Compounding Benefits of Air Pollution Control

A Revised View of Air Pollution Economics

Amanda J. Pappin, Amir Hakami --- Carleton University

Philip Blagden, Masoud Nasari, Mieczyslaw Szyszkowicz, Richard T. Burnett --- Health Canada

Applying Epidemiology in Health Impact Assessment

Integrate risk estimates from epidemiological studies with air quality models to estimate

The distribution of health impacts (affected populations), or

Applying Epidemiology in Health Impact Assessment

Integrate risk estimates from epidemiological studies with air quality models to estimate

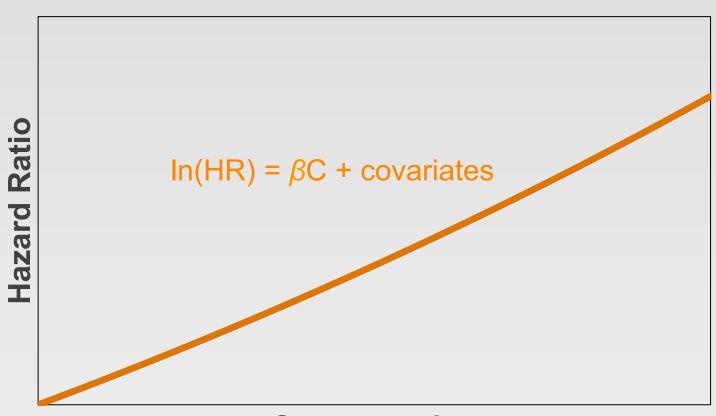
- The distribution of health impacts (affected populations), or
- Source attribution of health impacts (emissions responsible)

Identifying Sources of Health Impacts

$$\frac{\Delta\$}{\Delta \text{Emissions}} = \frac{\Delta\$}{\Delta \text{Mortality}} \times \frac{\Delta \text{Mortality}}{\Delta \text{Concentrations}} \times \frac{\Delta \text{Concentrations}}{\Delta \text{Emissions}}$$
Monetized benefits
Economics
Epidemiology
Air quality modeling

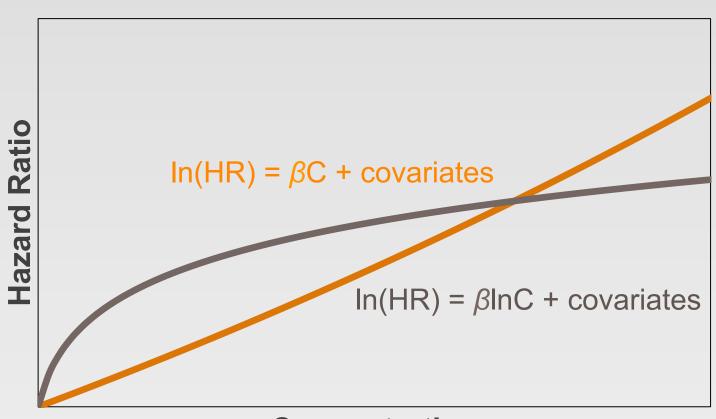
-- per-ton basis (benefits-per-ton or marginal benefit)

Traditional Linear CRF



Concentration

Linear vs Log-linear CRFs



Concentration

Traditional Health Impact Function

$$M(\$) = M_0 Pop(1-e^{-\beta C})V_{SL}$$

Mortality rate and population

Derived from Value of the CRF statistical life

--- best model fit for O₃ and mortality in Canada

Revised Health Impact Function

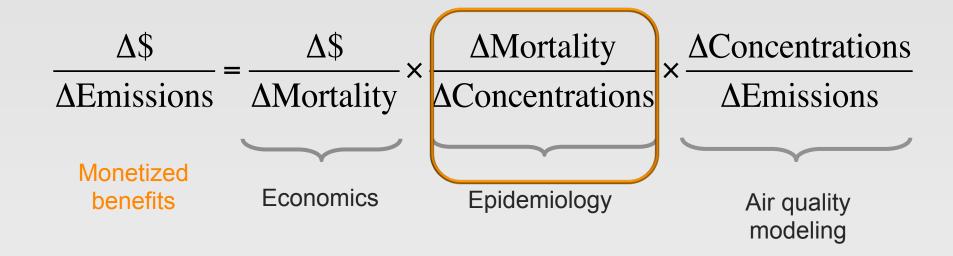
$$M(C) = M_0 Pop(1-e^{-\beta lnC})V_{SL}$$

