



# American Climate Prospectus

Economic Risks in the United States

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Stanford University

**James Rising**  
Columbia University

**Shashank Mohan**  
Rhodium Group

**Climate Change Impacts/Integrated Assessment | Snowmass, CO | July 22, 2014**  
(slides updated August 16, 2014)

# Overview

# An Independent Assessment for a Climate Risk Committee

Analytical Support for the Risky Business Project ([riskybusiness.org](http://riskybusiness.org))



**“If you can’t measure it, you can’t manage it” ...**

**- Mike Bloomberg**

# Research Objectives

1. Where possible, micro-found the damage function for the United States using the highest-quality, identified, empirical measurements
2. Make the calculation transparent (and hopefully open source)
3. Make updating the calculation easy
4. Leverage state-of-the-art physical climate models to describe the spatial structure of impacts
5. Quantify uncertainty and risk
6. Make the results interpretable and relatable to ordinary citizens
7. Finish in 12 months

# Scope of coverage

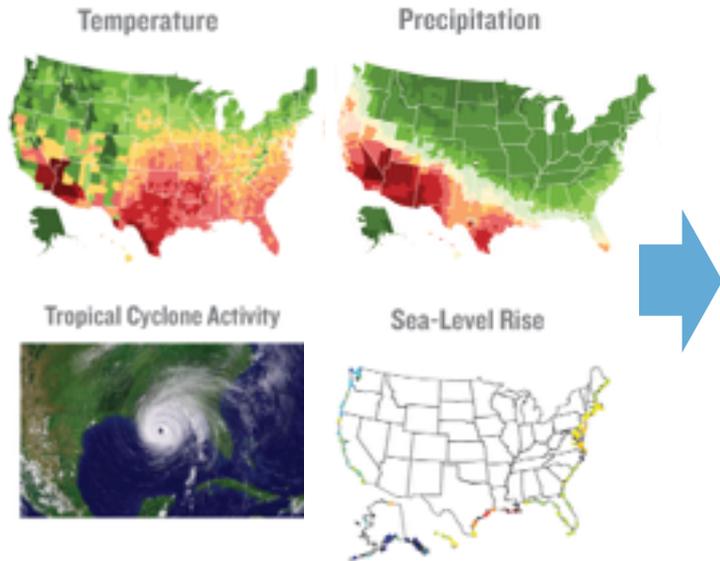
Far from comprehensive



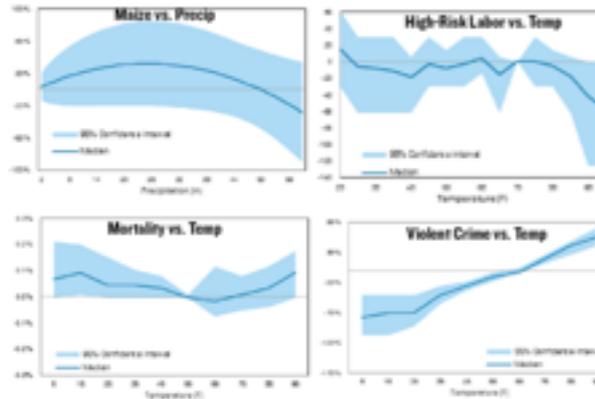
# Research approach

Spatial Empirical Adaptive Global-to-Local Assessment System (SEAGLAS)

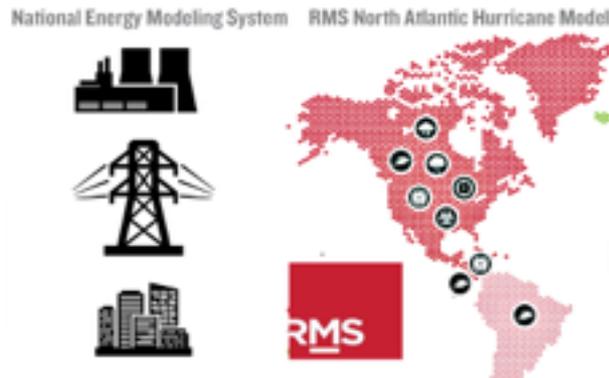
**Downscaled, probabilistic  
Physical Climate Projections**



**Impact estimates based on meta-analysis of econometric research**



**Integrated Economic Analysis  
with CGE model, consideration of  
potential adaptations**



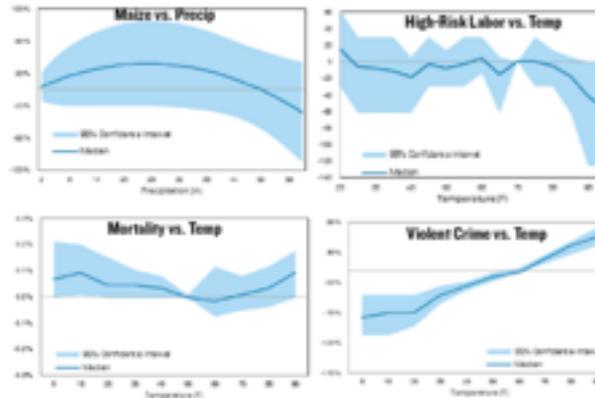
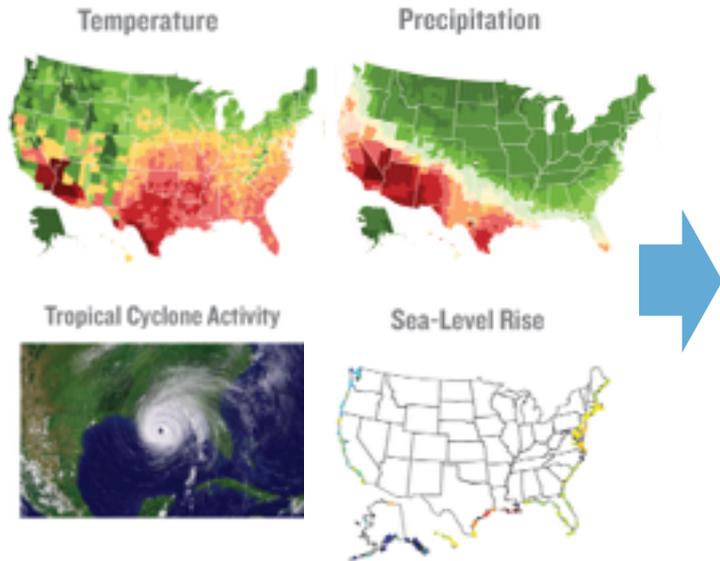
**Complementary detailed sectoral models**

# Research approach

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**Impact estimates based on meta-analysis of econometric research**

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Physical Climate Projections**



