

# Effects of Confounder Control and Causal Modelling in MAPLE, ELAPSE, and Medicare Cohorts

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ELAPSE



**HARVARD T.H. CHAN**  
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# Introduction

**Goal:** all three studies assess health effects of long-term exposure to low levels of ambient air pollution, PM<sub>2.5</sub>.

- Ideally you would randomize patients to low/high air pollution, but not feasible.
- Observational studies (e.g. claims data) have limitations due to lack of randomization.
  - Factors that are associated both with both exposures and health outcomes (e.g. SES-related factors) may **confound** exposure comparisons.

# Challenges

## 1) Unmeasured confounders

- Individual level covariates may confound exposure comparisons
- Some individual level covariates are missing across all three studies
- Each team approached this slightly differently

## 2) Adjusting for measured confounders

- Confounders can be included as linear terms in statistical models → could be sensitive to model misspecification

# MAPLE Overview

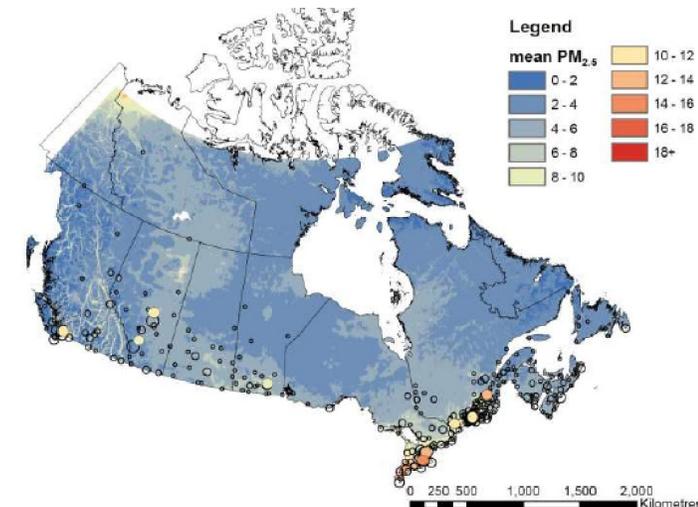


- **Canadian Census Health and Environment Cohort (CanCHEC):** three census-based cohorts (1991, 1996, 2001) of 7.1 million respondents, 128 million person-years with individual- and contextual-level covariates
- **Canadian Community Health Survey (CCHS):** additionally has information about health behaviors (450,900 respondents and 4.4 million person-years)
- Age  $\geq 25$  y at baseline, censored at 89 y
- Non-accidental mortality

**Individual covariates:** income quintile, visible minority status, Indigenous identity, labor force status, marital status, educational attainment

**Contextual covariates:** community size, airshed, urban form, neighborhood marginalization (instability, deprivation, dependency ethnic concentration)

**Behavioral covariates:** smoking behaviors, alcohol consumption, body mass index category, fruit and vegetable consumption, exercise behaviors





# MAPLE Covariate Adjustment

- Relationship between **PM<sub>2.5</sub>** and **non-accidental mortality** was robust to the addition of behavioral covariates in the CCHS

CanCHEC (1991, 1996, 2001)	HR	Lower 95%CI	Upper 95%CI
Unadjusted model (stratified by age, sex, cohort)	1.20	1.013	1.027
+ Individual level covariates	1.063	1.055	1.070
+ Contextual covariates (Full model)	1.084	1.073	1.096
mCCHS	HR	Lower 95%CI	Upper 95%CI
Unadjusted model (stratified by age, sex, cohort)	1.002	0.966	1.039
+ Individual level covariates	1.034	0.997	1.073
+ Contextual covariates	1.123	1.056	1.194
+ Behavioural covariates (Full model)	1.086	1.021	1.155

HR estimates per 10  $\mu\text{g}/\text{m}^3$  increase in PM<sub>2.5</sub>



# MAPLE Indirect Adjustment

- Use an indirect adjustment via partitioned regression to adjust for unmeasured covariates (Shin et al., 2014).
- Method uses a representative ancillary dataset to estimate the association between variables missing in a primary dataset with the complete set of variables of the ancillary dataset to produce an adjusted risk estimate for the variable in question.
- Used the association between fine particle air pollution ( $PM_{2.5}$ ) with mortality in the 2001 CanCHEC (N = 2.4 million, 10-years follow-up) as primary dataset.
- Used the 2001 cycle of the CCHS (N = 80,630) as the ancillary matching dataset that contained confounding risk factor information not available in CanCHEC (e.g., smoking).