Mortality rate and population

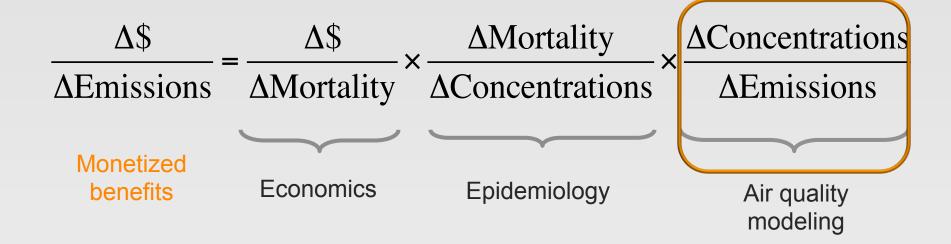
Derived from Value of the CRF statistical life

--- best fits for NO₂ and PM_{2.5} in Canada

Nonlinearity in Sensitivities

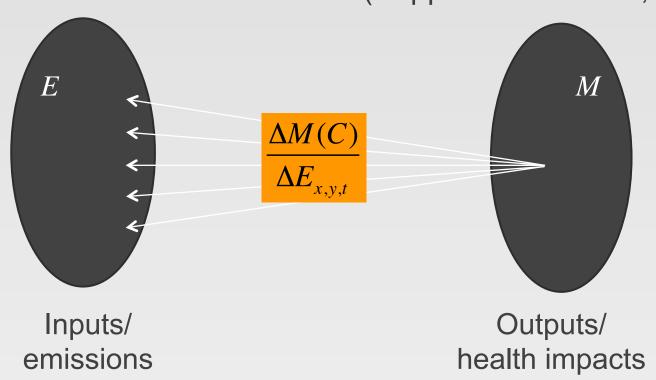


Nonlinearity in Sensitivities



Adjoint Air Quality Modeling

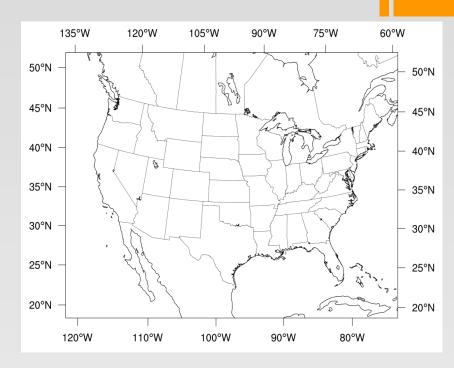
Tracing mortality backwards to emissions: where influences come from (Pappin and Hakami, 2013).



Case Study

CMAQ-Adjoint

- May-Sept 2007
- 36 km resolution
- SAPRC99



Case Study

Cost Function, J = monetized non-accidental mortality in Canada attributable to

- Long-term O₃ exposure
- Long-term NO₂ exposure

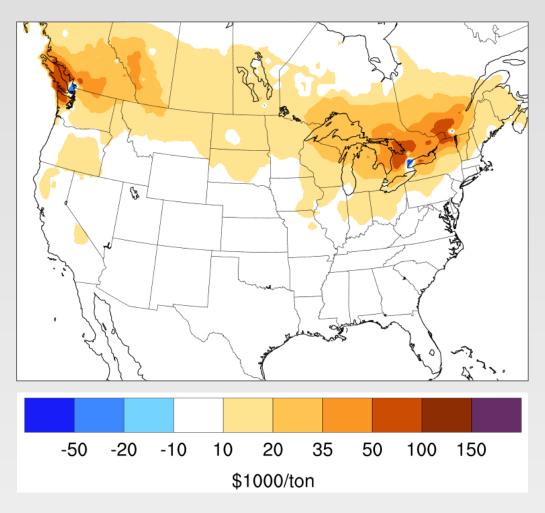
Canadian Epidemiological Data (Crouse et al. EHP 2015)