**Integrated Economic Analysis  
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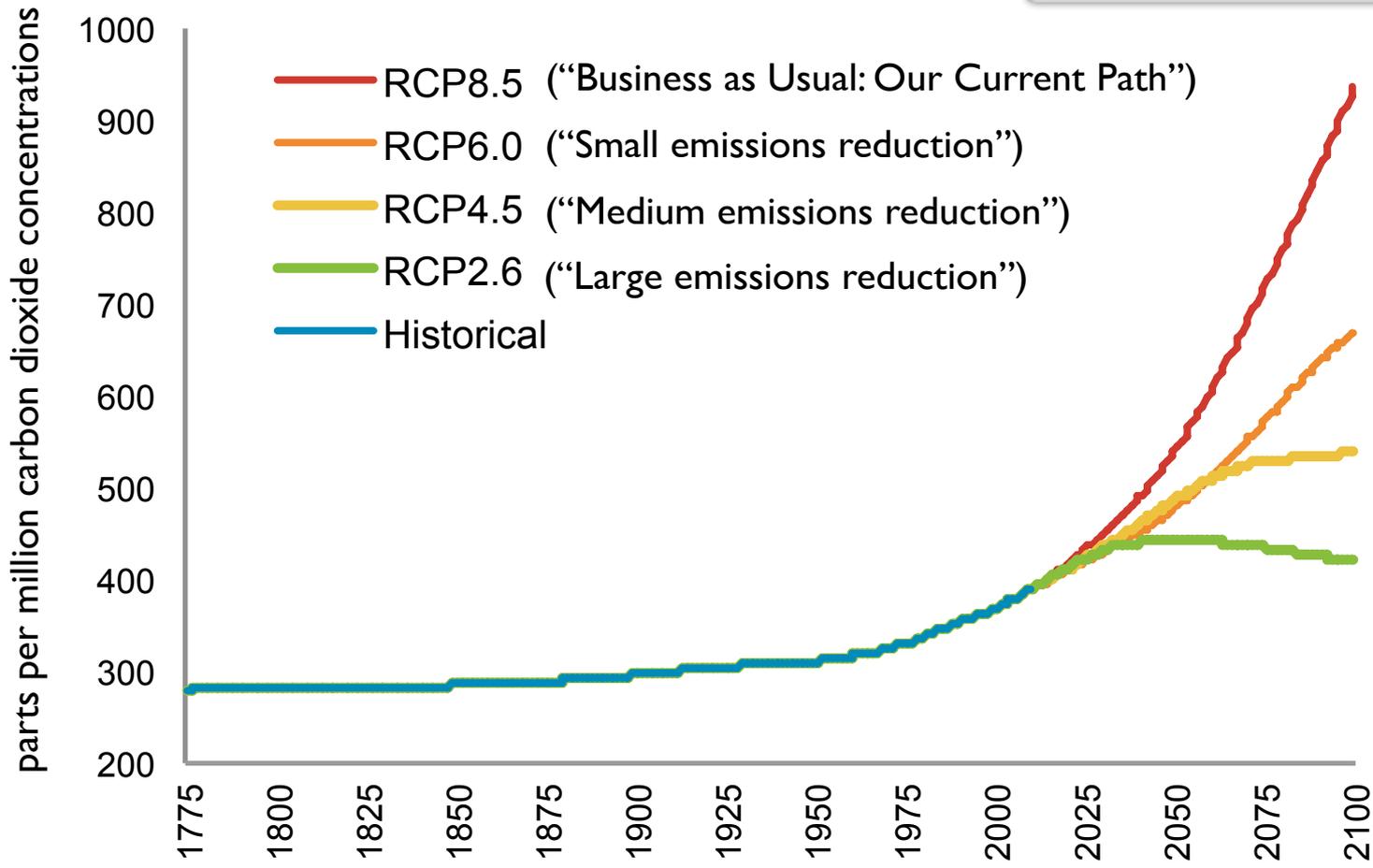
**Note: We did not try to construct a full IAM. In particular, we made the simplifying assumption of a constant spatial and sectoral economic structure**

**Complementary detailed sectoral models**

# Physical Science Projections

# Multiple emissions pathways reflect different socio-economic, technological and policy futures.

- Sources of uncertainty
- ◆ Socio-economic/Emissions
  - ◆ Global climate response
  - ◆ Regional climate response
  - ◆ Natural variability
  - ◆ Tipping points & the unknown