# MAPLE Indirect Adjustment: Results (Erickson et al, 2019)



- Results from the CCHS cohort analysis and the indirect adjustment suggest that behavioral covariates (e.g., smoking and diet) only slightly confounded the  $PM_{2.5}$ -mortality association.
- Based on sensitivity analyses in the 2001 CanCHEC cohort → missing data on behavioral covariates were unlikely to significantly confound the  $PM_{2.5}$ -mortality relationship in the Canadian population.

## Non-Accidental Mortality

	Unadjusted	Adjusted**	Weighted Adjustment
Model 1	1.103 (1.084 – 1.123)	1.135 (1.110 - 1.160)	1.131 (1.106 - 1.156)
mCCHS (Model 1)	1.177 (1.073 - 1.291)	1.202 (1.095 - 1.319)	
Model 2	1.201 (1.171 - 1.231)	1.161 (1.126 - 1.197)	1.178 (1.142 - 1.214)
mCCHS (Model 2)	1.163 (1.002 - 1.348)	1.136 (0.979 - 1.318)	

**Supplemental Table s3: Indirect adjustment of missing risk factors on  $PM_{2.5}$  hazard ratios (HR, 95% CI)\* for different causes of mortality with comparison to the CCHS-mortality cohort**

Model 1: Individual covariate model - stratified by 5-year age-sex groups, adjusted marital status, visible minority, Aboriginal identity, employment, income quintile, education, Model 2: Individual + ecological covariate model - Model 1, plus Can-Marg Index, Community Size, Airshed

mCCHS: equivalent model using CCHS-mortality linked cohort, adjusted model includes smoking, alcohol use, exercise, diet

\* HR estimates per 10  $\mu g/m^3$  increase in  $PM_{2.5}$  \*\*Model 1 and 2 are indirectly adjusted for smoking, alcohol use, exercise, diet

Weighted adjustment: indirect adjustment uses sample weight (W-matrix)

# ELAPSE Overview

- Large administrative cohorts from seven countries in Europe (N = 28 million)
- Age  $\geq 30$  y at baseline
- Non-accidental mortality



# ELAPSE Covariate Adjustment

- Specified three models to evaluate association between  $PM_{2.5}$  and non-accidental mortality;
  - **Model 1:** included only age (time axis), sex (as strata), and calendar year(s) of enrollment.
  - **Model 2:** added individual level variables available within each cohort.
  - **Model 3:** added to the model 2 area-level socio-economic status (SES) variables.

# ELAPSE Covariate Adjustment

## Non-Accidental Mortality

	Model 1	Model 2	Model 3
Pollutant	HR (95% CI)	HR (95% CI)	HR (95% CI)
<b>Belgian cohort (N=5,474,548)</b>			
PM2.5	0.960 (0.946, 0.974)	0.968 (0.956, 0.981)	1.023 (1.011, 1.035)
<b>Danish cohort (N=3,083,235)</b>			
PM2.5	1.221 (1.185, 1.257)	1.216 (1.190, 1.243)	1.141 (1.118, 1.164)
<b>Dutch cohort (N=10,465,727)</b>			
PM2.5	1.062 (1.034, 1.090)	1.036 (1.013, 1.059)	1.021 (0.999, 1.044)
<b>English cohort (N=1,368,740)</b>			
PM2.5	1.114 (1.090, 1.137)	1.082 (1.059, 1.105)	1.023 (1.001, 1.045)
<b>Norwegian cohort (N=2,309,001)</b>			
PM2.5	1.034 (1.018, 1.050)	1.071 (1.061, 1.082)	1.076 (1.066, 1.086)
<b>Rome cohort (N=1,263,712)</b>			
PM2.5	0.964 (0.888, 1.047)	1.019 (0.969, 1.071)	1.066 (1.033, 1.099)
<b>Swiss cohort (N=4,188,175)</b>			
PM2.5	0.979 (0.966, 0.993)	1.007 (0.997, 1.018)	1.026 (1.015, 1.038)

HR estimates per 5 µg/m<sup>3</sup> increase in PM<sub>2.5</sub>

- In all cohorts, HRs were sensitive to more complete adjustment for potential confounders.
- Adjustment for individual and especially area-level confounders increased the HRs in some cohorts (e.g. Rome, Swiss, Norwegian) and decreased HRs in other cohorts (e.g. Dutch, English).