- $O_3 \beta = 0.0026 \text{ ppb}^{-1} \text{ (summertime average DM8A)}$
- $NO_2 \beta = 0.0059 \text{ ppb}^{-1} \text{ (summertime average)}$
- NO_2 log-linear β = 0.0732 (---note difficulty interpreting)

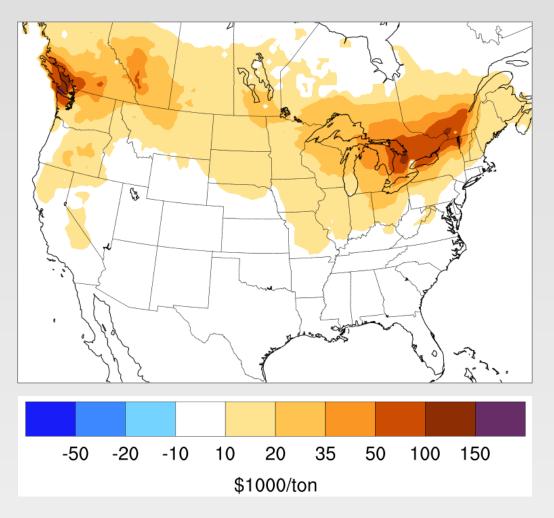
Findings: O₃ Mortality

- Linear CRF
- Non-linear atmospheric response

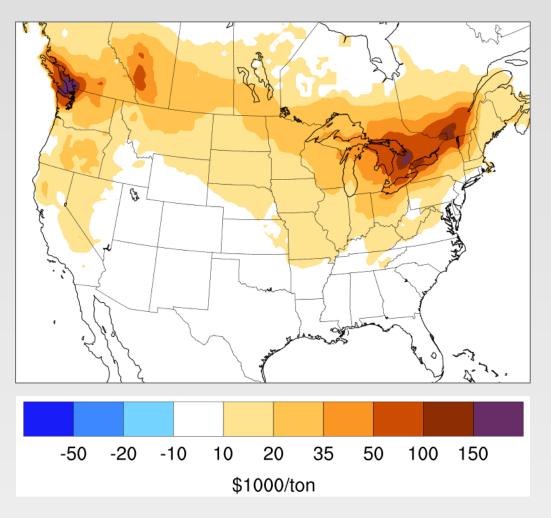
O₃, At 2007 Emission Levels



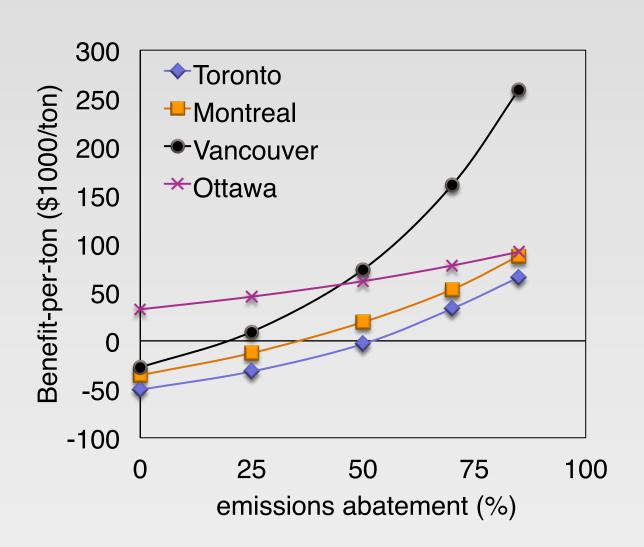