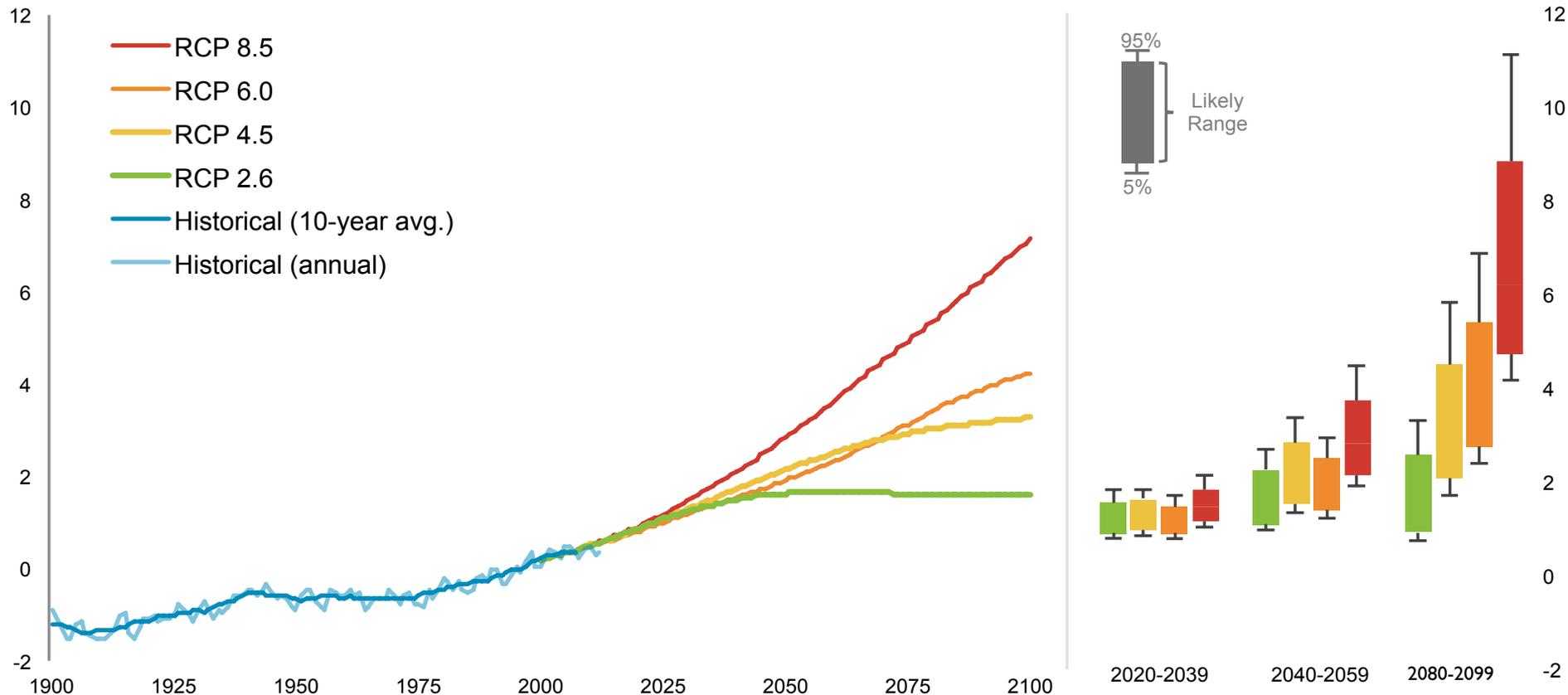


# The global temperature response to greenhouse gas forcing is uncertain.

## Sources of uncertainty

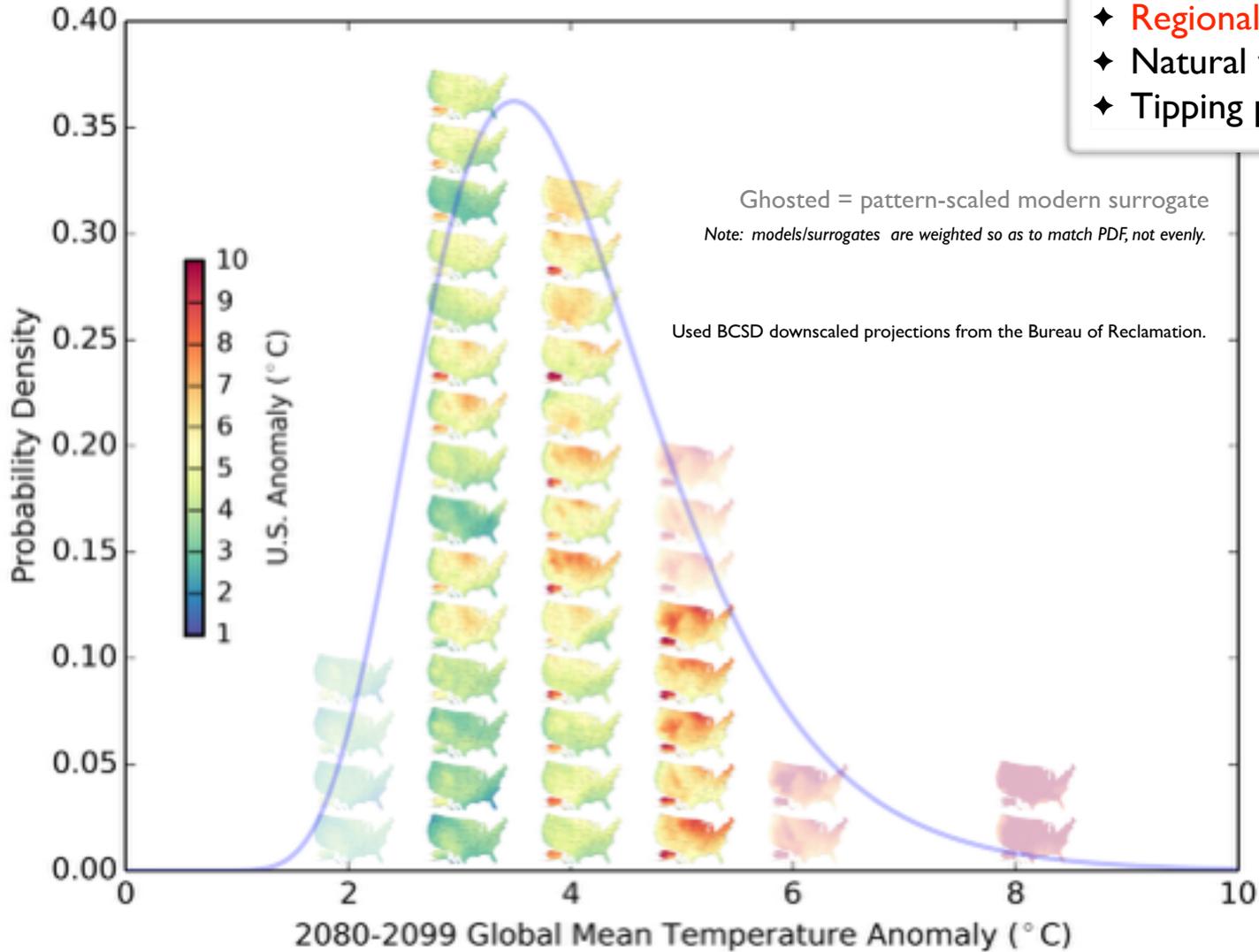
- ◆ Socio-economic/Emissions
- ◆ **Global climate response**
- ◆ Regional climate response
- ◆ Natural variability
- ◆ Tipping points & the unknown

Temperature projections (°F) from the MAGICC simple climate model, courtesy Malte Meinshausen



