

Black Carbon Reduced Form Modeling through Green's Functions for Policy Scenarios

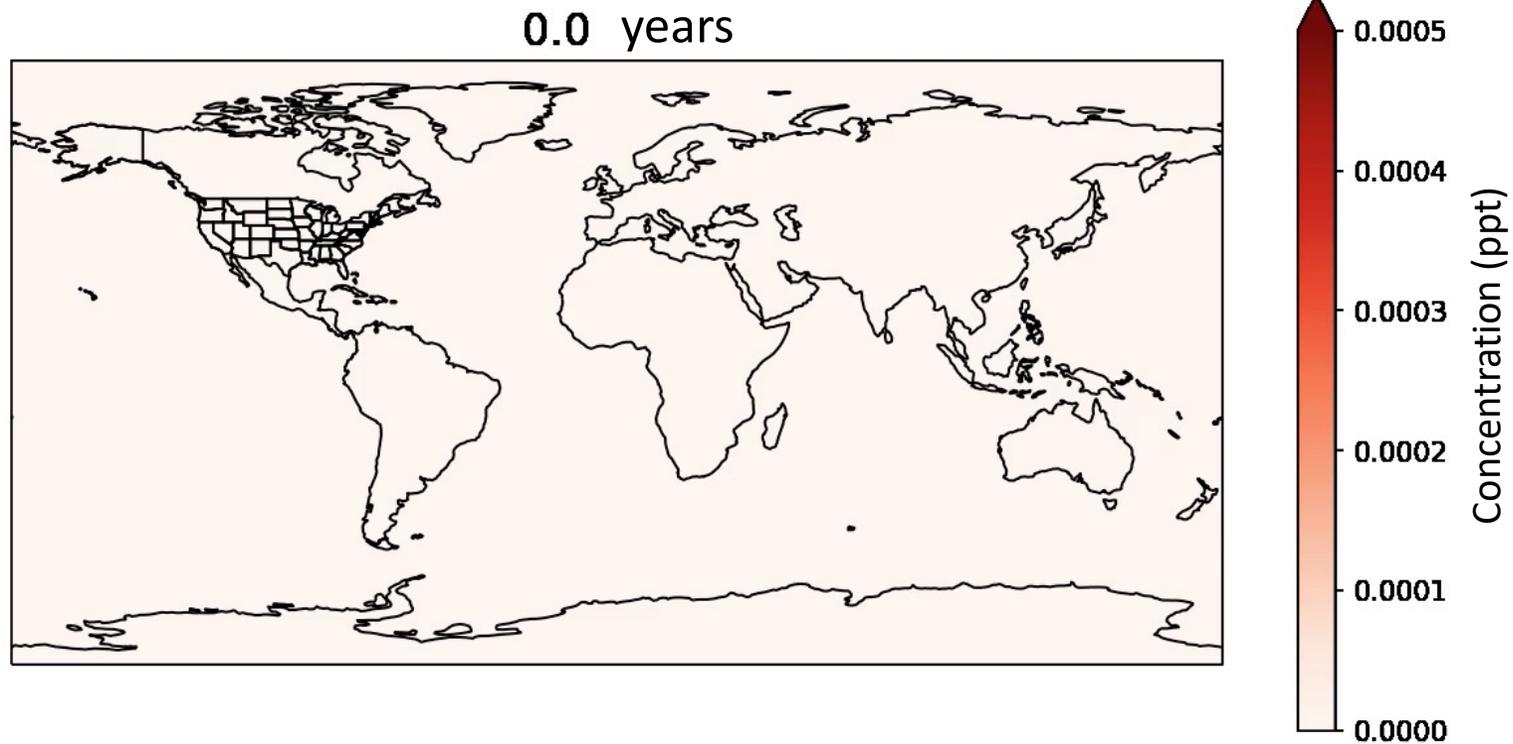
Lyssa Freese, Sebastian Eastham, Cecilia Han
Springer, Noelle Selin



MARTIN FAMILY
SOCIETY OF
FELLOWS FOR
SUSTAINABILITY

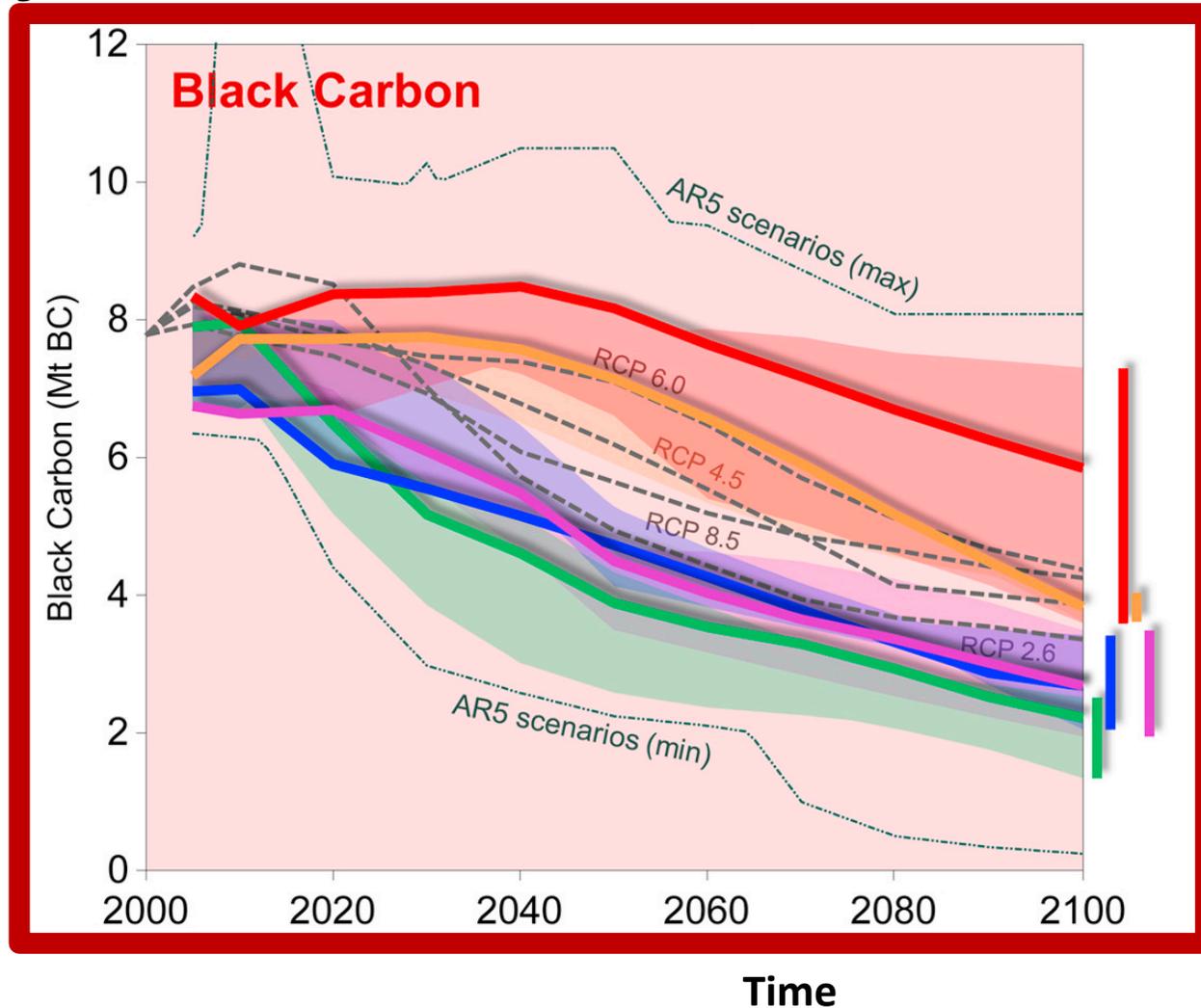


Reduced form model of Black Carbon



- Black carbon concentration from coal power plants in Southeast Asia
- 50 years of simulation
- 3 minutes to simulate in a Jupyter notebook

We need to be able to assess various policy trajectories



We take a few emissions trajectories and use them in a chemical transport model

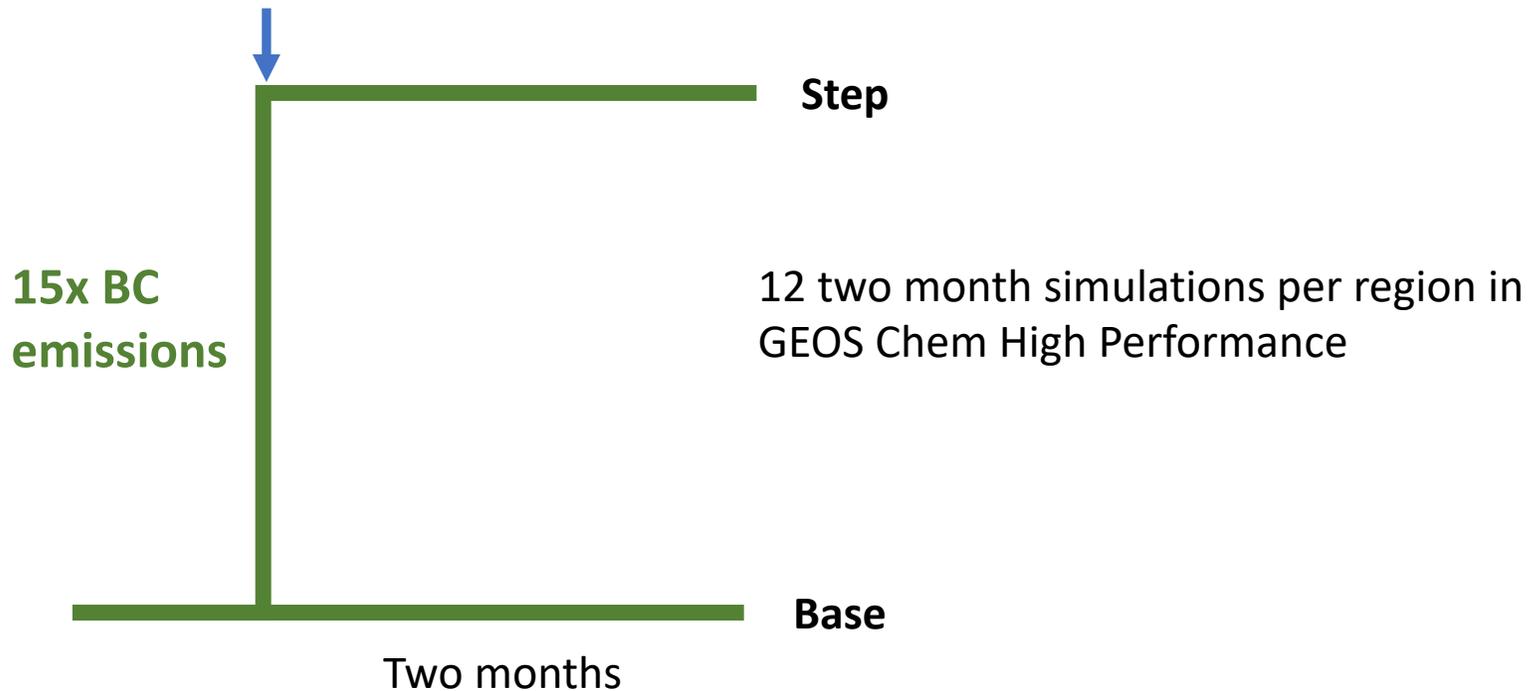
Missing: Spatial and Temporal variability

