amount of pollutant *p* in source *l*
- $u_{ip}$  Known observation-specific uncertainties corresponding to  $x_{ip}$
- No prior information is needed

# Positive matrix factorization (PMF)

- PMF has a graphical user interface
  - Generally finds solutions that minimize Q without additional constraints
  - Run PMF multiple times, find smallest Q
  - Uses the multilinear engine (ME) to solve

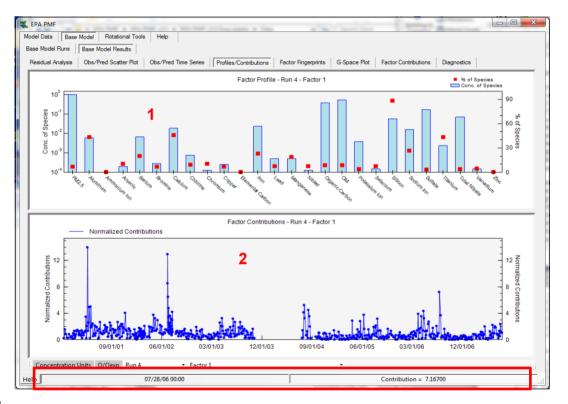


Image: epa.gov

# Multilinear engine (ME-2)

#### ME-2

- more flexible: can incorporate constraints
- minimizes  $Q + Q^a$ , where  $Q^a$  consists of auxiliary equations:

$$Q^{a} = \sum_{j=1}^{J} Q_{j}^{a} = \sum_{j=1}^{J} (r_{j}/s_{j})^{2}$$

where

#### $r_j$ is the residual of equation j $s_j$ is the "softness" of the equation

 $r_i$  may be something like  $(f_{il} - f^*)$ , where  $f^*$  is some fixed value.

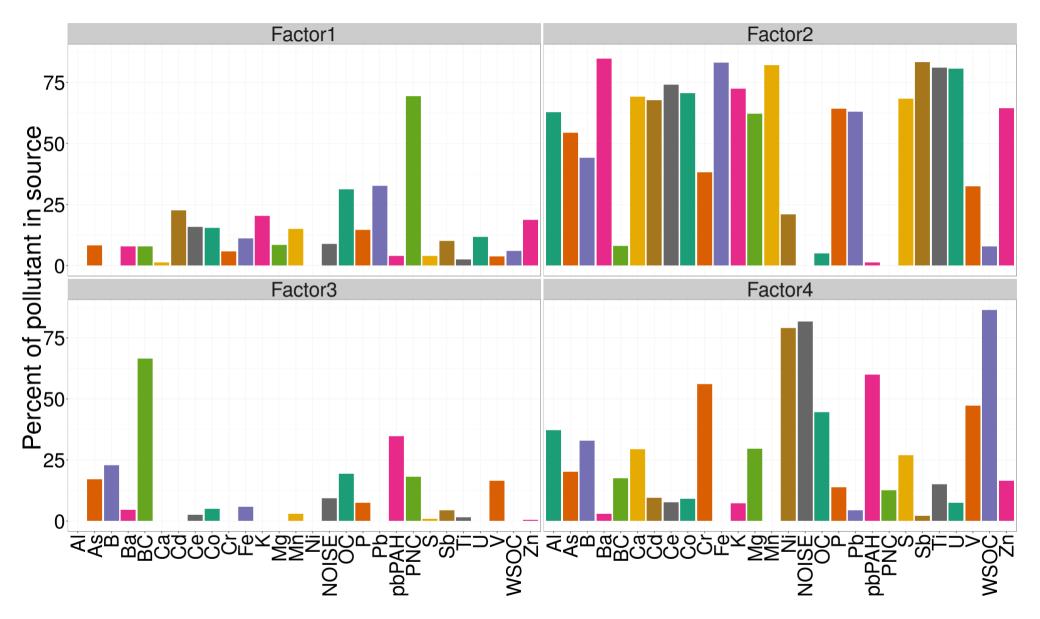
### Health model

Treat estimated source concentrations  $f_{ij}$  as known in standard health effects regression model:

$$Y_{it} = \beta_0 + \beta_1 f_{il} + \beta_2 z_i + \beta_3 f_{il} z_i + \gamma t + b_i + \epsilon_{it}$$

- $Y_{it}$  Difference from baseline for an outcome for observation *i* at time *t*
- $f_{ii}$  (estimated) Amount of source I for observation i
- $z_i$  Asthma status (1 = yes) for observation *i*
- $b_i$  Random intercept
- $\epsilon_{it}$  Random error