# As is the regional response that accompanies global warming...

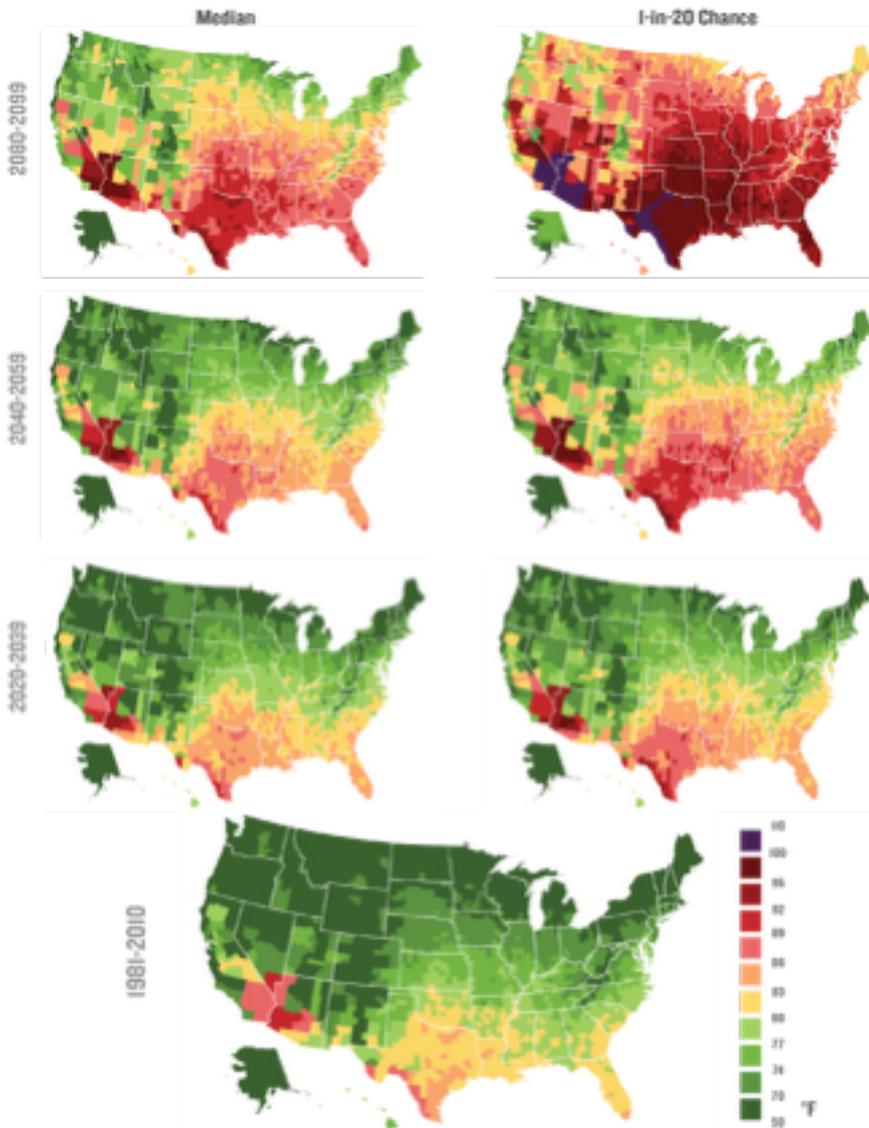
- Sources of uncertainty
- ◆ Socio-economic/Emissions
  - ◆ Global climate response
  - ◆ **Regional climate response**
  - ◆ Natural variability
  - ◆ Tipping points & the unknown



*Probability distribution developed from simple climate model and downscaled global climate model projections with the surrogate/model mixed ensemble method of Rasmussen & Kopp (in prep.).*

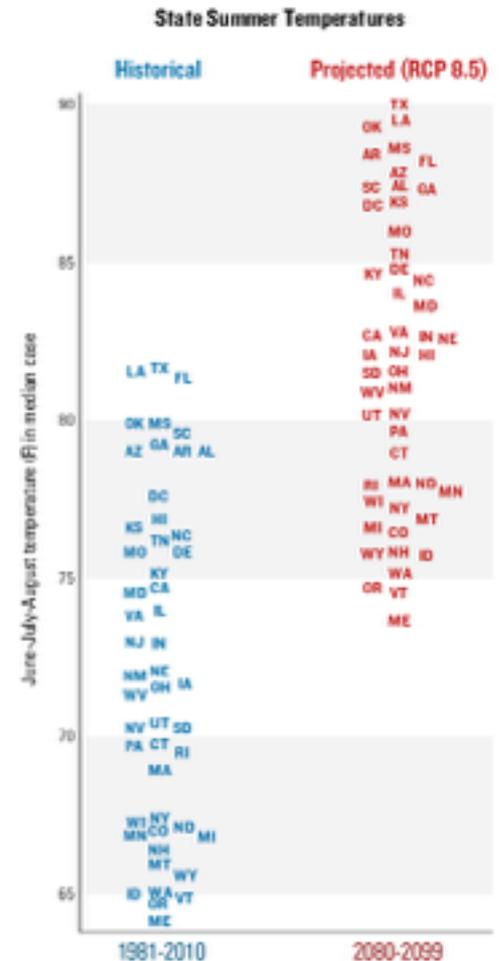
# which gives rise to uncertain projections of temperature change...

Median and 1-in-20 chance summer temperature projections (°F), RCP 8.5 (high emissions)



## Sources of uncertainty

- ◆ Socio-economic/Emissions
- ◆ Global climate response
- ◆ **Regional climate response**
- ◆ Natural variability
- ◆ Tipping points & the unknown



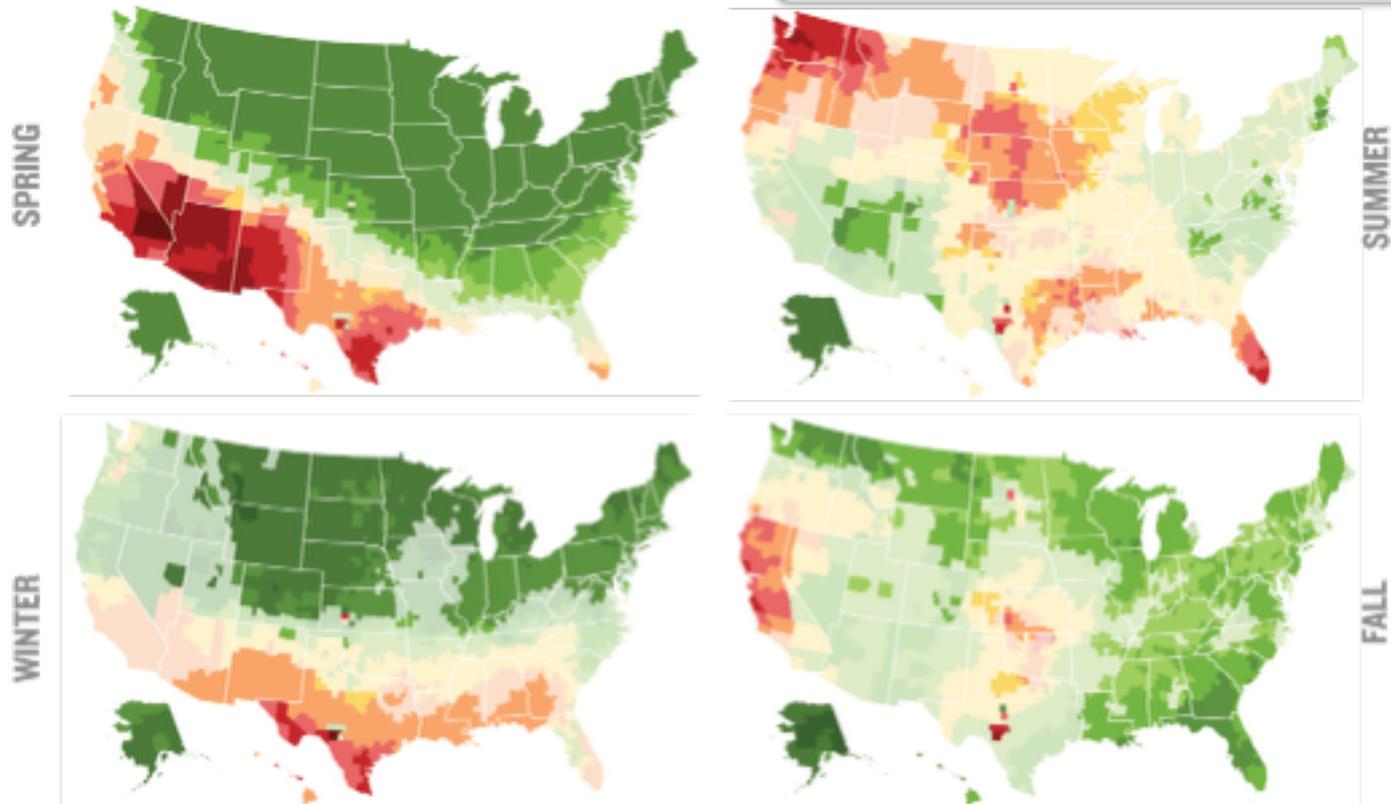
# ...and of precipitation change.