# ELAPSE Indirect Adjustment

- Applied the indirect adjustment method proposed by Shin et al., 2014.
- Applied the Shin method for natural cause mortality and the key potential confounders smoking status and BMI.
- In the Netherlands, Switzerland, Norway, Belgium and Denmark data from large population surveys on variables such as smoking, alcohol use, exercise and BMI were available.
  - English cohort has individual data on smoking and BMI in full cohort → no indirect adjustment needed.

# ELAPSE Indirect Adjustment: Results

- In the Dutch, Swiss and Norwegian cohort, HRs were attenuated but remained (borderline) significant.
- In the Rome and Belgian cohort, HRs increased after indirect adjustment.

Table 22 Hazard ratios for associations between PM<sub>2.5</sub> and NO<sub>2</sub> and **natural-cause mortality** in six administrative cohorts: indirect adjustment, adjustment for area-level disease

	Cohort	HR, main Model	HR with indirect adjustment	HR, main Model, reduced population area level adjustment	HR with area-level lung cancer, COPD, diabetes
PM <sub>2.5</sub>	Belgian	1.023 (1.011, 1.035)	1.049 (1.036, 1.062)	NA	1.029 (1.017, 1.041) <sup>1</sup>
	Danish	1.141 (1.118, 1.164)	1.118 (1.095, 1.140)	NA	1.143 (1.120, 1.167) <sup>2</sup>
	Dutch	1.021 (0.999, 1.044)	1.015 (0.993, 1.038)	NA	1.007 (0.985, 1.030) <sup>3</sup>
	Norwegian	1.076 (1.066, 1.086)	1.055 (1.045, 1.065)	1.095 (1.080, 1.109) <sup>4</sup>	1.081 (1.067, 1.095) <sup>1</sup>
	Rome	1.066 (1.033, 1.099)	1.111 (1.080, 1.142)	NA	1.041 (1.009, 1.075) <sup>3</sup>
	Swiss	1.026 (1.015, 1.038)	1.015 (1.003, 1.027)	NA	NA