### Results: source profiles $\Lambda$ from PMF

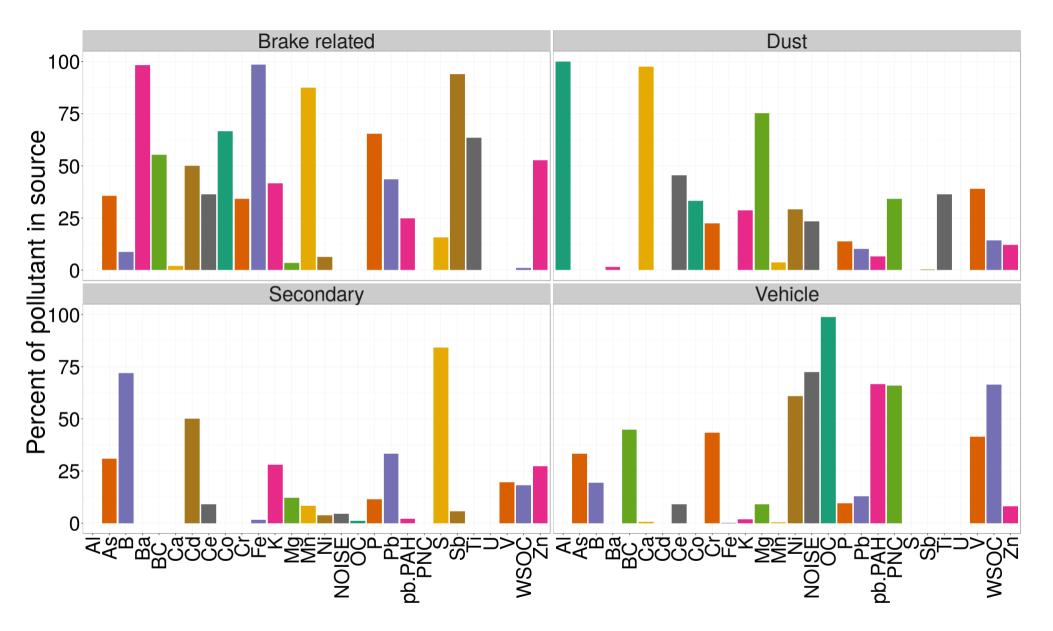


# Multilinear engine (ME-2)

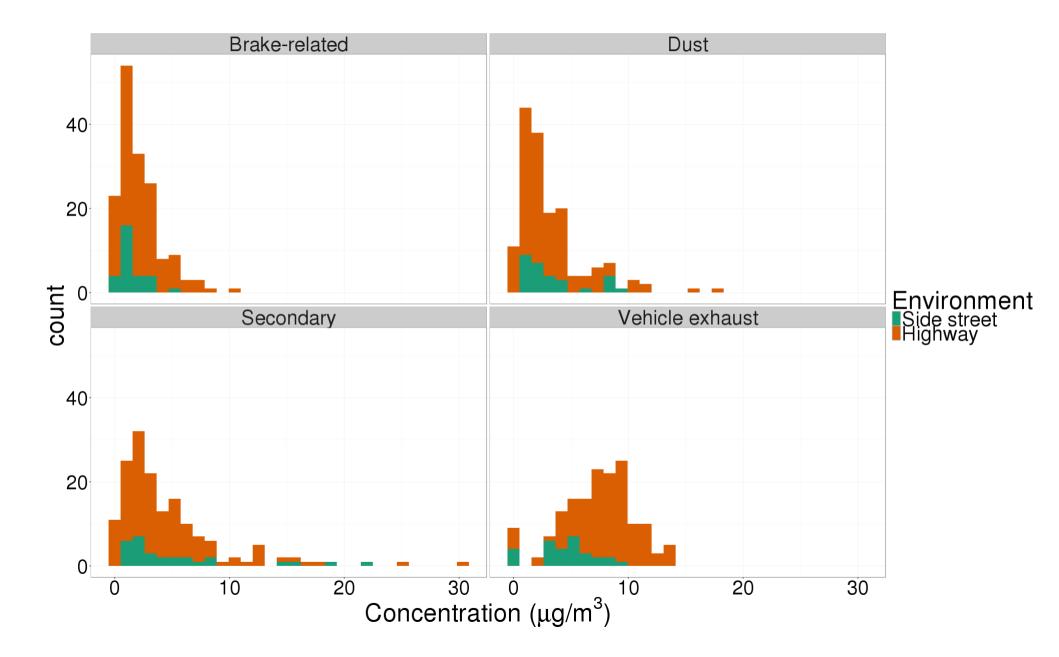
- What auxiliary equations to include?
- Sources:
  - Brake-related (high in metals elements: Zn, Mg)
  - Dust (high in crustal constituents: AI, Ca)
  - Secondary (high in sulfate)
  - Vehicle exhaust (high in carbon-containing constituents: OC, BC)