## Sources of uncertainty

- ◆ Socio-economic/Emissions
- ◆ Global climate response
- ◆ **Regional climate response**
- ◆ Natural variability
- ◆ Tipping points & the unknown

Median projected % precipitation change, RCP 8.5 (high emissions) in 2080-2099.

In the faded regions, an increase and an decrease are both about equally likely.



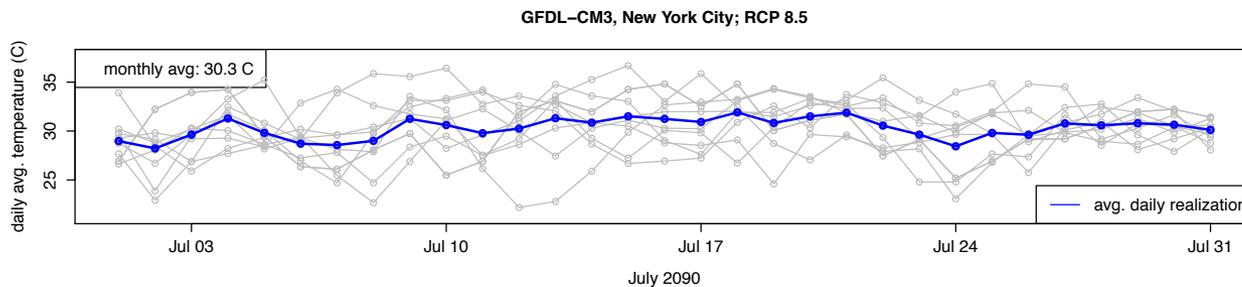
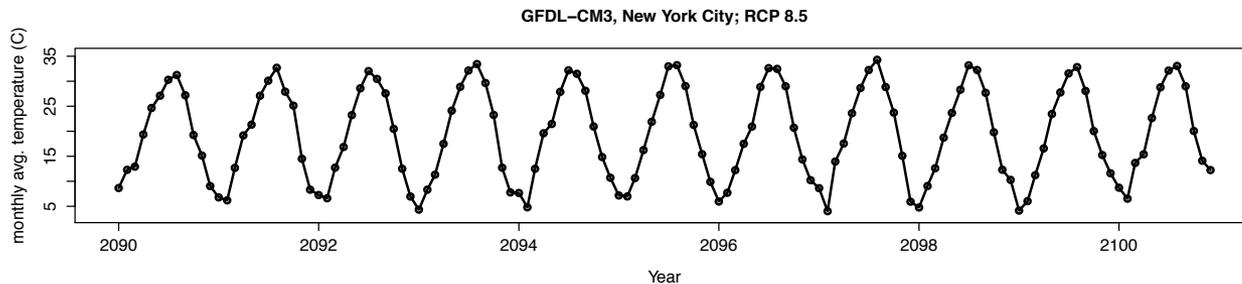
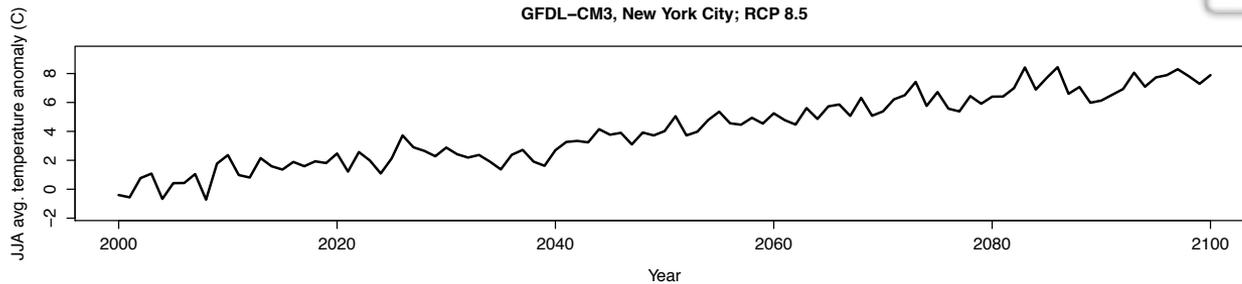
Increase and decrease about equally likely  
(both between 33% and 67% probability)



# Natural variability occurs at time scales ranging from the daily to the decadal.

## Sources of uncertainty

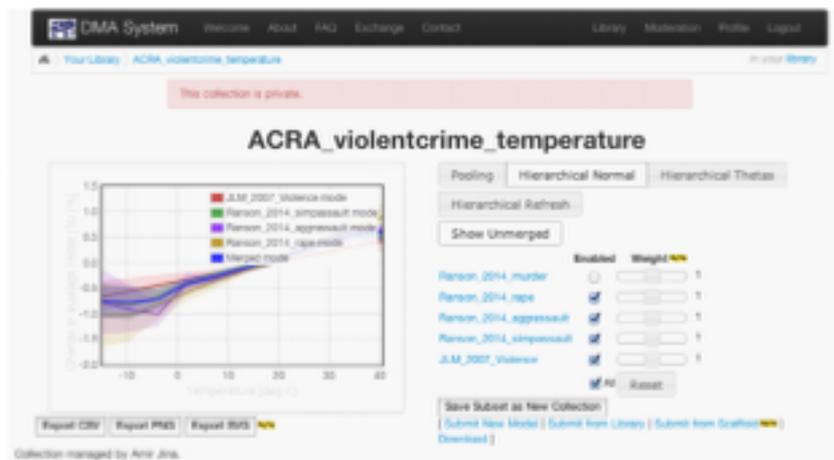
- ◆ Socio-economic/Emissions
- ◆ Global climate response
- ◆ Regional climate response
- ◆ **Natural variability**
- ◆ Tipping points & the unknown