Years Later

Shared Socio-economic Pathways; Riahi, 2017

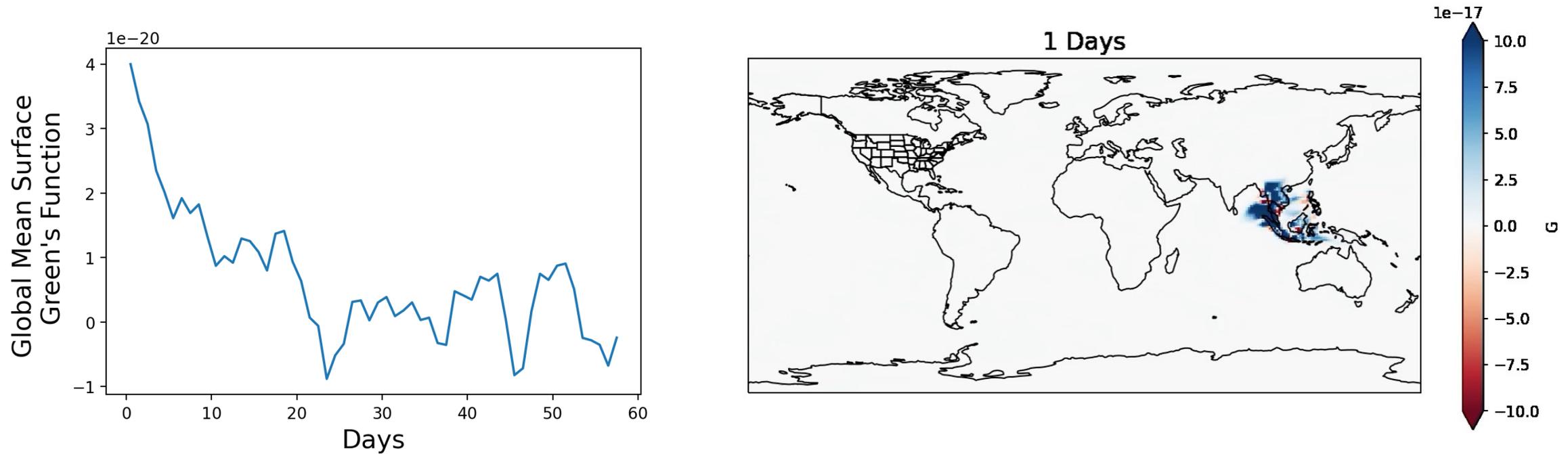
Using GEOS Chem High Performance to Diagnose the Green's Functions

- January, April, July, October
- 2000 and 2016



GCHP v13.3.4
CEDS Emissions Patterns
Resolution: C90 ($\sim 1^\circ \times 1^\circ$) – Also testing stretch grid and C180

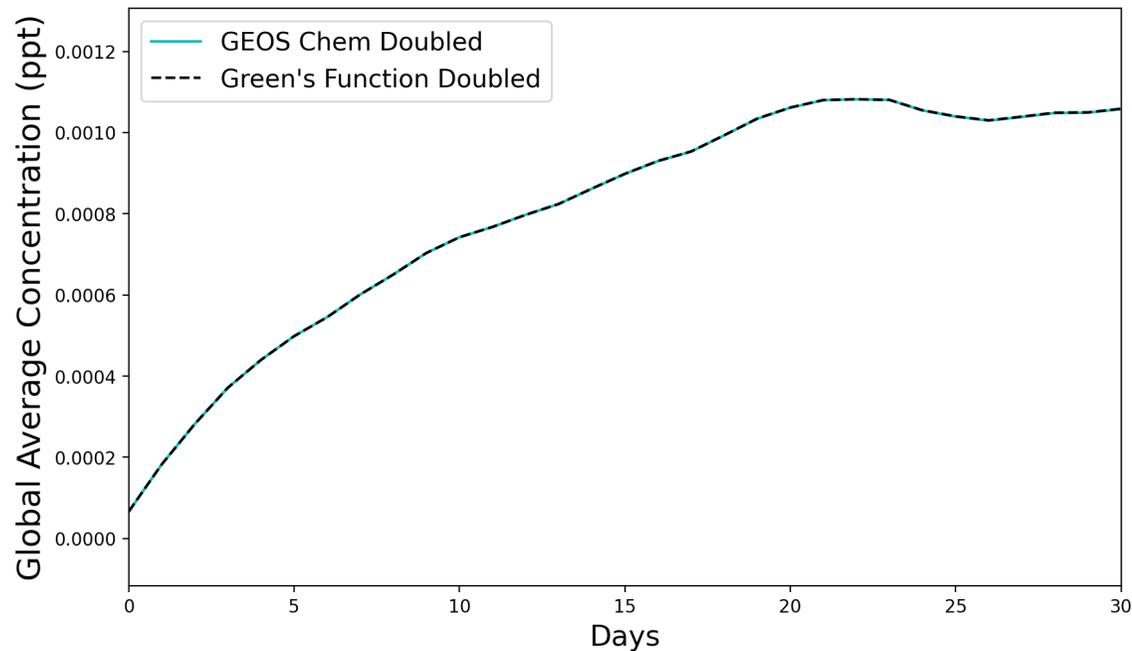
Black Carbon Green's Function over Southeast Asia (Jan 2016 run)



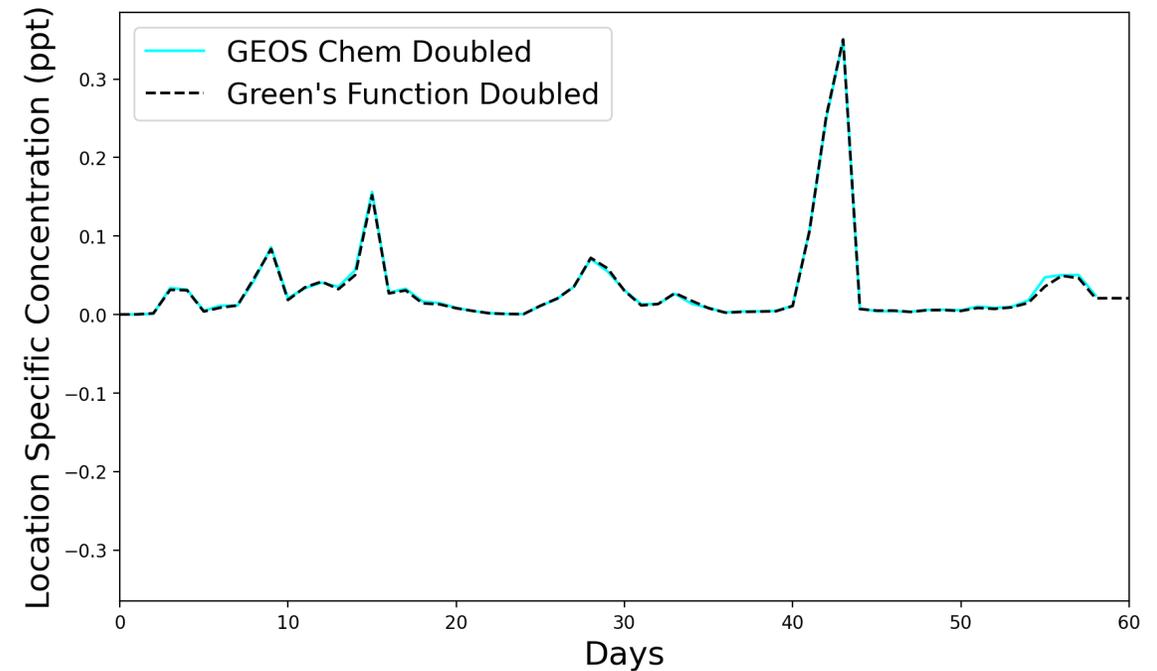
Validation of our Green's Function with a 2x SEA Black Carbon Simulation

Using Green's Function from one experiment to recreate a doubled GEOS Chem output

Global Average



One individual grid box (Guangzhou, China)

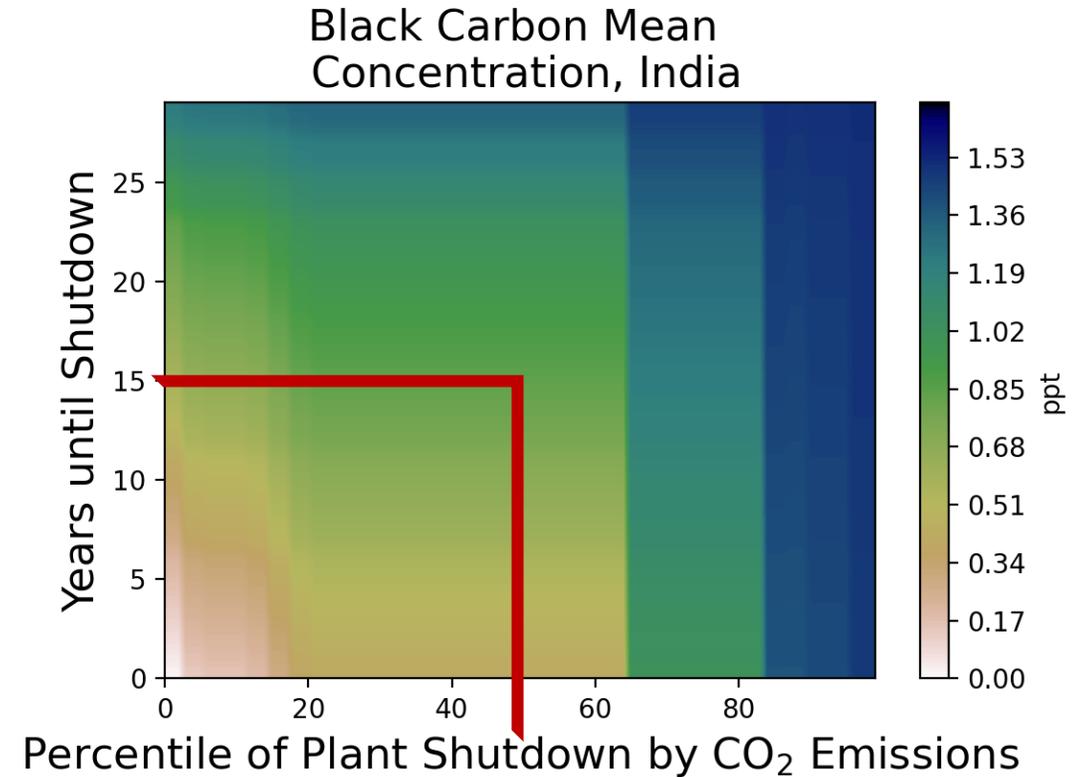
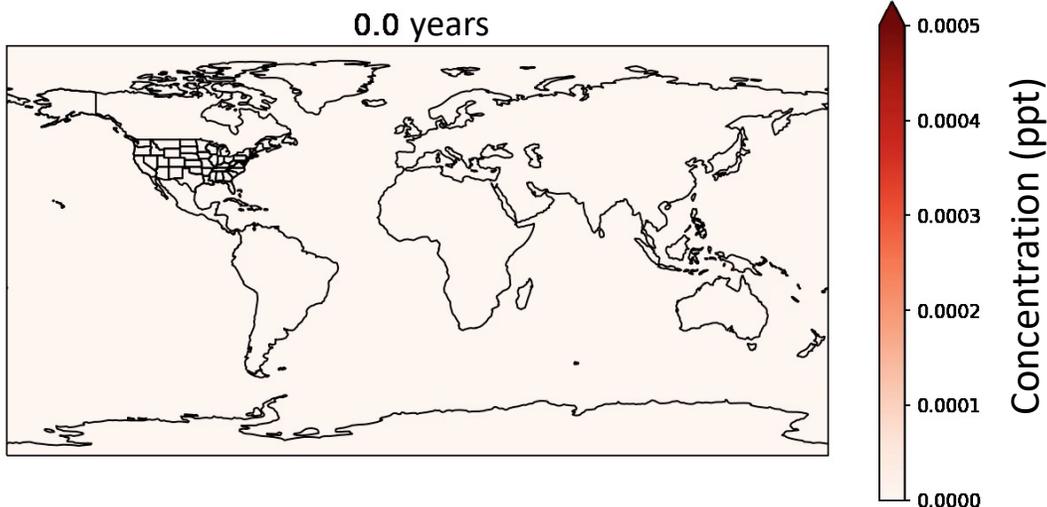


Early Retirement of Coal Power in Southeast Asia

Closures in Southeast Asia driven by three key reasons:

1. CO₂ emissions factors
2. Plant Age
3. Local air pollution policies

Early Retirement After 20 Years



This can and will be applied to other important black carbon sources, such as wildfires and residential coal burning

Today's takeaways

1. Emission and pollution trajectories over space and time determine cumulative impacts
2. We can use Green's functions for reduced form Black Carbon modeling with low computational load
3. Applying Green's functions across dozens of coal early retirement trajectories in Southeast Asia informs us of cumulative impacts, policies to prioritize, and transboundary pollutant transport
4. Future aspects: applying this to Hg and CO₂

Contact me:



emfreese@mit.edu



<https://lfreese.github.io/>



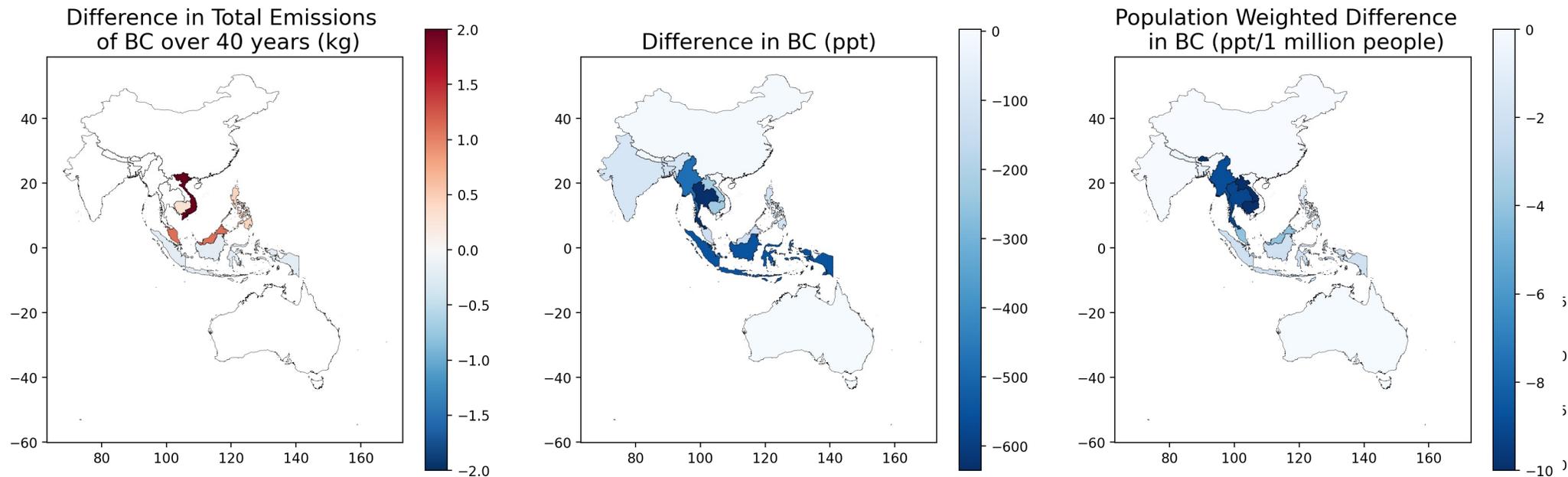
[@lyssafreese](https://twitter.com/lyssafreese)

Cumulative impacts over time— comparing across policies

Scenario 1: Shutdown all plants after 20 years of operation

Scenario 3: Shutdown top 50% of CO₂ emitters after 5 years, all other plants have 40-year lifetime

Scenario 3 < Scenario 1



This can and will be applied to other important black carbon sources, such as wildfires and residential coal burning

Green's Functions in Chemical Transport Modeling

$$\frac{\delta n}{\delta t} = \overset{\text{Transport}}{-\nabla \cdot (n\tilde{U})} + \overset{\text{Production}}{P(n)} - \overset{\text{Loss}}{L(n)} \quad \text{Continuity Equation}$$

$$E(s) = \frac{\delta n}{\delta t} + \nabla \cdot (n\tilde{U}(s)) - \underset{\substack{\text{Chemical} \\ \text{Production}}}{C(n, s)} + \underset{\text{Chemical Loss}}{L_C(n, s)} + \underset{\text{Deposition}}{L_D(n, s)}$$

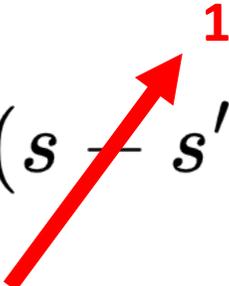
s is any combination of x, y, t

Green's Functions in Chemical Transport Modeling

$$E(s) = \mathcal{L}n \quad \text{Linear operator = continuity equation}$$

$$\mathcal{L}G(s - s') = \delta(s - s') \quad \text{Definition of a Green's Function}$$

Delta Function

$$\int \mathcal{L}G(s - s')E(s')ds = \int \delta(s - s')E(s')ds$$


$$\mathcal{L} \int G(s - s')E(s')ds = E(s)$$

$$\int G(s - s')E(s')ds = n \quad \text{Taking the relationship between the linear operator acting on } n \text{ and the Emissions}$$

Green's Functions in Chemical Transport Modeling

$$E(s) = \mathcal{L}n \quad \text{Linear operator = continuity equation}$$

$$\mathcal{L}G(s - s') = \delta(s - s') \quad \text{Definition of a Green's Function}$$

Delta Function

If we have G, we now know we can convolve that with Emissions to find our concentration

$$\int \mathcal{L}G(s - s')E(s')ds$$

$$\mathcal{L} \int G(s - s')E(s')ds = E(s)$$

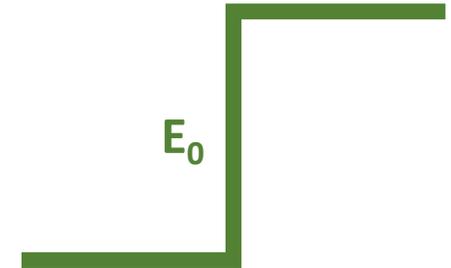
$$\int G(s - s')E(s')ds = n$$

Taking the relationship between the linear operator acting on n and the Emissions

How do we get G from a step increase in emissions?

$$\mathcal{L}c = H(t)E_0$$

Continuity equation acting on a specific pollutant in a given location



$$\frac{\delta}{\delta t} \mathcal{L}c = \frac{\delta}{\delta t} H(t)E_0$$

Time derivative of the Heaviside function = delta function

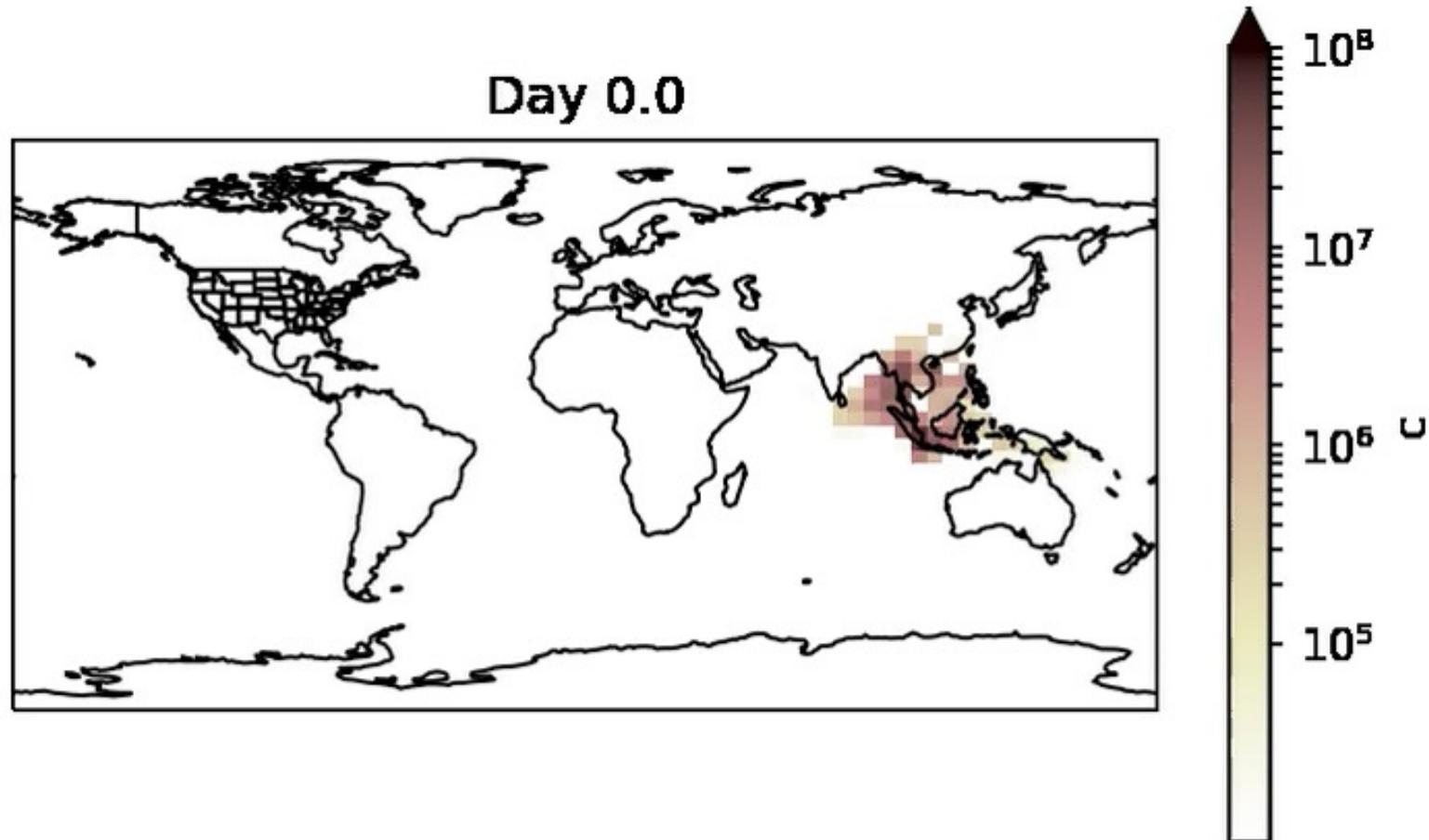
$$\mathcal{L} \frac{\delta c}{\delta t} = \delta(t)E_0$$

$$\mathcal{L} \left(\frac{\delta c}{\delta t} \frac{1}{E_0} \right) = \delta(t)$$

Definition of a Green's Function

$$\mathcal{L}G(s - s') = \delta(s - s')$$

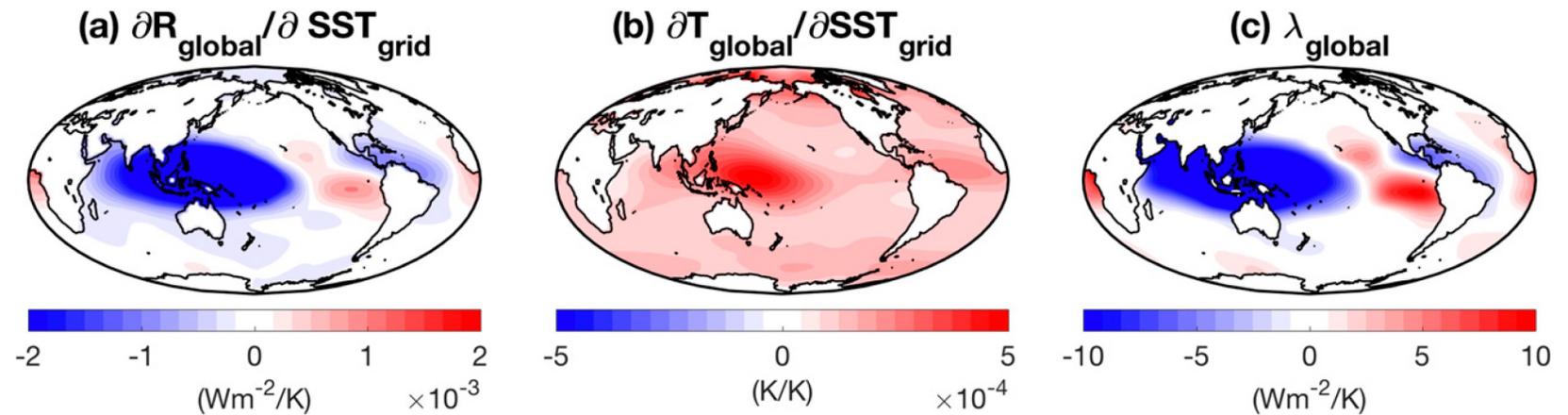
Green's Functions in Chemical Transport Modeling