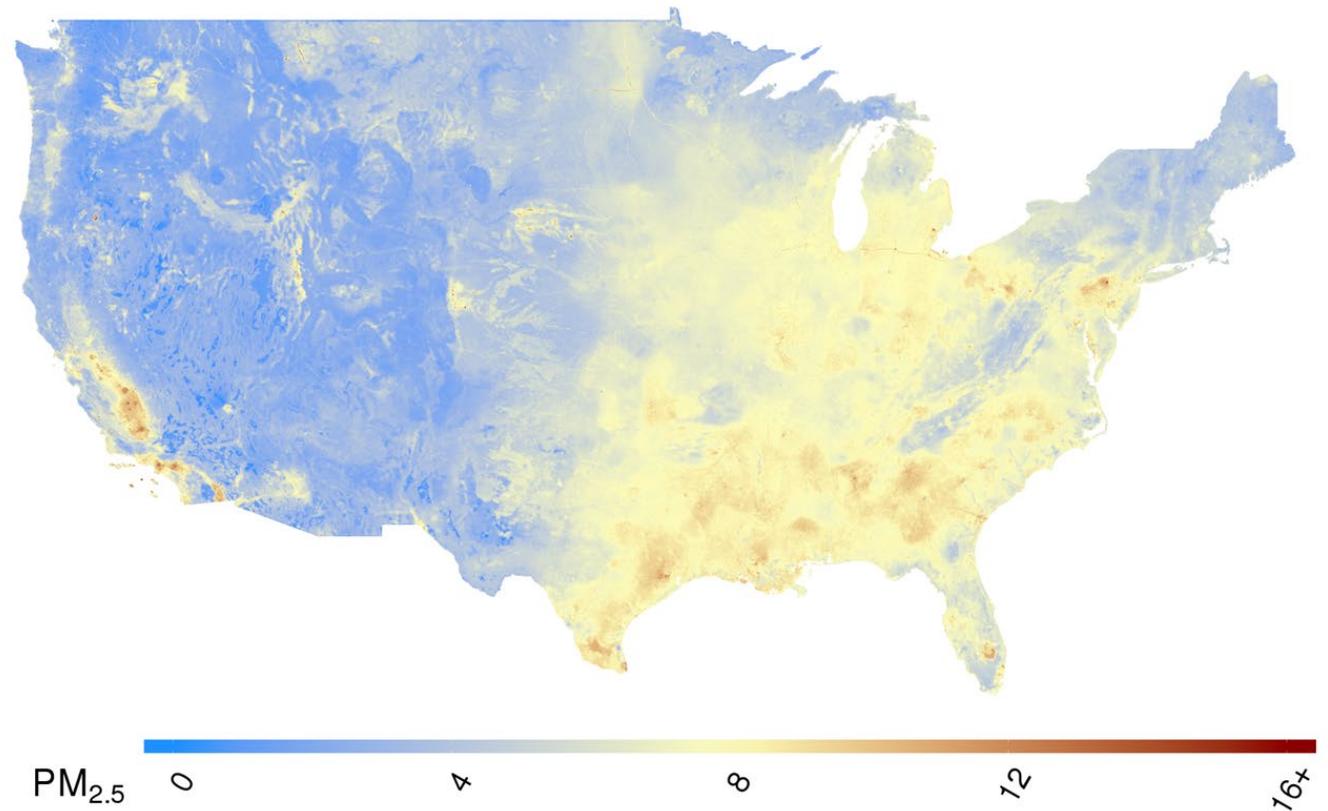
HR estimates per 5 µg/m<sup>3</sup> increase in PM<sub>2.5</sub>

# Harvard Overview



- Medicare enrollees (N = 68.5 million)
- Age  $\geq 65$  y
- All-cause mortality