# Meta-analysis of Econometric Research

# Distributed Meta-Analysis System

Researcher 1  
Researcher 2  
Researcher 3



American  
Climate  
Prospectus

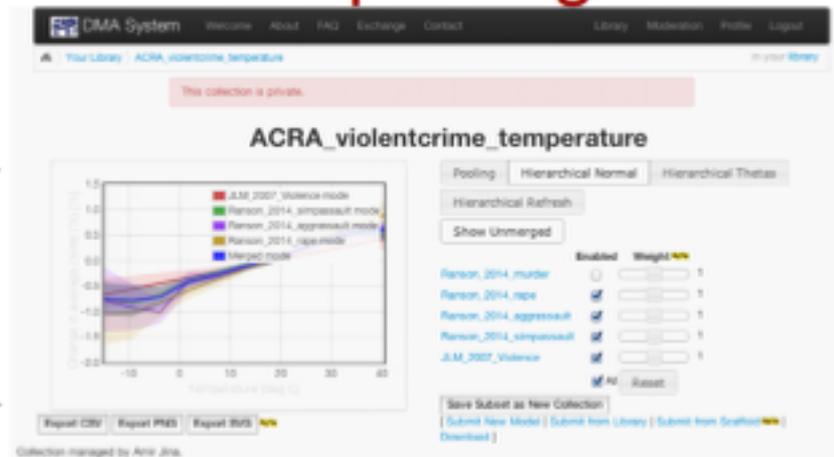
[dmas.berkeley.edu](http://dmas.berkeley.edu)

(Rising & Hsiang, 2014)

# Distributed Meta-Analysis System

Bayesian updating

Researcher 1  
Researcher 2  
Researcher 3  
Researcher 4



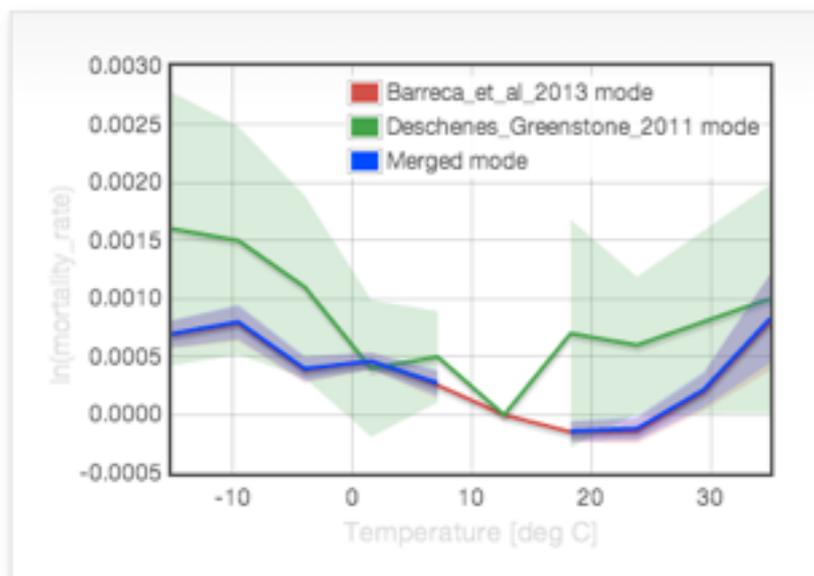
Revised American Climate Prospectus

[dmas.berkeley.edu](http://dmas.berkeley.edu)

(Rising & Hsiang, 2014)

# ACRA\_mortality\_temperature

Mortality temperature response function



Export CSV Export PNG Export SVG new

No Distribution ▼

	Enabled	Weight <small>new</small>
Deschenes_Greenstone_2011	<input checked="" type="checkbox"/>	<input type="text" value="1"/>
Barreca_et_al_2013	<input checked="" type="checkbox"/>	<input type="text" value="1"/>
	<input checked="" type="checkbox"/> All	<input type="button" value="Reset"/>

[\[ Submit New Model | Submit from Library | Submit from Scaffold new | Download \]](#)

Collection managed by Amir Jina.