Application of Green's Functions in Sea Surface Temperature Perturbations and Climate Science

Radiative feedback,
 λ in response to sea
surface
temperature

$$\lambda(t) = \frac{\overline{\mathbf{R}}}{\overline{\mathbf{T}}},$$



Criteria for Applying Green's Functions to Atmospheric Chemistry

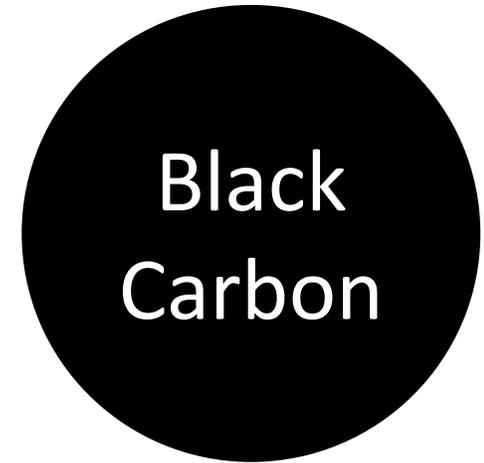
Similar to previous work, we need a variable that:

1. Changes as a function of time
2. Is linearizable
3. Is a direct output or can be indirectly calculated from outputs of a complex model

And is relevant to our scientific research questions:

1. Emitted from coal power
2. Has climate/air quality impacts

Criteria for Applying Green's Functions to Atmospheric Chemistry



Similar to previous work, we need a variable that:

1. Changes as a function of time
2. Is linearizable
3. Is a direct output or can be indirectly calculated from outputs of a complex model

And is relevant to our scientific research questions:

1. Emitted from coal power
2. Has climate/air quality impacts

Using a step increase to calculate the Green's Function

1. Two Chemical Transport model simulations are needed:
 1. A base scenario where emissions are normal
 2. A step increase scenario where emissions are increased by x amount in the region of interest
2. Normalize this by the emissions forcing

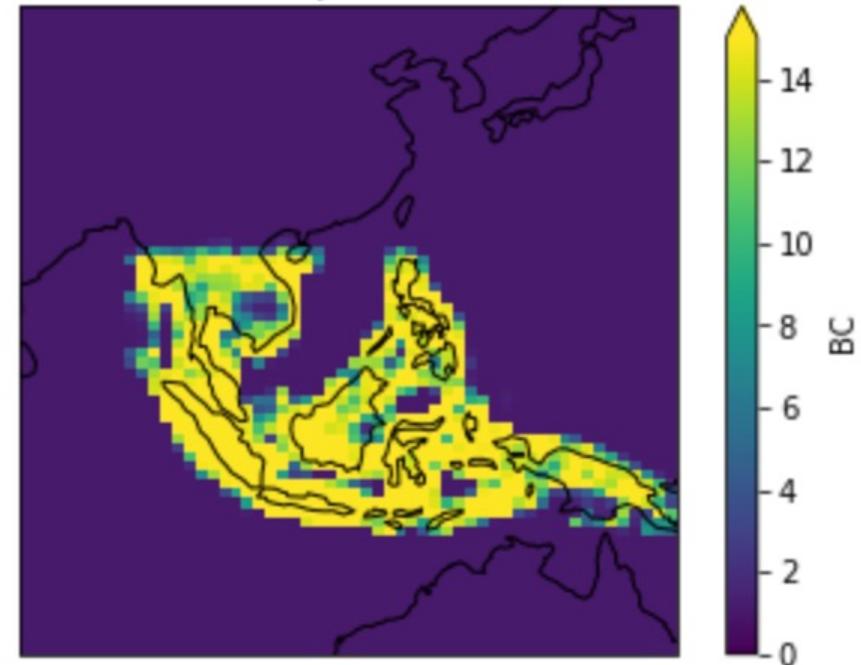
The outcome of this is our Green's Function, which we can convolve with an emissions profile to estimate the spread of a pollutant from the region of interest

$$G(t) = \frac{\left(\frac{d C_{step} - C_{base}}{dt} \right)}{(E_{step,t=0} - E_{base,t=0})}$$

1. Simulation details– Base and Step increase

- GEOS-Chem High Performance
- Black Carbon
- Two 60 day simulations
 - Base scenario uses BC emissions from CEDS
 - Step increase uses BC emissions from CEDS increased across all sectors in the region of interest by **15x**

Emissions Difference of our Step/Base

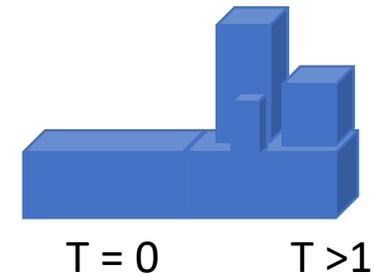


1.2: Three potential types of step increases:

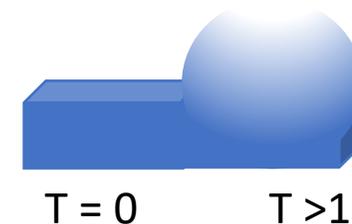
1. Pure step (one value added to all grid boxes)



2. Weighted by existing emissions (% of current emissions added to each grid box)

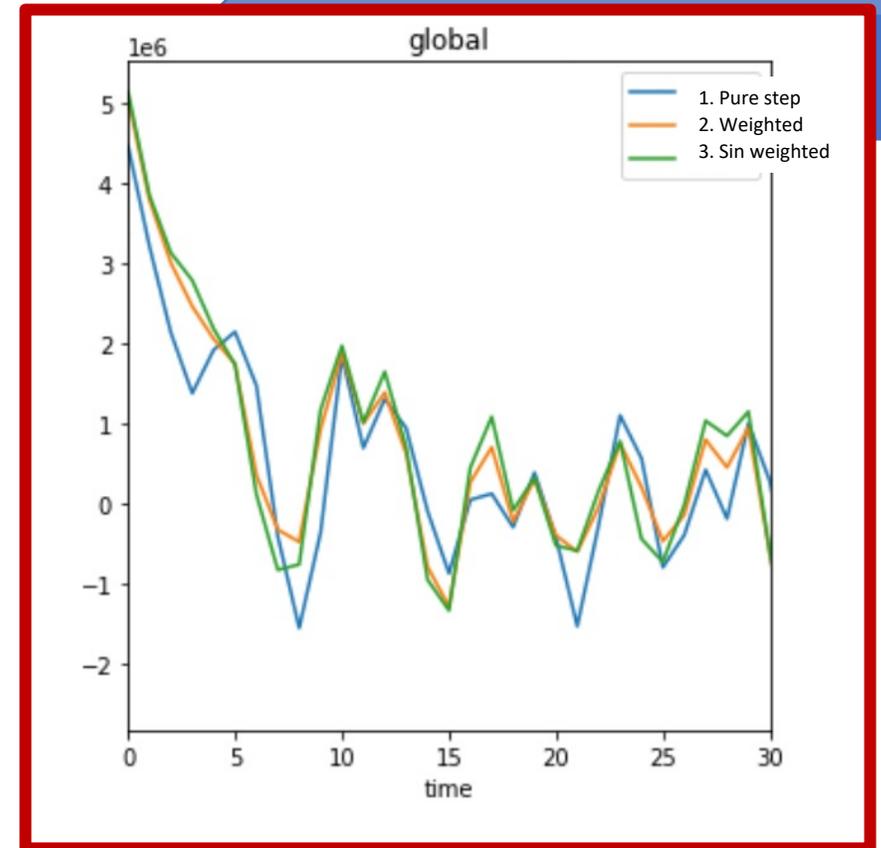


3. Sin weighted (reduces the impact of the boundaries on spread)



1.2: Three potential types of step increases:

1. Pure step (one value added to all grid boxes)
2. Weighted by existing emissions (% of current emissions added to each grid box)
3. Sin weighted (reduces the impact of the boundaries on spread)

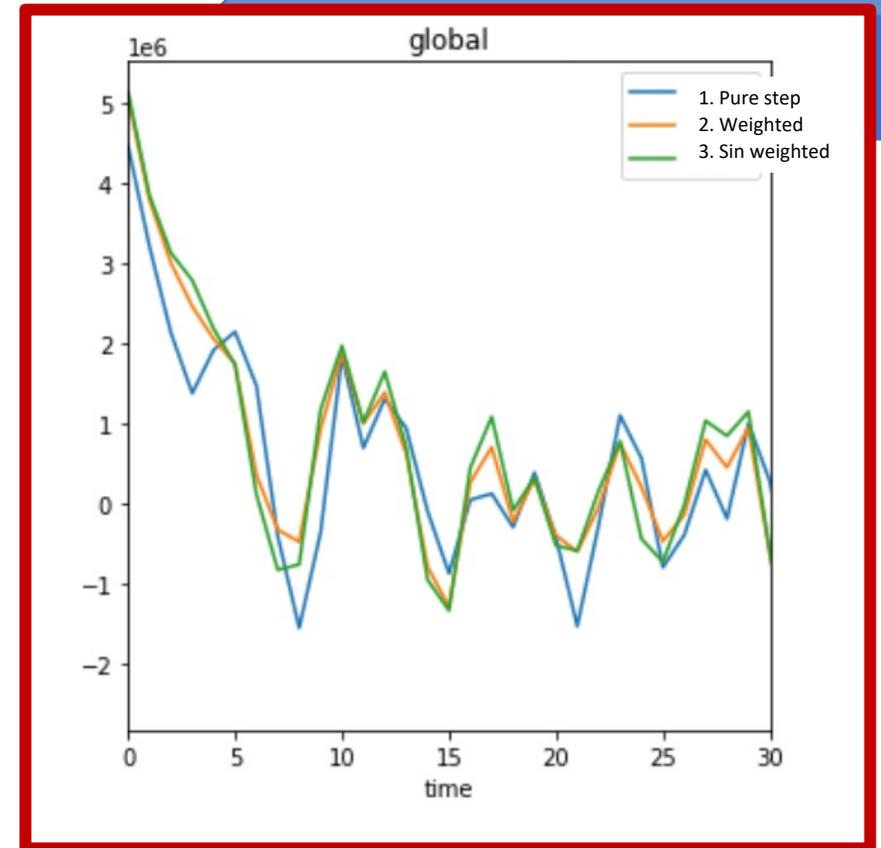


T = 0

T > 1

1.2: Three potential types of step increases:

1. Pure step (one value added to all grid boxes)
2. Weighted by existing emissions (% of current emissions added to each grid box)
3. Sin weighted (reduces the impact of the boundaries on spread)