Annual Average of PM<sub>2.5</sub> per  $\mu\text{g}/\text{m}^3$  in 2016

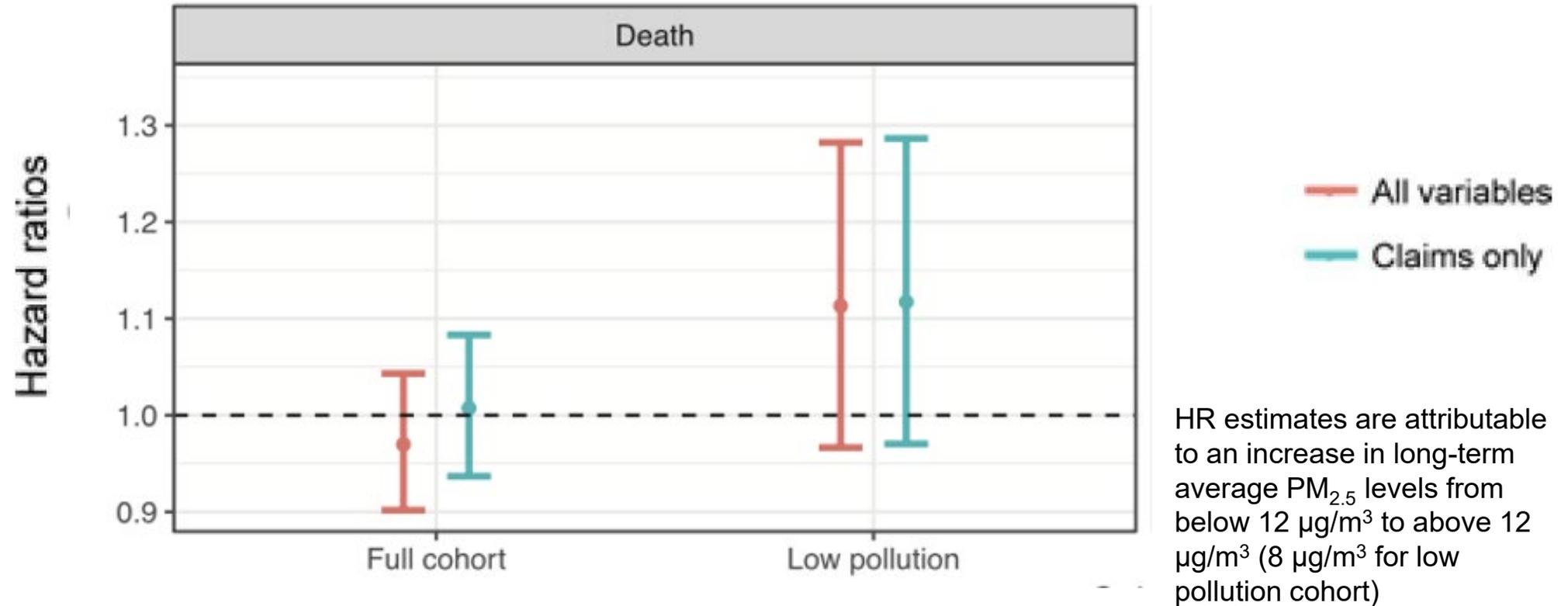




# Harvard Covariate Adjustment

- Medicare data contain little information about individual level covariates.
- Obtained data from Medicare Current Beneficiary Survey (MCBS), which is an annual phone survey of a nationally representative sample of Medicare beneficiaries
  - contains information on more than 150 potential individual confounders: individual risk factors (e.g., smoking, body mass index [BMI], and income)
- Analyzed data for a cohort of ~32,000 beneficiaries from the MCBS-Medicare linked dataset.

# Harvard Covariate Adjustment: Results (Makar et al, 2017)



**Figure 4.** Sensitivity to exclusion of MCBS variables: Hazard ratios and 95% confidence intervals based on robust, sandwich variance estimators computed including (red) and excluding (blue) MCBS variables.

# Causal Inference Adjustment for Measured Confounders

Causal inference methods are designed to bridge the gap between observational studies and randomized experiments → are more robust to model misspecification