Categories: ACRA Health

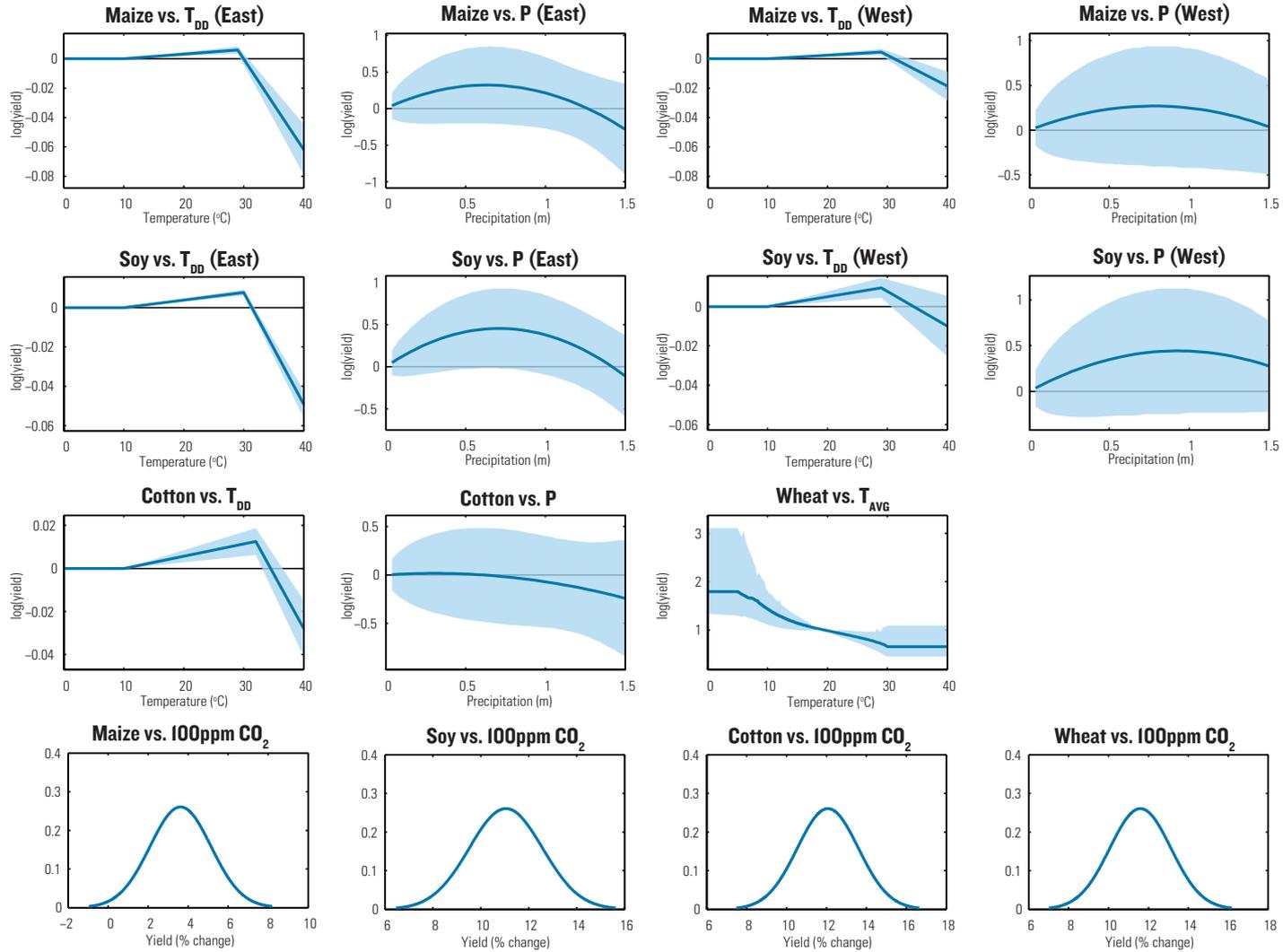
# Our criteria for inclusion

1. **Nationally representative**
2. **Analyze recent time-periods in US history**
3. **Robust to unobserved factors that differ across spatial units (jurisdictions, counties, or states)**
4. **Identify responses to high-frequency climatic variables (days or weeks)**
5. **Identify responses to the full distribution of temperature and rainfall measures**
6. **Account for temporal displacement**
7. **Account for seasonal patterns and trends in the outcomes**
8. **Ecologically valid (not laboratory studies)**

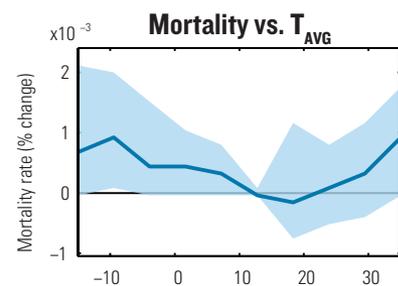
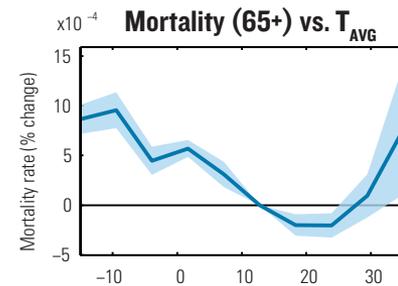
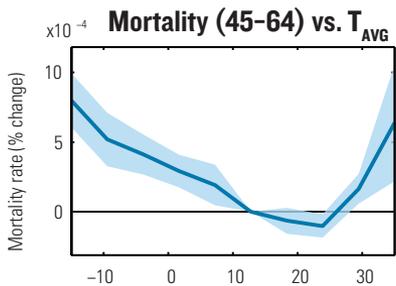
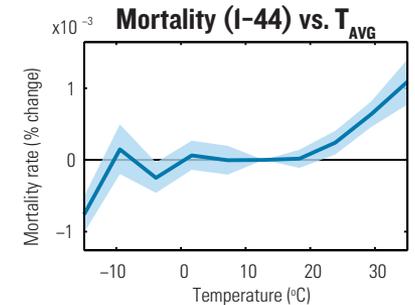
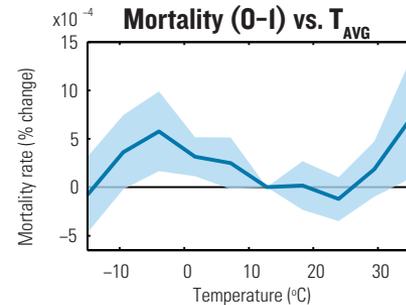
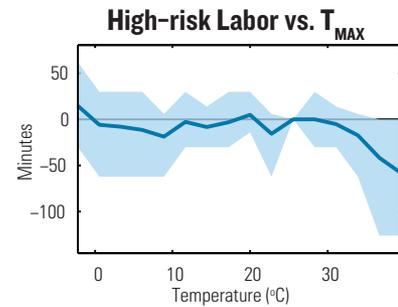
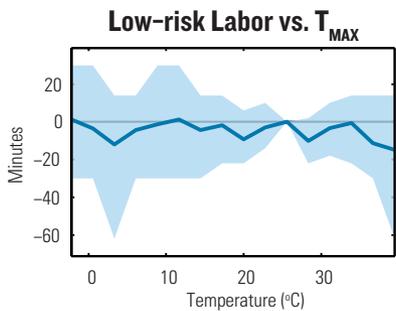
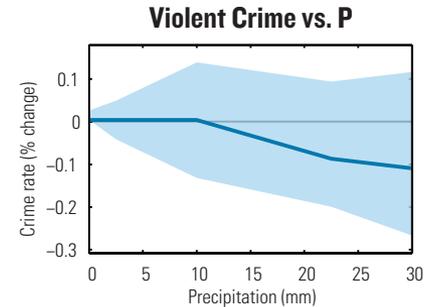
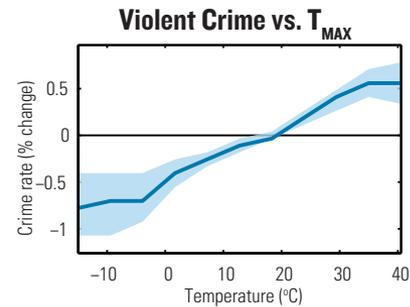
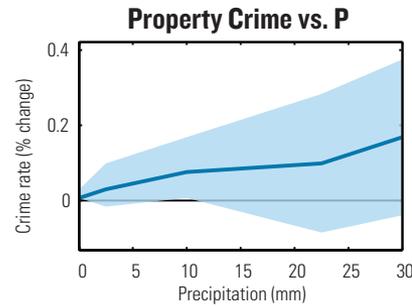
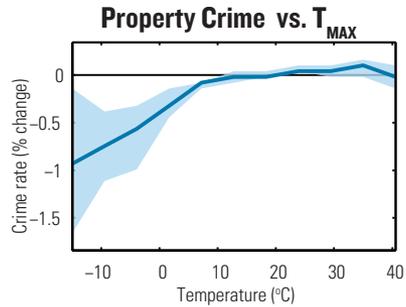
# Sectors where we apply this approach

- **Agriculture (maize, soy, cotton, wheat)**
  - Schlenker and Roberts (PNAS, 2009) - Cotton, Soy, Maize
  - Hsiang et al. (2013) - Wheat
  - McGrath and Lobell (ERL, 2013) - Carbon fertilization
  - Fisher et al. (AER, 2012) - Crop storage
- **Labor productivity (extensive margin only)**
  - Graff Zivin and Neidell (JLE, 2014)
- **Heat- and cold-related mortality (age resolved)**
  - Deschenes and Greenstone (AEJ, 2011)
  - Barreca et al. (2013)
- **Crime (violent and property)**
  - Jacob et al. (JHR, 2007)
  - Ranson (JEEM, 2014)
- **Electricity demand (residential)**
  - Auffhammer and Aroonruengsawat (CC, 2011)