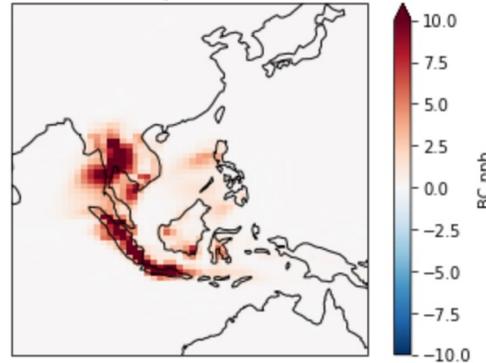
T = 0

T > 1

2. Normalizing our Green's Function by the emissions forcing

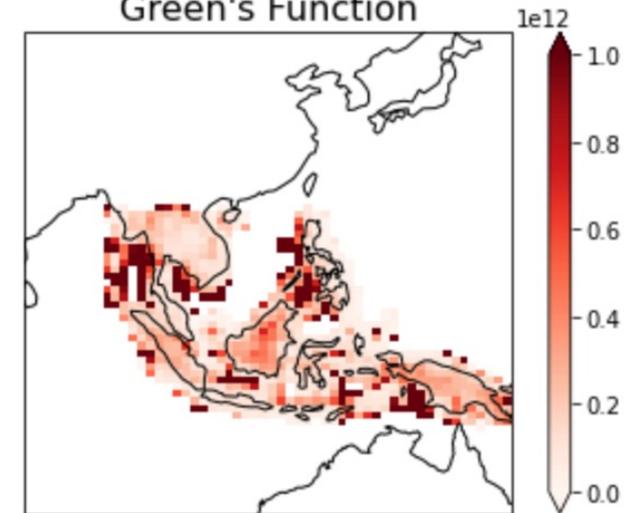
$$G(t) = \frac{\left(\frac{d C_{step} - C_{base}}{dt} \right)}{(E_{step,t=0} - E_{base,t=0})}$$

Time Derivative Concentration
Difference of our
Step - Base



E_0

Green's Function

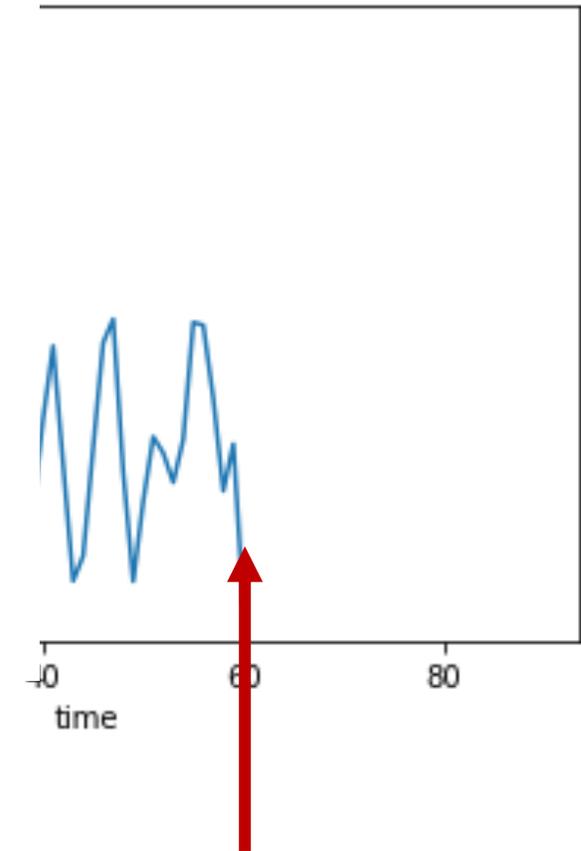
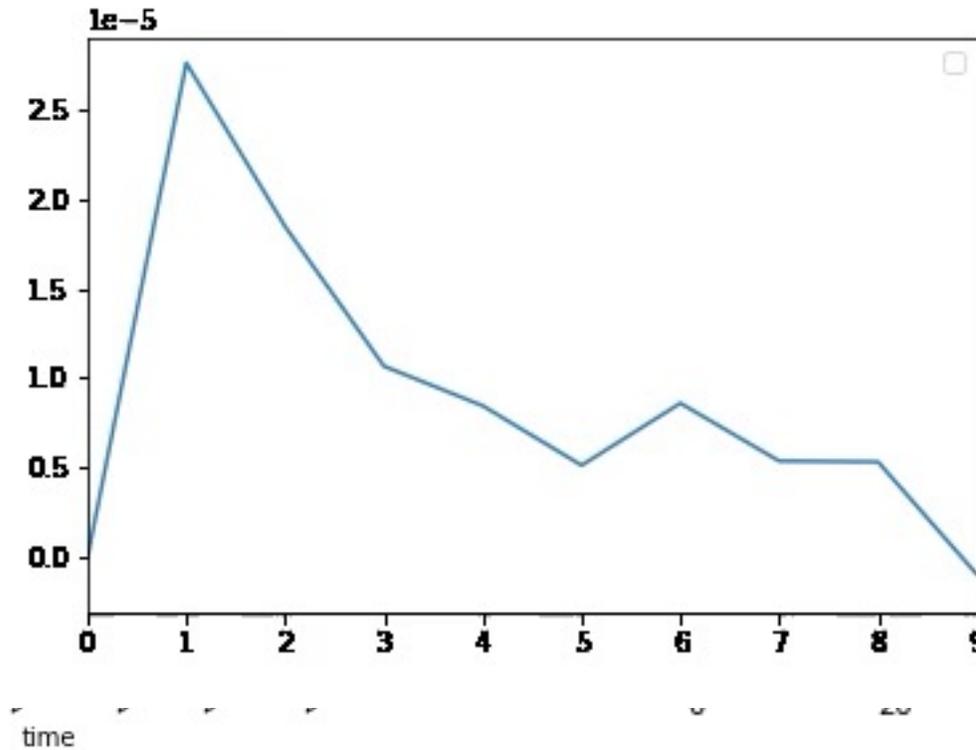
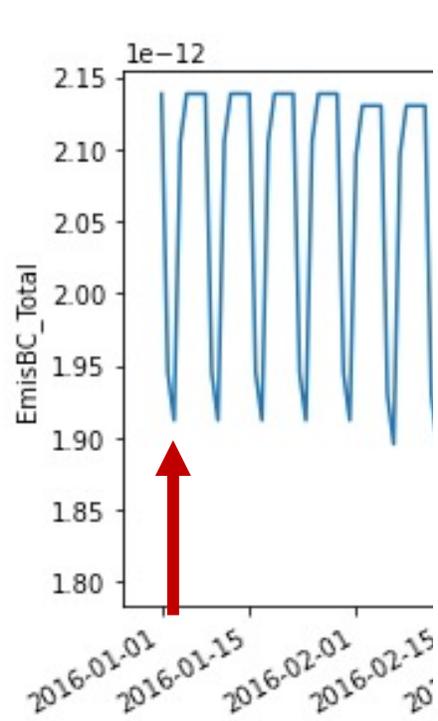


Convolution of our Green's Function with the Emissions Profile

1. Convolve the Emissions in each location with the Green's Function
2. Sum over time (and space if looking for a global response)

$$\Delta[X] = \int_{t_0}^{t_f} \int_{x_0}^{x_f} E(t') G_{(t-t', x-x')} dt' x' \approx \sum_{t_0}^{t_f} \sum_{x_0}^{x_f} E(t') G_{(t-t', x-x')} \Delta t$$

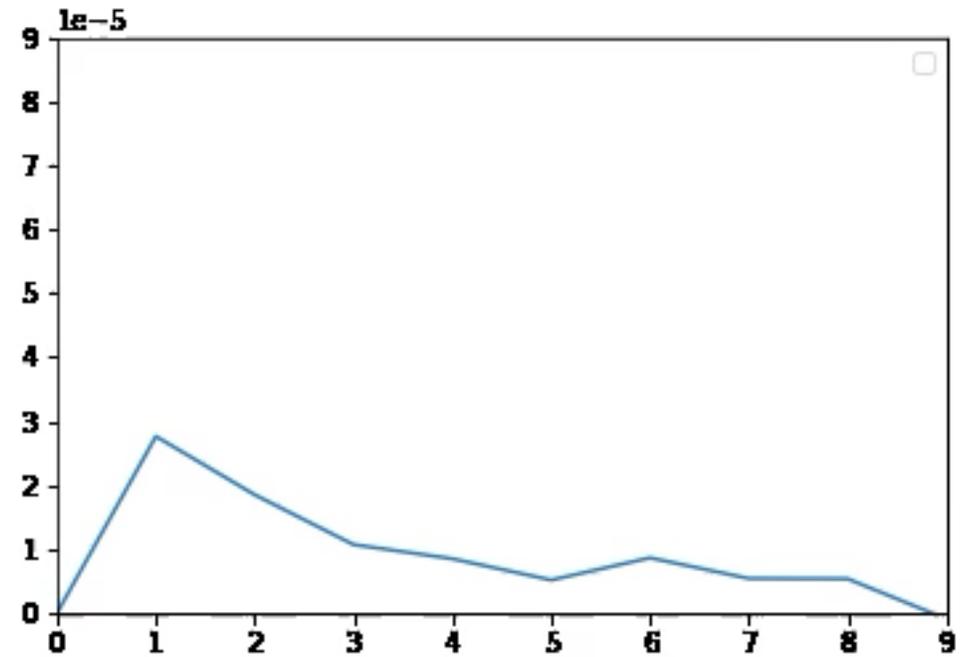
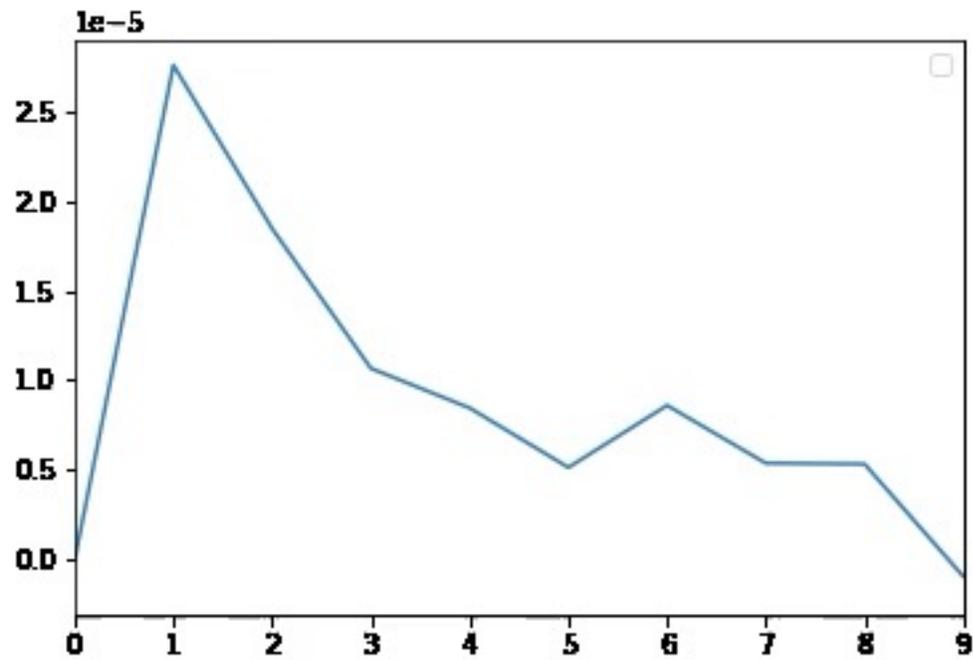
1. Convolution of Emissions