O₃, At 50% Emissions Abatement



O₃, At 85% Emissions Abatement

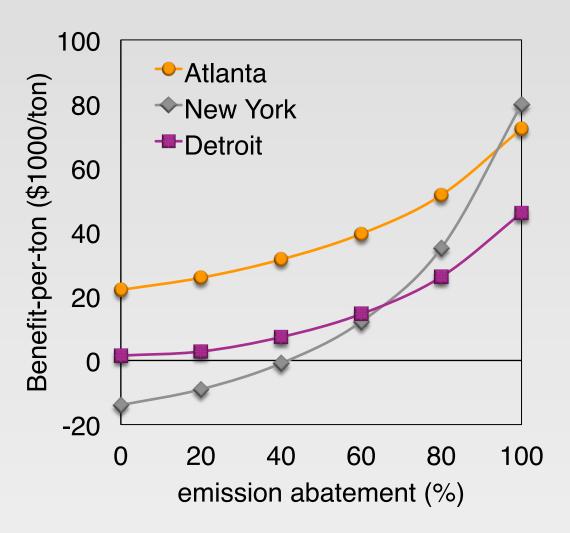


Nonlinear Behavior by Source



Similar Behavior in the U.S.

(Pappin et al. ES&T 2015)



Conclusions - 1

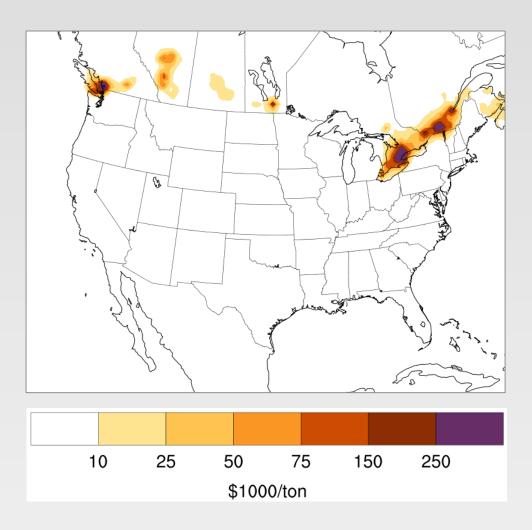
- Non-linearity in O₃ based benefits are due entirely to atmospheric chemistry
- This becomes increasingly important as we move towards lower pollution levels