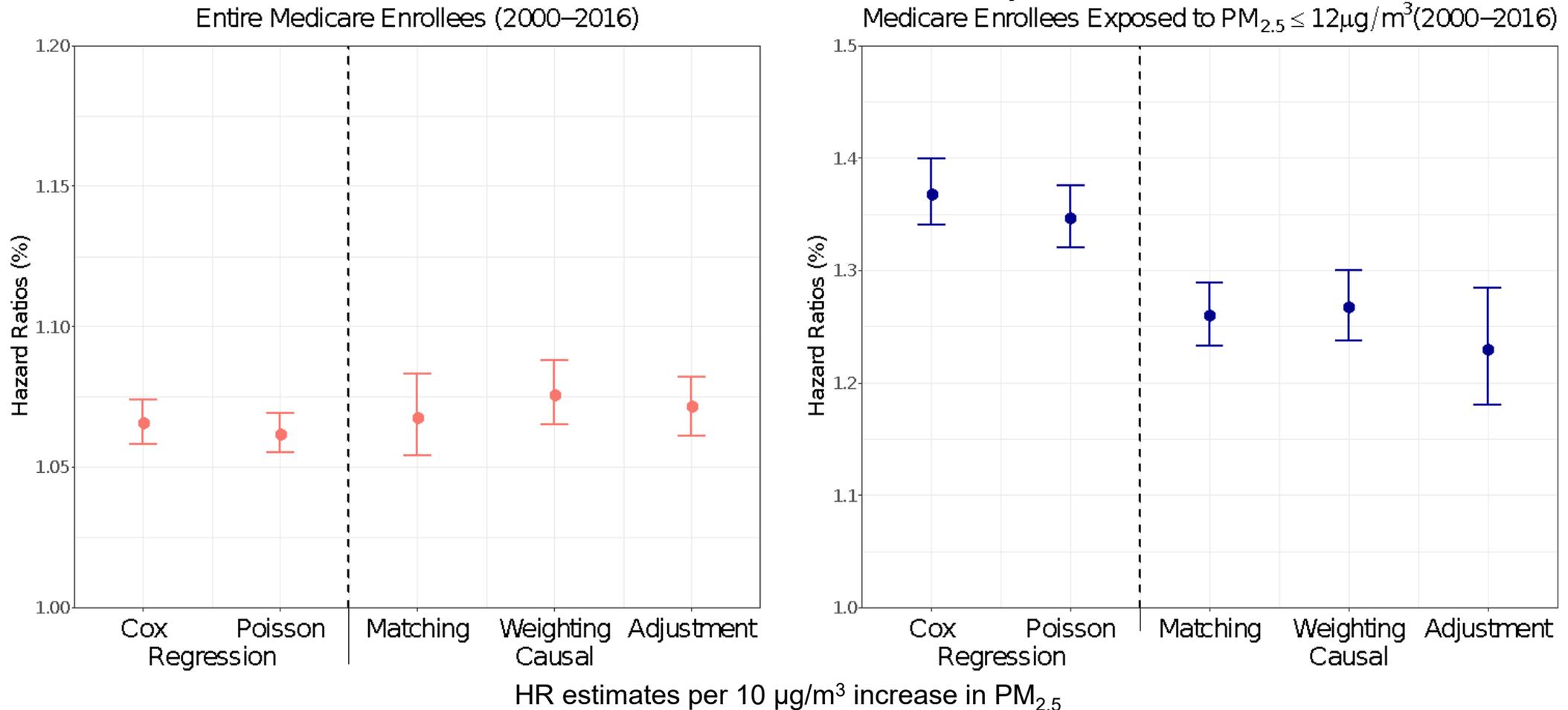
# Harvard Causal Inference Modeling

- Applied various traditional and causal inference approaches, conducted extensive sensitivity analyses.
- Relied on publicly available data and provide code that allows for reproducibility [https://github.com/wxwx1993/National\\_Causal](https://github.com/wxwx1993/National_Causal).