$$E(t')G(t - t')$$

After 60 days I am treating it as 0

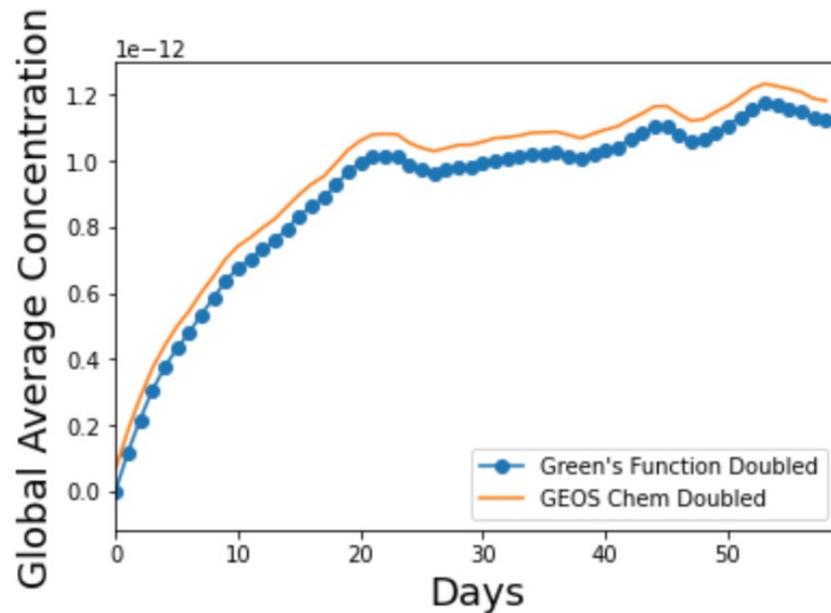
2. Sum over time



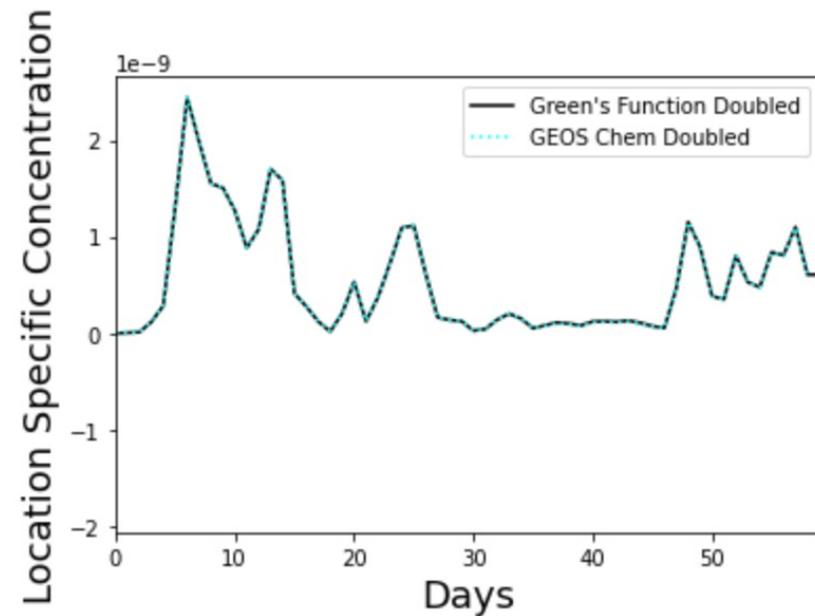
Validation of our Green's Function with a 2x Simulation

Use Green's Function from one experiment to recreate a doubled GEOS Chem output

Global Average



One individual grid box



Research Questions and Scenario Application

What are the impacts of trajectories of coal plant retirements in Southeast Asia on a black carbon?

Simulation Setup across Southeast Asia

Locations of step changes:

Southeast Asia

Indonesia

Singapore

Malaysia

Vietnam

Cambodia

Execution time: 15-30 seconds per every 10 years (in an interactive notebook)

In GEOS Chem High Performance it takes 48 days to run one year on 60 cores (50 years of simulations for one scenario would take longer than my PhD)

Times of step changes:

January

April

July

October

2016

2000

Coal Early Retirements due to four key reasons

1. Location based
2. CO₂ emission factors
3. Plant Age
4. Funding type

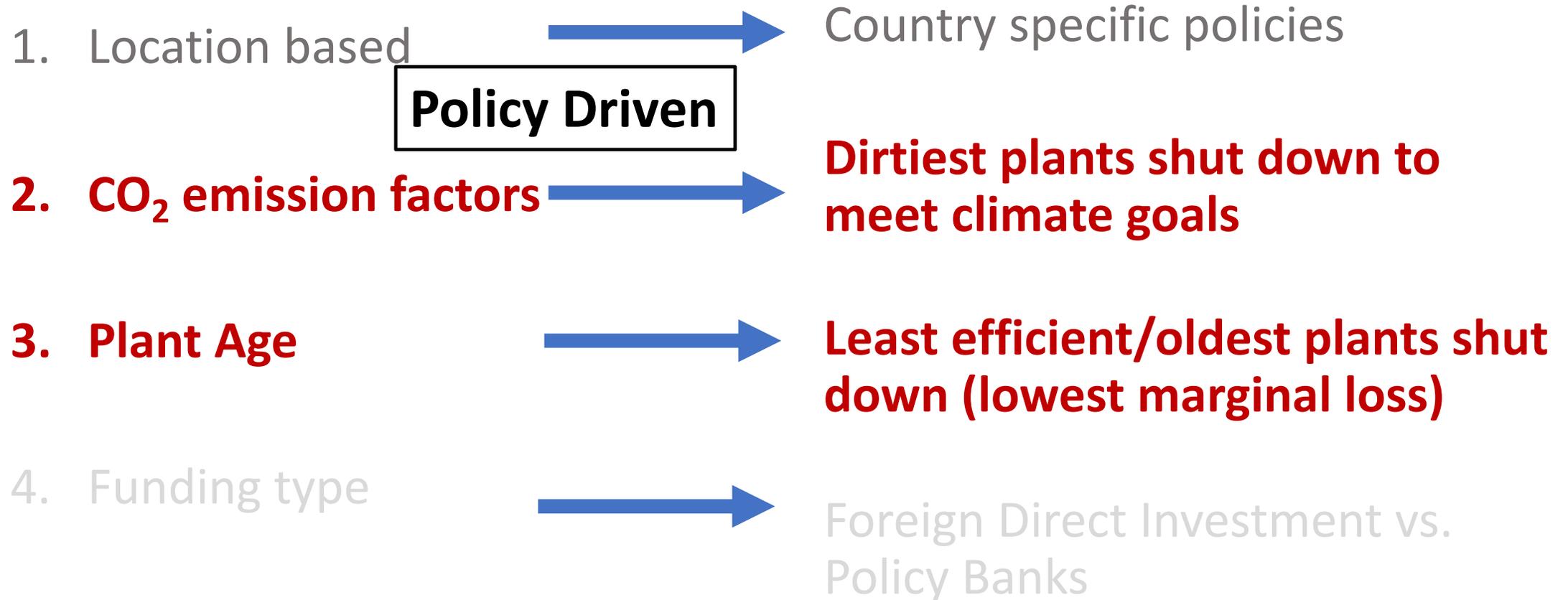
Coal Early Retirements due to four key reasons

1. Location based  Country specific policies
2. CO₂ emission factors  Dirtiest plants shut down to meet climate goals
3. Plant Age  Least efficient/oldest plants shut down (lowest marginal loss)
4. Funding type  Foreign Direct Investment vs. Policy Banks

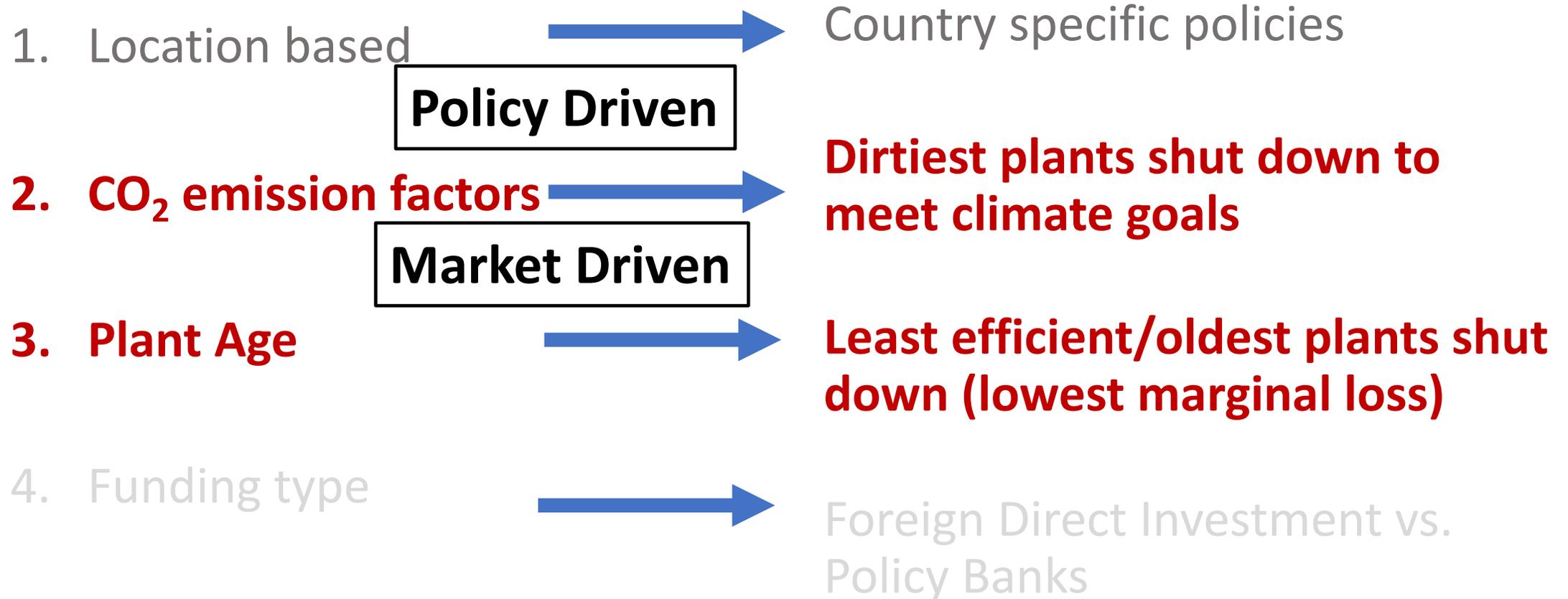
Coal Early Retirements due to four key reasons

1. Location based  Country specific policies
2. **CO₂ emission factors**  **Dirtiest plants shut down to meet climate goals**
3. **Plant Age**  **Least efficient/oldest plants shut down (lowest marginal loss)**
4. Funding type  Foreign Direct Investment vs. Policy Banks

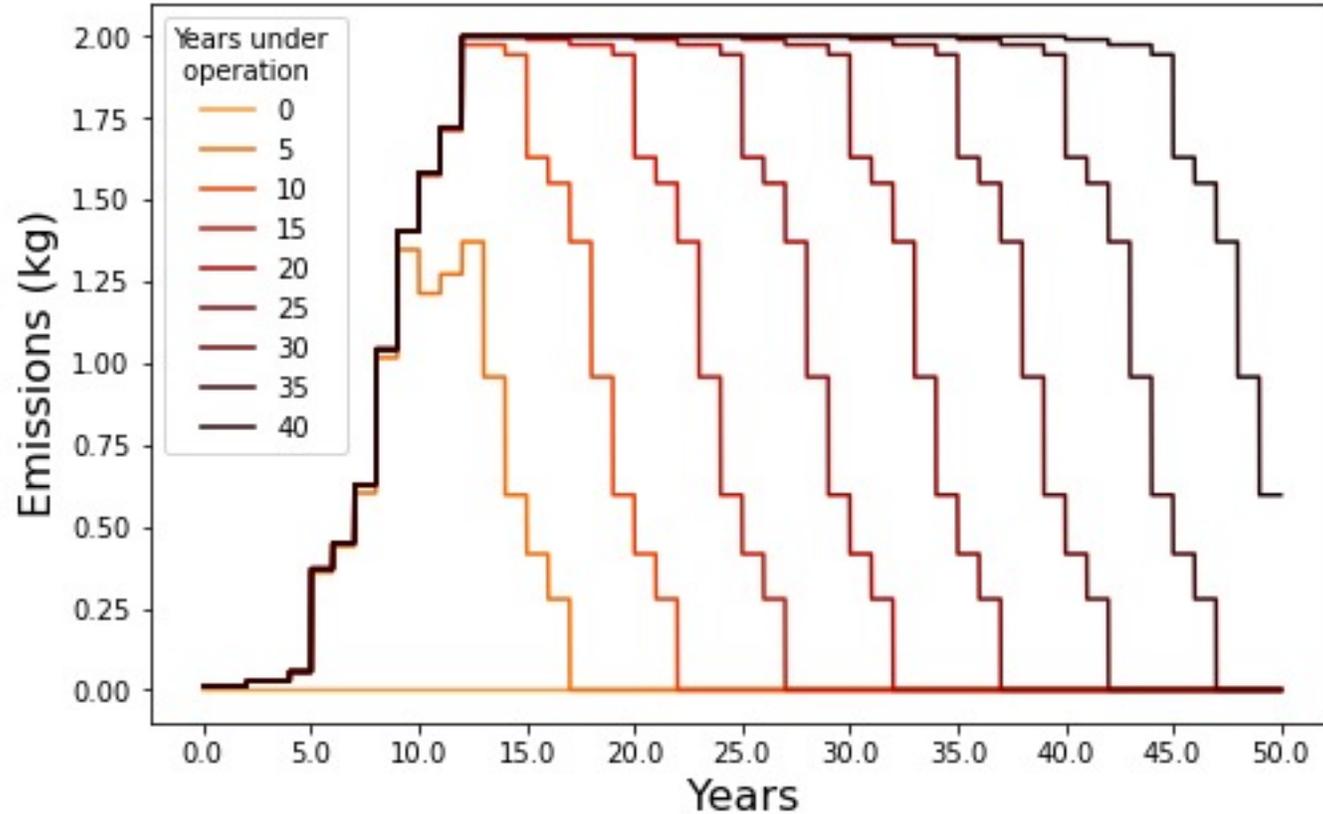
Coal Early Retirements due to four key reasons