Findings: NO₂ Mortality

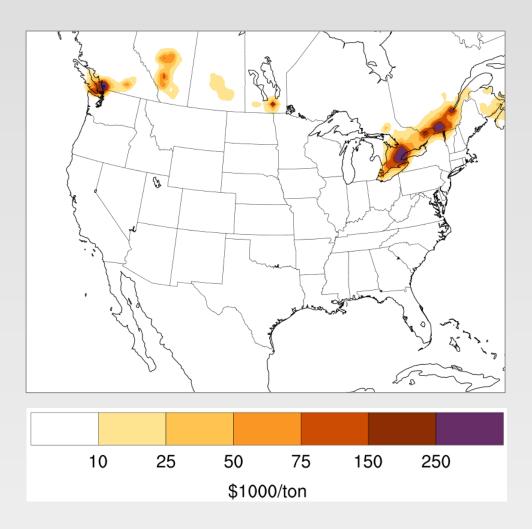
Linear vs log-linear CRFs

Traditional, Linear CRF

NO₂, At 2007 Emission Levels

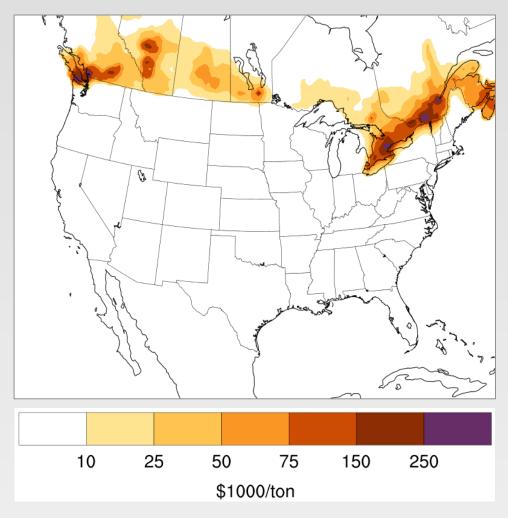


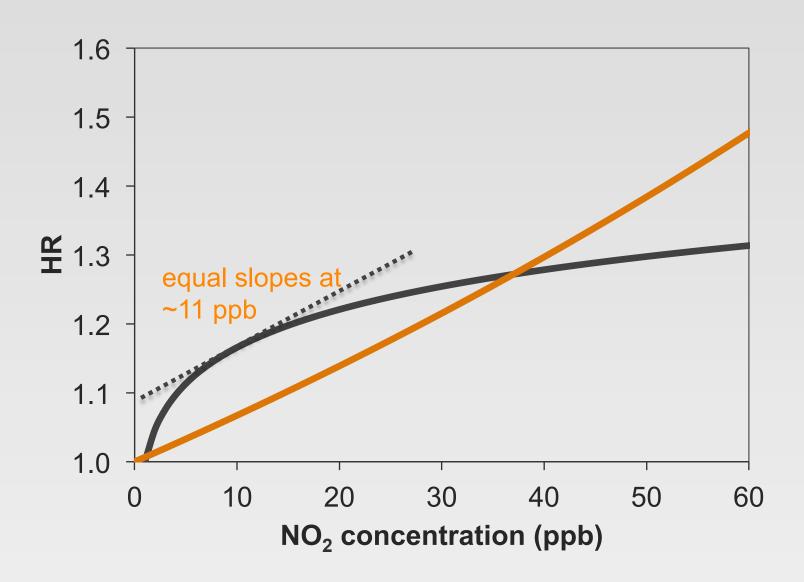
NO₂, At 85% Emissions Abatement



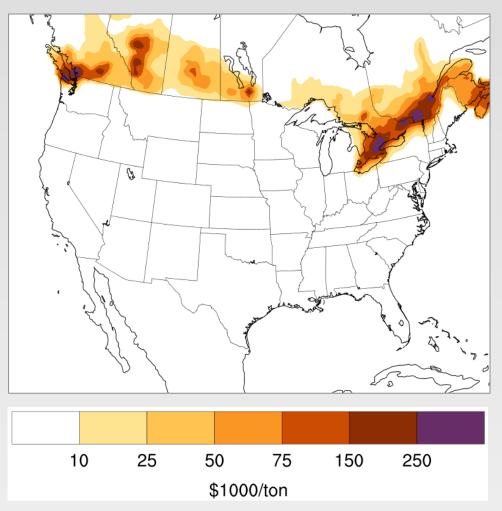
Log-linear CRF

NO₂, At 2007 Emission Levels

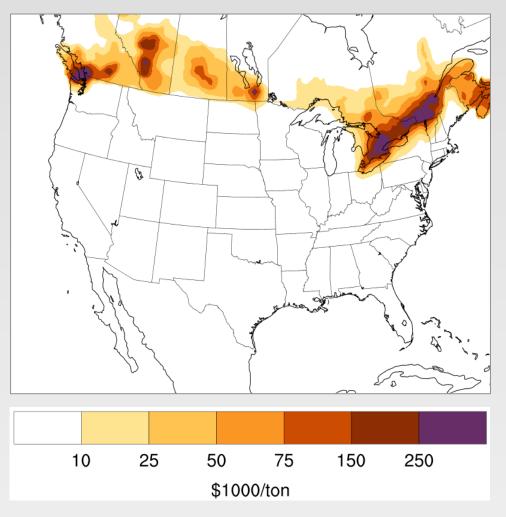




NO₂, At 50% Emissions Abatement

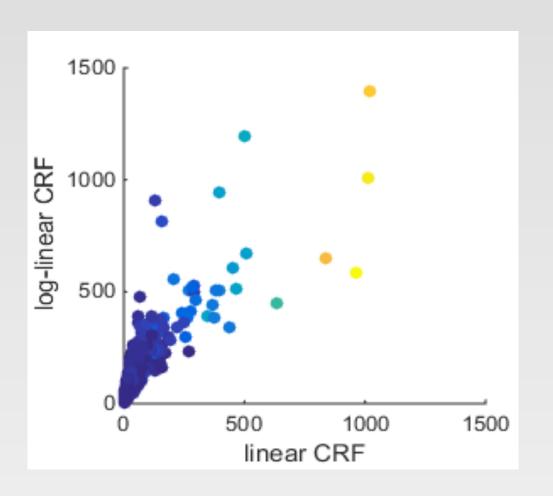


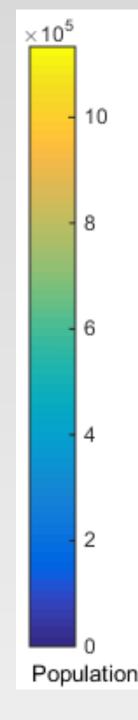
NO₂, At 85% Emissions Abatement



Linear vs Log-linear

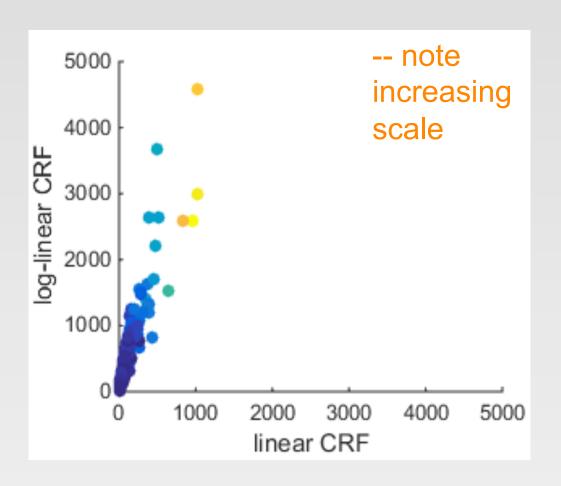
NO₂, At 2007 Emission Levels

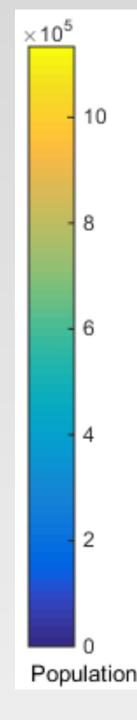




Linear vs Log-linear

NO₂, At 85% Emissions Abatement

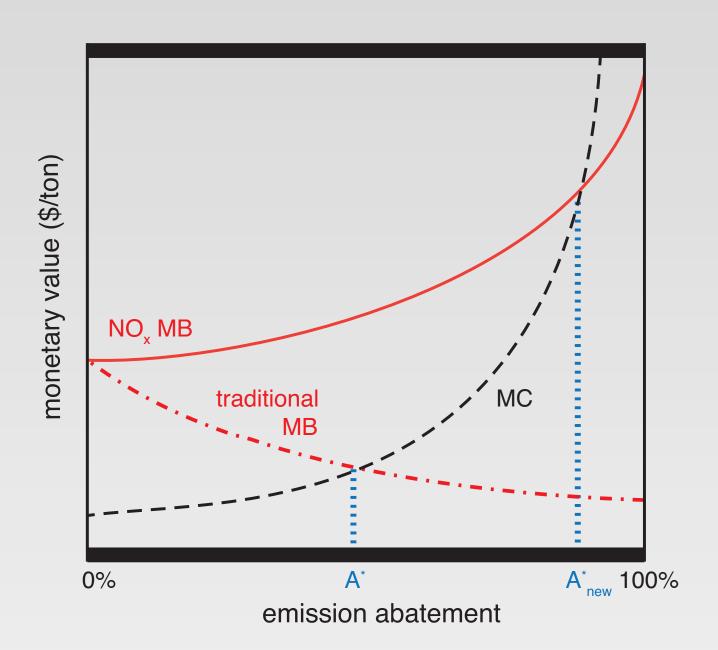




Conclusions - 2

- Important differences between linear and log-linear CRFs for NO₂, particularly in cleaner environments