Source	Brake-related	Dust	Secondary	Vehicle exhaust
Contributing pollutants	Zn, Fe	Al, Ca	S	OC, BC
ME-2 constraints	↑ Zn, Fe	↑ Al, Ca, noise	↑ OC, WSOC	
		↓ WSOC	↓ Al, Ca	↓ OC, BC

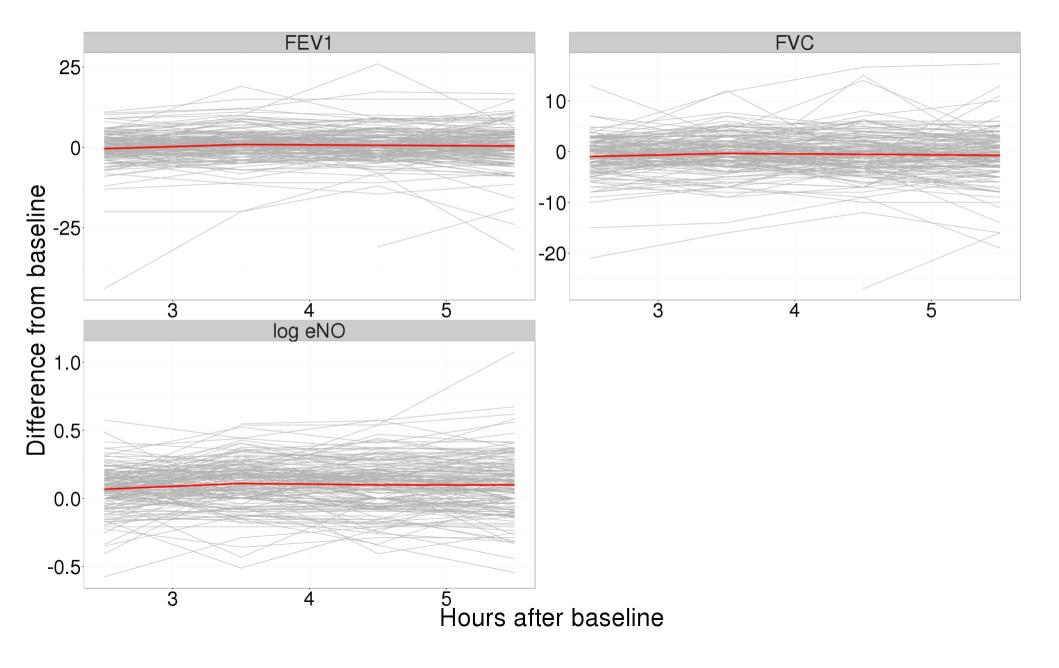
### Results: source profiles $\Lambda$ from ME-2



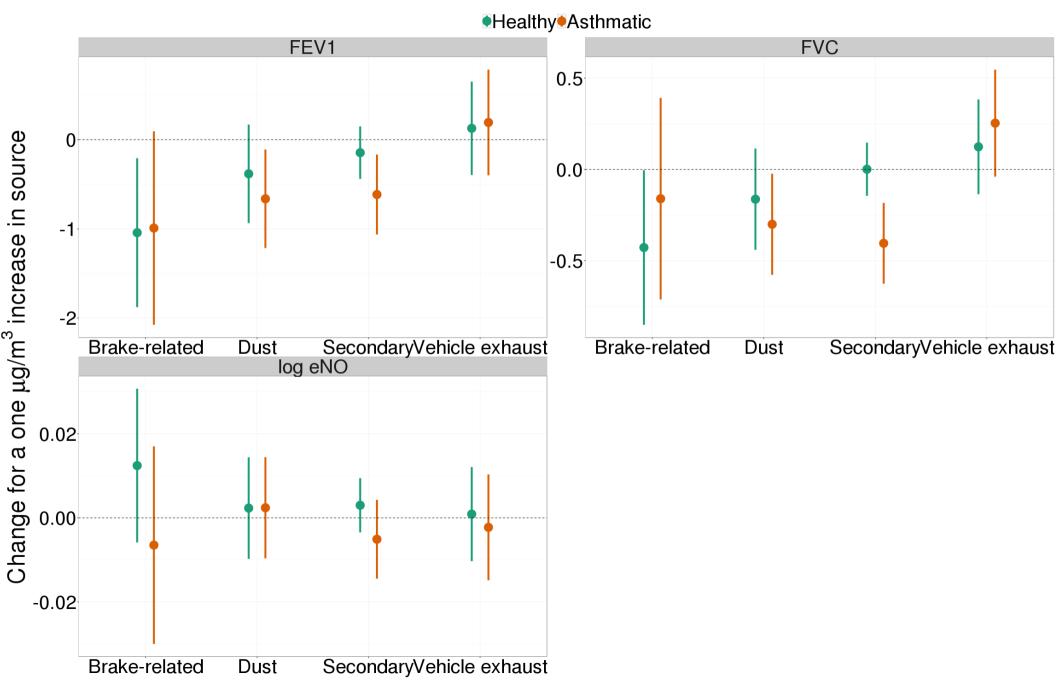
### Results: source concentrations F from ME-2