Coal Early Retirements due to four key reasons



Early Retirement based on plant age: Trajectories



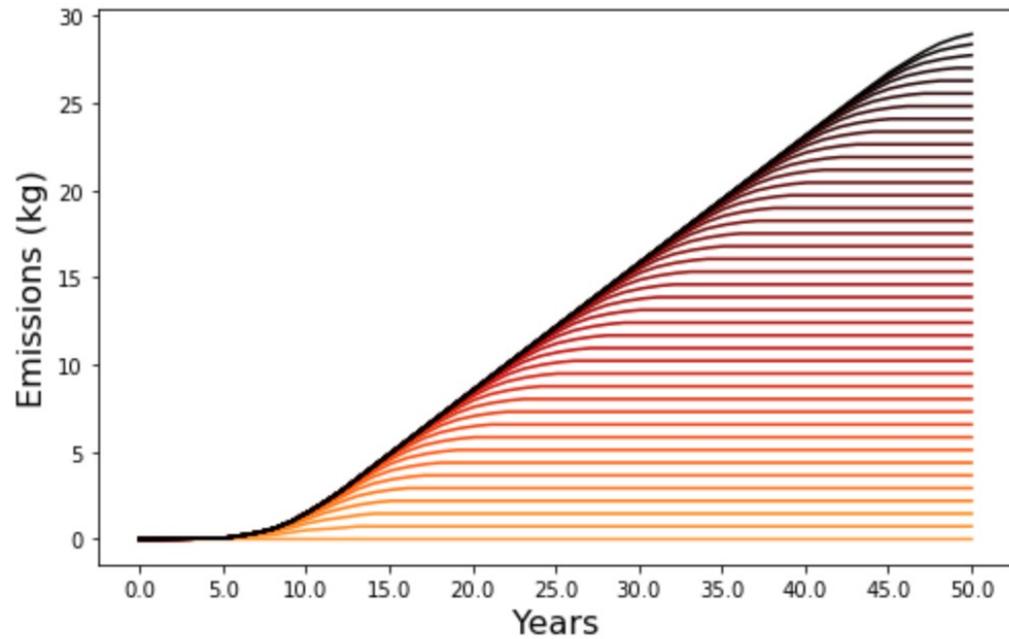
Emissions of black carbon from all Chinese funded Southeast Asian coal plants

Assuming emission factor is constant at .026 g/kg of coal burned, assuming a 22.51 GJ/t conversion factor

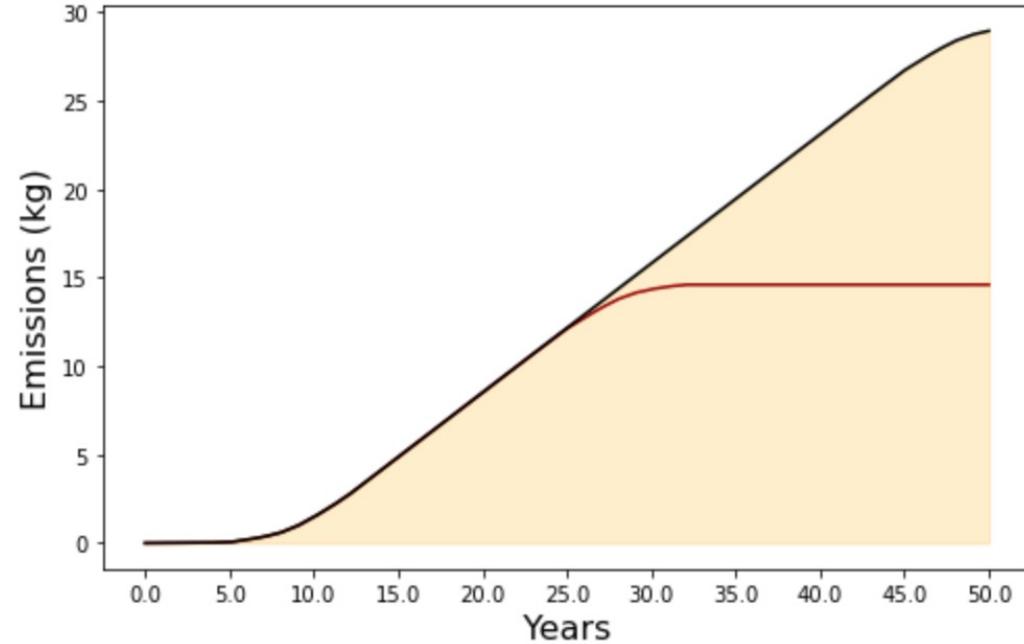
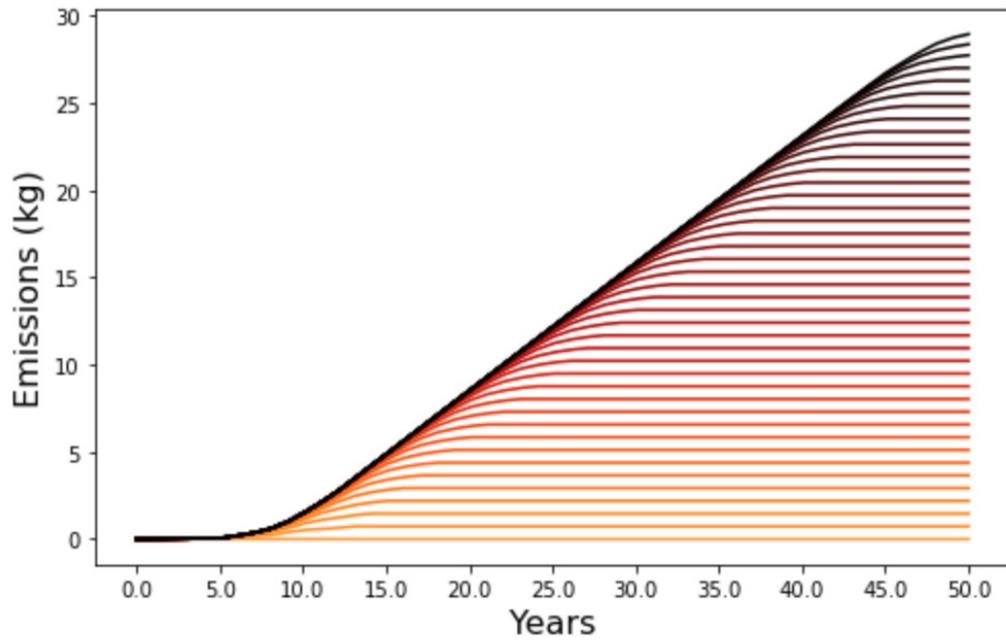
Wang et al, 2012 and National Research Council

National Academies of Sciences, Engineering, and Medicine. 2000. Cooperation in the Energy Futures of China and the United States. Washington, DC: The National Academies Press. <https://doi.org/10.17226/9736>.

Early Retirement based on plant age: Cumulative Emissions



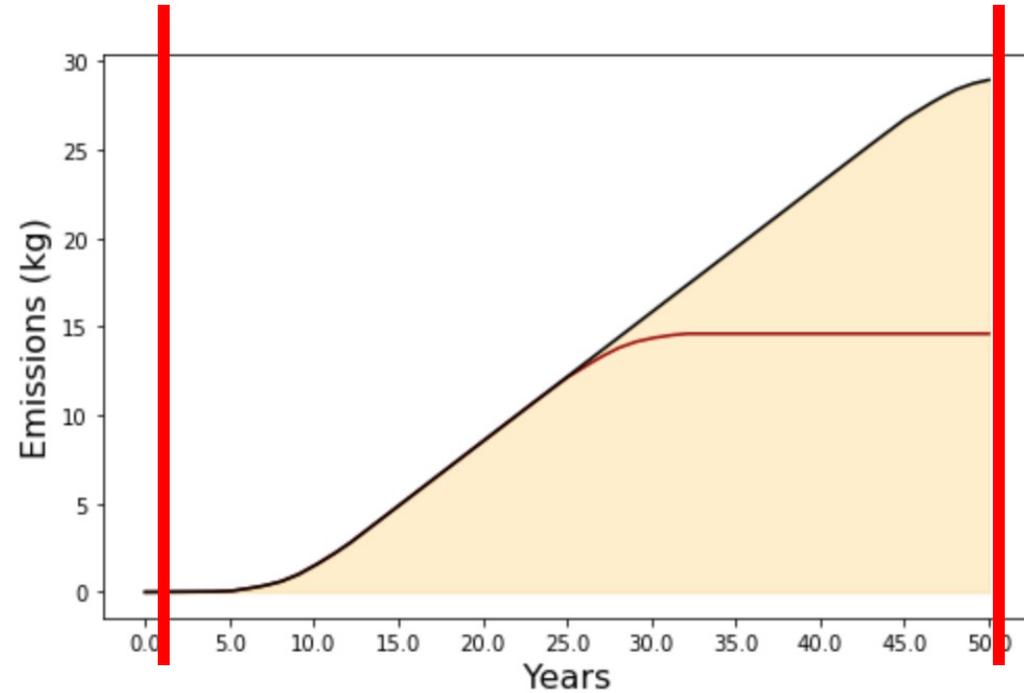
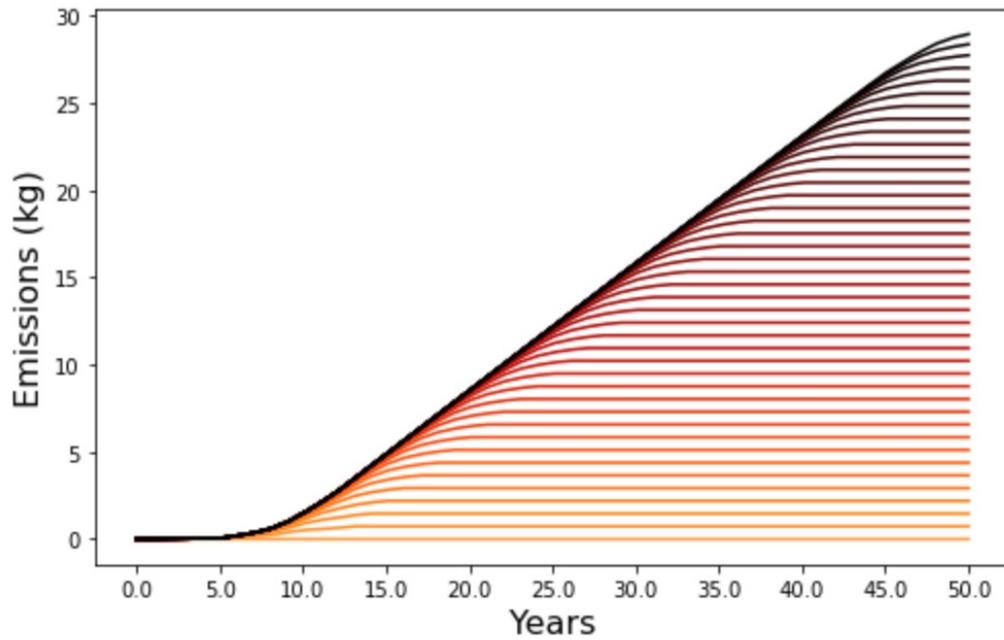
Early Retirement based on plant age: Cumulative Emissions



