Understanding the Sources of Climate Uncertainty in Projections of Climate Impacts

Kevin Schwarzwald and Nathan Lenssen

COLUMBIA CLIMATE SCHOOL LAMONT-DOHERTY EARTH OBSERVATORY



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Studies that:

1. Take historical weather data



(Population-weighted continental USA historical temperature distribution)

Studies that:

2. Find a relationship between that data and some societal variable of interest (a "dose-response function")



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Studies that:

3. Project that relationship in the future using climate projections



 Δ Mortality = +3% under climate change due to heat, in continental USA

Exist for many variables; used e.g. to estimate Social Cost of Carbon





Looking at the three criteria from before...

- 1. Take historical weather data
- 2. Find a relationship between that data and some societal variable of interest
- 3. Project that relationship in the future using climate projections

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Sources of climate uncertainty in projections (e.g. Hawkins & Sutton 2009)

- Scenario uncertainty
- Model uncertainty
- Internal uncertainty

- Scenario uncertainty outcomes differ due to differing plausible assumptions about the future
- Model uncertainty
- Internal uncertainty



Sources of climate uncertainty in projections (e.g. Hawkins & Sutton 2009)

- Scenario uncertainty
- Model uncertainty outcomes differ due to different model responses to the same inputs
- Internal uncertainty



Sources of climate uncertainty in projections (e.g. Hawkins & Sutton 2009)

- Scenario uncertainty
- Model uncertainty
- Internal uncertainty outcomes differ due to internal variability of the climate system / models





Model uncertainty

 Internal uncertainty

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	0-)

Global, decadal mean surface air temperature



(from Hawkins and Sutton 2009)



Model uncertainty

 Internal uncertainty

Relative importance likely different for impacts studies due to nonlinearities vs. climate!

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Global, decadal mean surface air temperature



(from Hawkins and Sutton 2009)

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Impact shown for a set of dose-response functions using CMIP3 models by Burke et al. (2014); many impacts studies now use an ensemble of models / scenarios

- Scenario uncertainty
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- Internal uncertainty



Impact shown for a set of dose-response functions using CMIP3 models by Burke et al. (2014); many impacts studies now use an ensemble of models / scenarios

Effect on impacts projections has yet to be explicitly quantified. Now feasible using Single Model Initial-condition Large Ensembles (LEs - Deser et al. 2020)

Single Model Initial-Condition Large Ensembles (LEs)

Running the same climate model, with the same inputs (scenario), multiple times to sample internal variability

From Deser, Clara, Adam S. Phillips, Michael A. Alexander, and Brian V. Smoliak. 2014. "Projecting North American Climate over the Next 50 Years: Uncertainty Due to Internal Variability." Journal of Climate 27 (6): 2271–96. https:// doi.org/10.1175/JCLI-D-13-00451.1.



FIG. 1. Winter SAT trends $[2010-60; °C (51 \text{ yr})^{-1}]$ from each of the 40 CCSM3 ensemble members.

Plan

dose-response functions to all three sources of climate uncertainty

Does internal variability matter, and if so, under what conditions?

A sensitivity analysis of several well-known

Outline

- 1. Choice of dose-response functions
- 2. Partitioning climate uncertainty in projections using these dose-response functions
- 3. How to fully incorporate climate uncertainty into climate impacts projections

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1. Choice of dose-response functions

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Dose-response functions





Dose-response functions





outcome = f(T)

estimated over continental USA

40



Outline

- 1. Choice of dose-response functions
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2010-2039













Uncertainty partitioning depends on shape of dose-response function

Mortality



per capita

Corn yields



Uncertainty partitioning depends on shape of dose-response function

Mortality



per capita

Corn yields



Uncertainty partitioning depends on shape of doseresponse function



Geography of uncertainty

Early 21st. century 2010-2039

























Late 21st. century 2070-2099



Interaction between geography of climate variability and human variables (population, corn production,...) matters!

Relative contribution

to uncertainty

Scenario

<0.0 €. €. €. €. €.

0.33 0.67

0.

Model 👌

00.001

0.67 0.33 00.00

0

Internal



Outline

- 1. Choice of dose-response functions
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Sources of uncertainty

Relative importance of source of uncertainty (scenario, model, internal) depends on:

- Timeframe of interest
- Shape of dose-response function
- Geography

Sources of uncertainty

Ignoring internal variability can result in underestimating worst-case impacts!

- Non-linear relationships

- Long-term impacts projections

- Scenario uncertainty

Internal

variability



• Near-term (decision-relevant?) impacts projections • Areas with known existing strong internal variability

 Monotonic dose-response functions • Areas with few existing patterns of internal variability

Thank you!







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RESEARCH ARTICLE

ENVIRONMENTAL SCIENCES

The importance of internal climate variability in climate impact projections

Kevin Schwarzwald^{a,1} and Nathan Lenssen^{a,b}

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@ks905383

xagg - raster data to polygons for python





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Extra Slides

Is climate uncertainty important?

Climate uncertainty >> dose-response **uncertainty** in projections of many climate damages...

Figure from Hsiang, Solomon, Robert Kopp, Amir Jina, James Rising, Michael Delgado, Shashank Mohan, D. J. Rasmussen, et al. 2017. "Estimating Economic Damage from Climate Change in the United States." Science 356 (6345): 1362–69. https://doi.org/10.1126/science.aal4369.



Supplementary Figure S7: Decomposing sources of uncertainty for damages over time. Same as Figure S6 but for multiple time periods and normalizing variances to by the sum of variances from climate models, weather, and econometric uncertainty. Approach is same as ref. (43). Green line indicates the total variance once the "interaction" component Λ is accounted for. Λ can be read as the difference between 1 and the green line at each moment, it is possible for $\Lambda < 0$.









Single Model Initial-Condition Large Ensembles (LEs)

Running the same climate model, with the same inputs (scenario), multiple times to sample internal variability

Initialization either:

- "Micro": bit-flipping initial conditions
- "Macro": branching off from a long control run

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Internal, model, scenario uncertainty

Projection / bias-correction method

FIG. 1. Winter SAT trends $[2010-60; ^{\circ}C (51 \text{ yr})^{-1}]$ from each of the 40 CCSM3 ensemble members.

What data are useful in impacts contexts?



Single Model Initial-Condition Large Ensembles (LEs)

Some limitations (of this generation):

- CMIP5 generation
- Some only have monthly data
- One scenario (RCP8.5)

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What data are useful in impacts contexts?



GFDL-CM3's odd changing seasonal cycle

The outlier in mortality projections due to dramatically increasing summertime temperatures...



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2010-2039 2040-2069 2070-2099