### Results: longitudinal health data



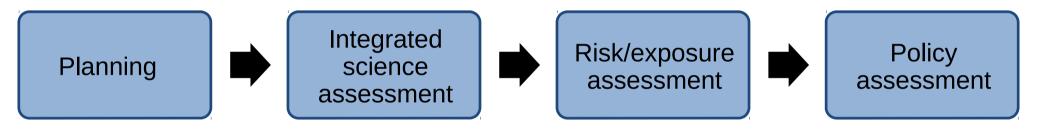
# Results: estimated effects of sources on health



# Air pollution policies

Some evidence of associations between exposure to secondary pollution and pollution from dust and lung function in individuals with asthma

National ambient air quality standards



#### **Emissions-specific policies**

Light-Duty Vehicles, Light-Duty Trucks, and Medium-Duty Passenger Vehicles: Tier 2 Exhaust Emission Standards and Implementation Schedule

Standard	Emission Limits at 50,000 miles					Emission Limits at Full Useful Life (120,000 miles) <sup>2</sup>				
Stanuaru	NOx (g/mi)	NMOG (g/mi)	CO (g/mi)	PM (g/mi)	HCHO (g/mi)	NOx (g/mi)	NMOG (g/mi)	CO (g/mi)	PM (g/mi)	HCHO (g/mi)
Bin 1	-	-	-	-	-	0	0	0	0	0
Bin 2		-	-	-	-	0.02	0.01	2.1	0.01	0.004
Bin 3	-	-	-	-	-	0.03	0.055	2.1	0.01	0.011
Bin 4	-	-	-	-	-	0.04	0.07	2.1	0.01	0.011
Bin 5	0.05	0.075	3.4	-	0.015	0.07	0.09	4.2	0.01	0.018
Bin 6	0.08	0.075	3.4	-	0.015	0.1	0.09	4.2	0.01	0.018

Image: epa.gov

# Conclusions

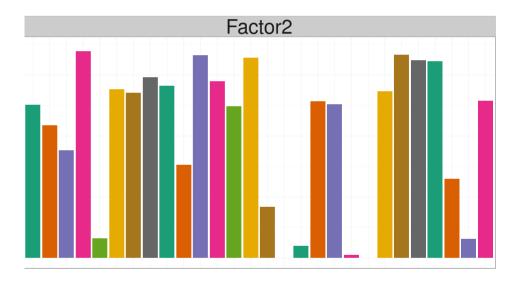
Some limitations:

- Healthy, young population
- Short-term exposure
- Lung function measurements
  - Biomarkers of inflammation
  - Cardiovascular health measures

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Evaluating source apportionment models

- Estimated sources may not align well with known sources
- Incorporating prior information
- Validation of results



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### Covariate summary

var	commute	n	% missing	mean	median	sd	min	max
age	0	73	5.2	30.72	28.13	9.03	0.33	59.47
age	1	79	6	30.43	28.33	8.86	0.08	58.44
bmi	0	56	27.3	23.8	23.12	4.08	17.82	38.98
bmi	1	79	6	23.76	23.29	4.08	16.7	38.98
cort_1	0	54	29.9	680.82	453.21	599.37	14.73	2886.51
cort_1	1	61	27.4	742.23	466	622.74	14.73	2886.51