Base
Scenario

Policy
Scenario

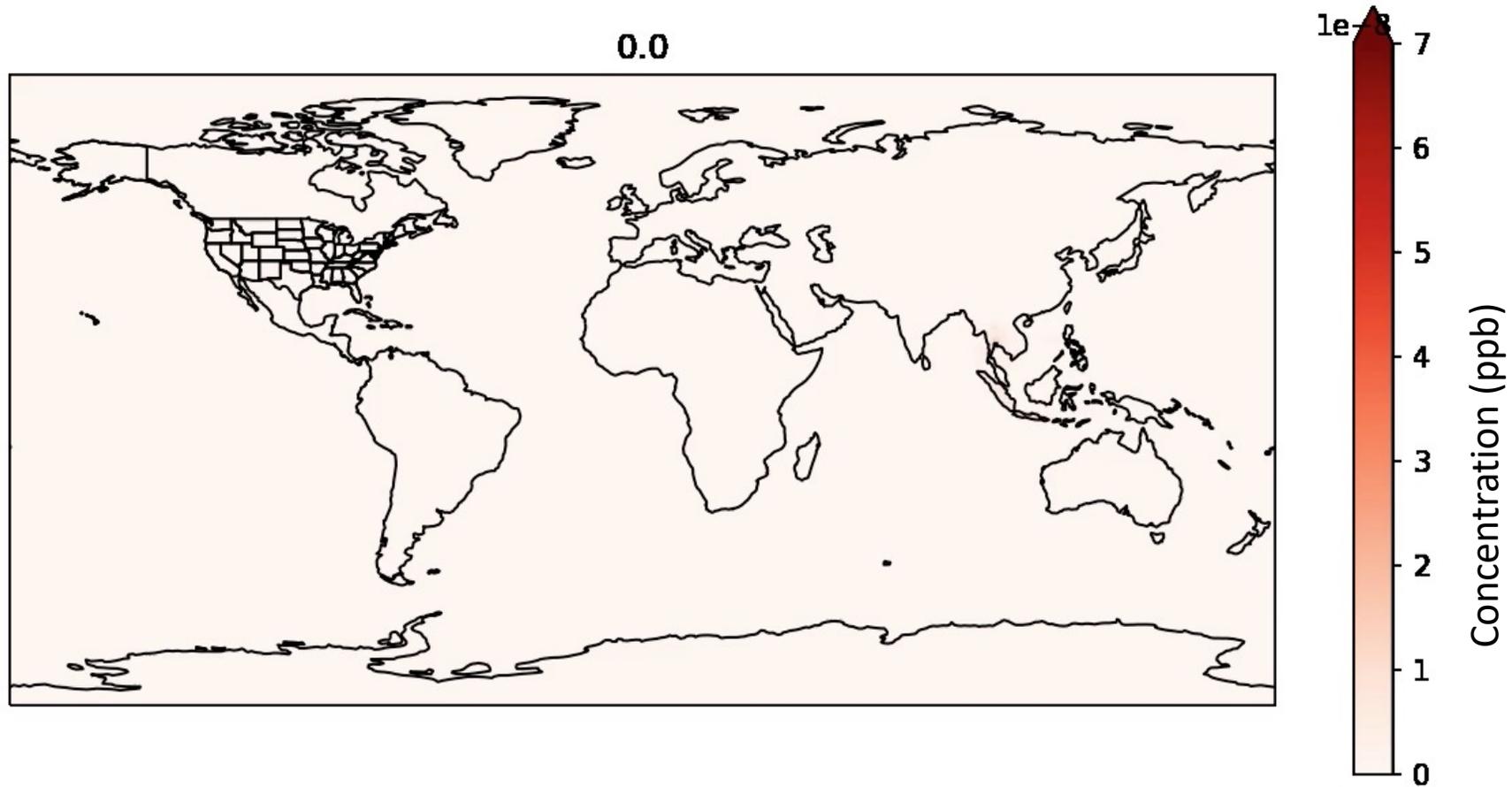
Early Retirement based on plant age: Cumulative Emissions



Base
Scenario

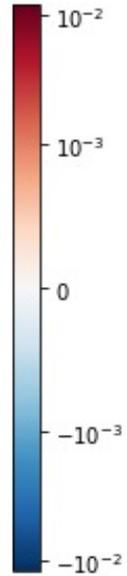
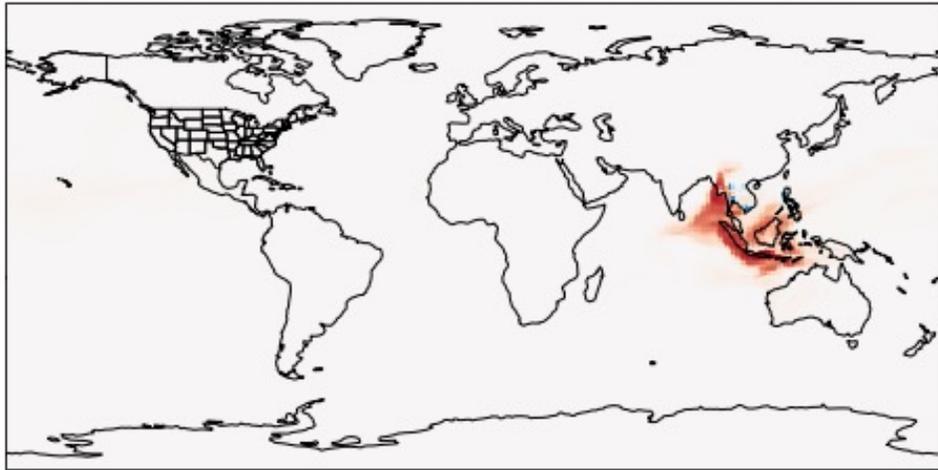
Policy
Scenario

Closing plants after 15 years of operation

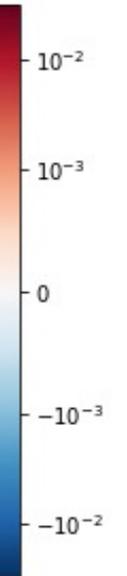
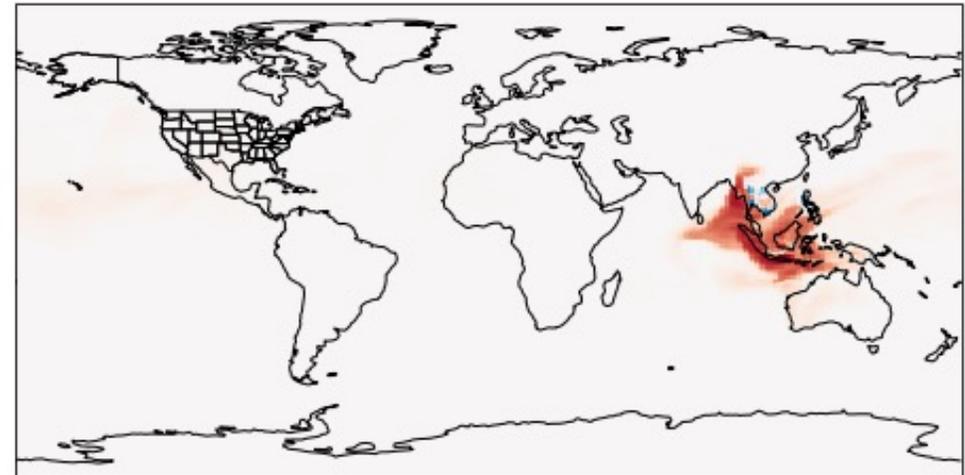


Early Retirement Cumulative Differences

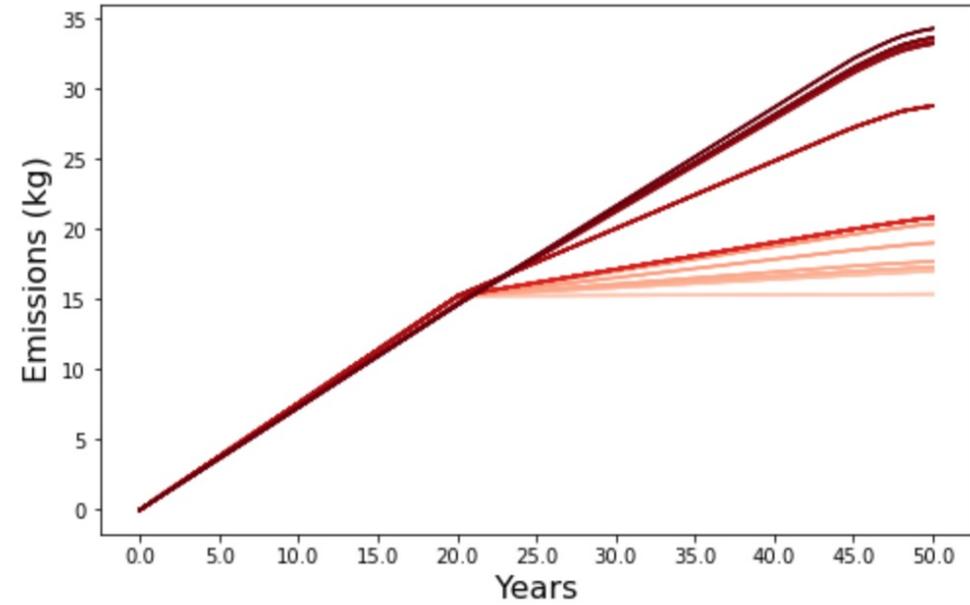
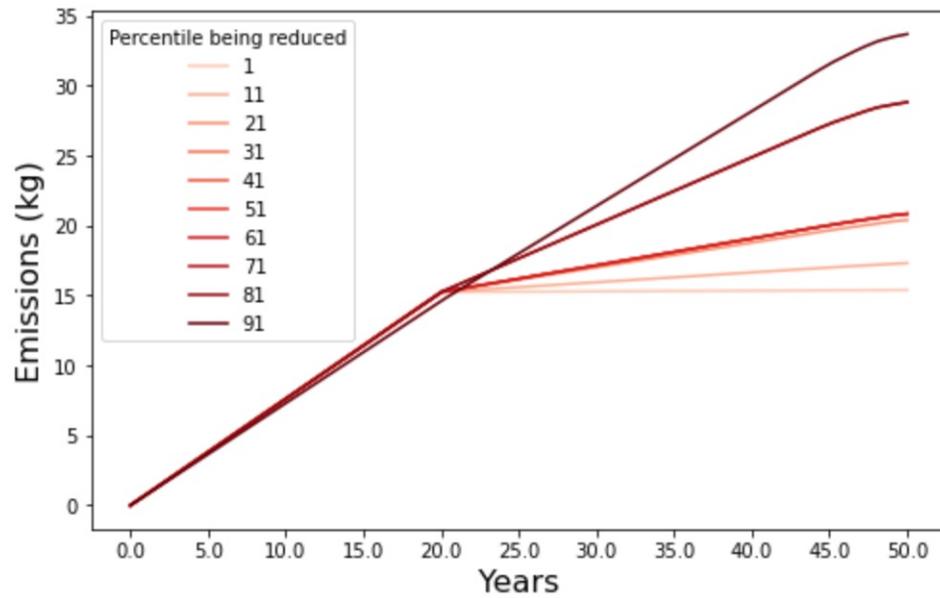
Early retirement after 30 - 15 years Cumulative BC difference



Early retirement after 40 - 15 years Cumulative BC difference



Shutting down high emitters by percentile after 20 years



Shutting down high emitters vs. early retirement

Shutting down highest 50% of emitters after 20 years
vs. early retirement after 20 years