# Harvard Causal Inference Modeling: Results (Wu et al, 2020)

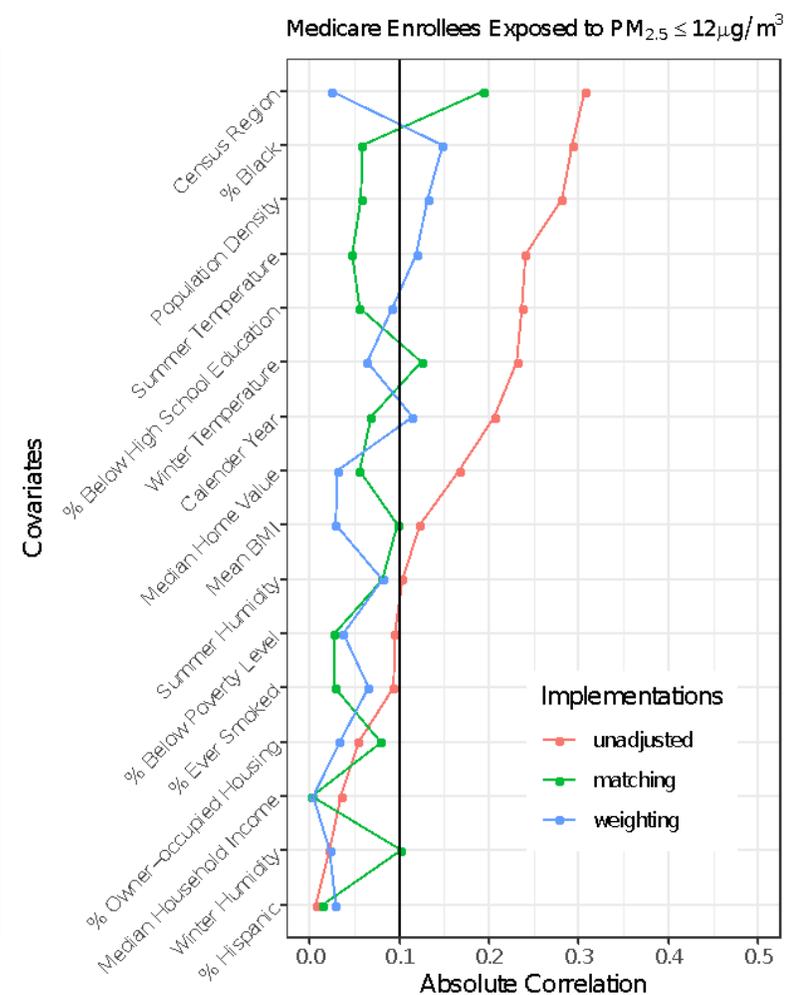
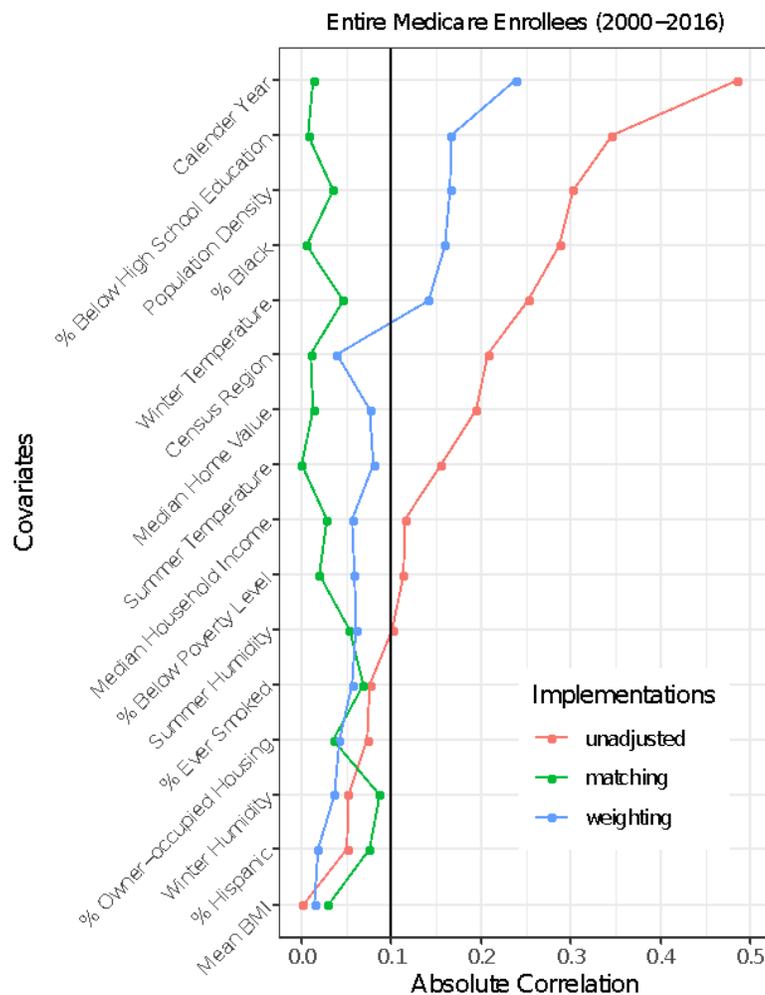
## All-cause mortality





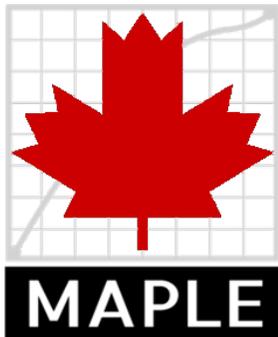
# Harvard Causal Inference Modeling: Results (Wu et al, 2020)

- Covariate balance used to assess the quality of the causal inference approaches at recovering randomized experiments.
- Evaluate covariate balances for each confounder.



# Conclusions

- All three studies; missing data on covariates were unlikely to significantly confound the  $PM_{2.5}$ –mortality relationship
- In Medicare, five approaches (traditional and causal) lead to similar results → robust and reproducible evidence on causal link between PM and mortality



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# Ongoing Work

- All three teams are working on harmonizing confounders and models across the datasets
- ELAPSE and MAPLE working on applying causal inference approaches



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**Additional Slides**