- Benefits are larger for NO₂ than O₃

Policy Relevance



Pappin et al. ES&T 2015

Considerations for PM

- Indications of atmospheric nonlinearity for PM_{2.5} exist in the literature (Fann et al. 2012; Holt et al. 2015; Hakami et al. 2003; Zhang et al. 2012)
- Combined with a potentially non-linear CRF, benefits-per-ton for PM_{2.5} may increase substantially towards lower pollution levels
- Further research using a multiphase adjoint model can shed light on this

Acknowledgements

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Thank you!

Limitations

- Constant mortality rates assumed over time
- Long-term benefits (i.e., chronic exposure mortality)
 modeled in a short, 5-month simulation episode
- Uncertainty in atmospheric modeling, CRFs, and economic valuation lead to uncertainties in benefitper-ton estimates

Canadian Census, Environment, and Health Cohort (CanCHEC)

- 2.6 million subjects > 25 years of age
- O₃, NO₂, PM_{2.5} and mortality analyzed (various causes-of-death)
- Log-linear models appropriate for NO₂ and PM_{2.5}
- Linear model most appropriate for O₃

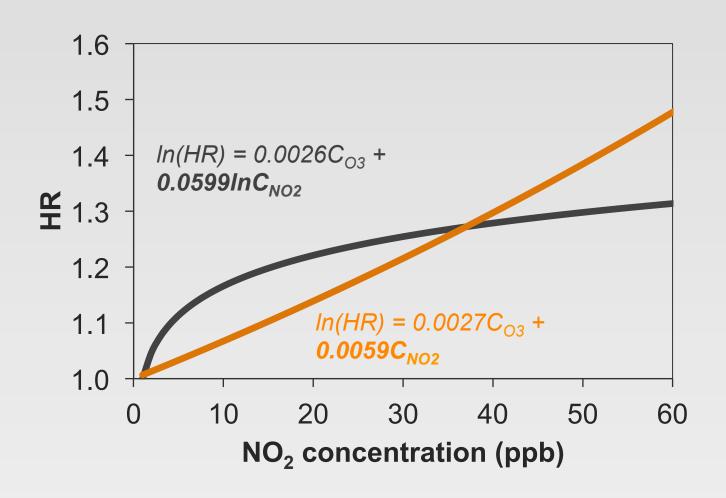
CanCHEC O₃ and NO₂ CRFs

Non-accidental mortality

- Linear: $In(HR) = 0.0027C_{O3} + 0.0059C_{NO2} + covariates$
- Log-linear: $In(HR) = 0.0026C_{O3} + \frac{0.0599In(C_{NO2}+1)}{0.0599In(C_{NO2}+1)} + covariates$

Linear and Log-linear NO₂ CRFs

Analysis of CanCHEC



Identifying Sources of Health Impacts

A question of sensitivity analysis

$$\frac{\Delta Mortality}{\Delta Emissions} = \frac{\Delta Mortality}{\Delta Concentrations} \times \frac{\Delta Concentrations}{\Delta Emissions}$$
Epidemiology
Air quality modeling