# Impact functions – agriculture

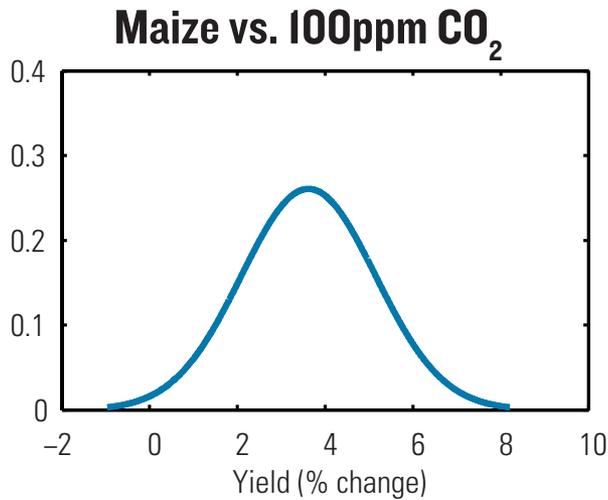
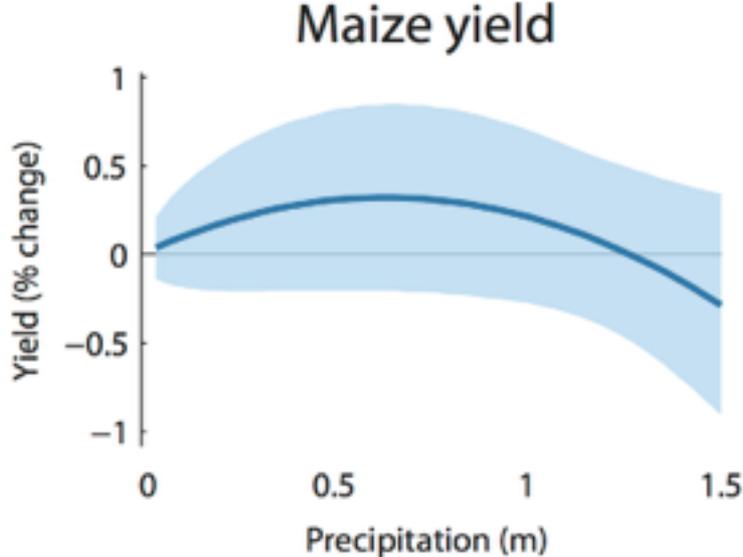
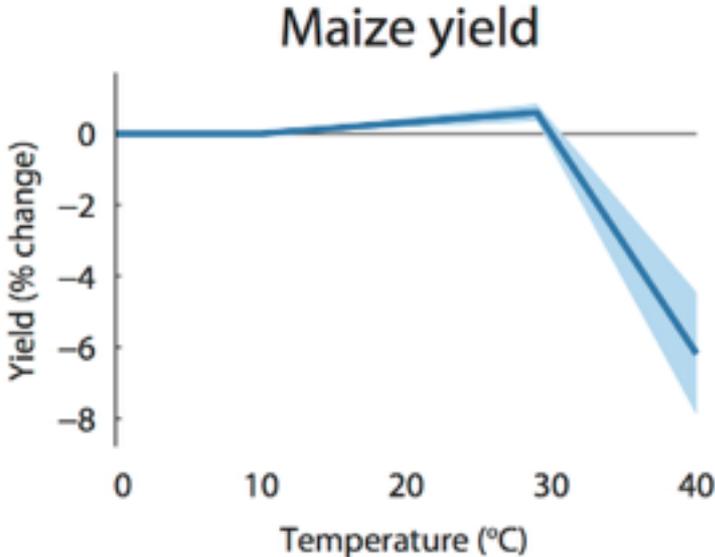


# Impact functions – other sectors



# Direct impacts

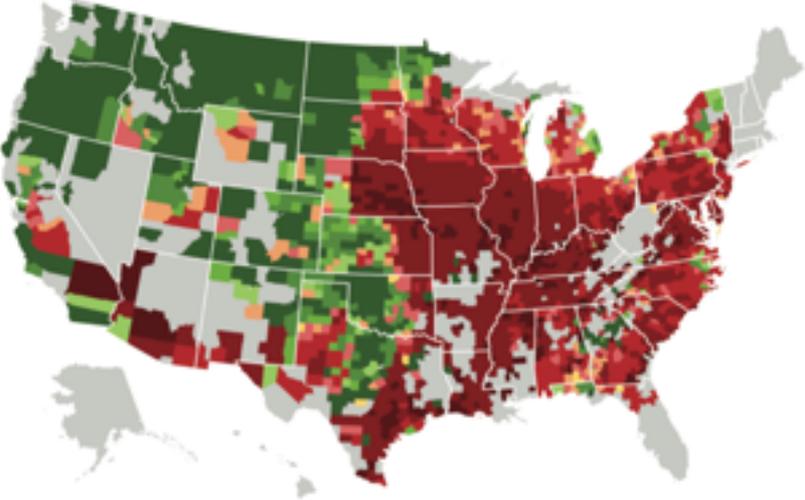
# Agriculture example: Dose-response



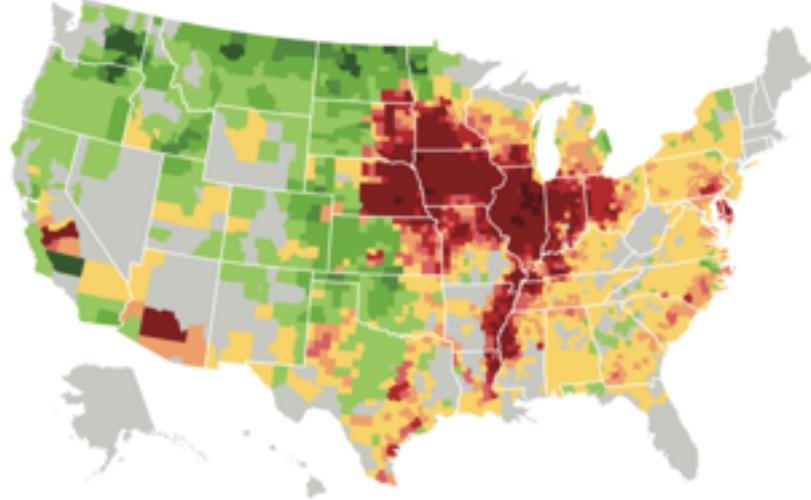
# Agriculture example: Median in RCP 8.5

2080-2099

### Relative Change in Yields Percent

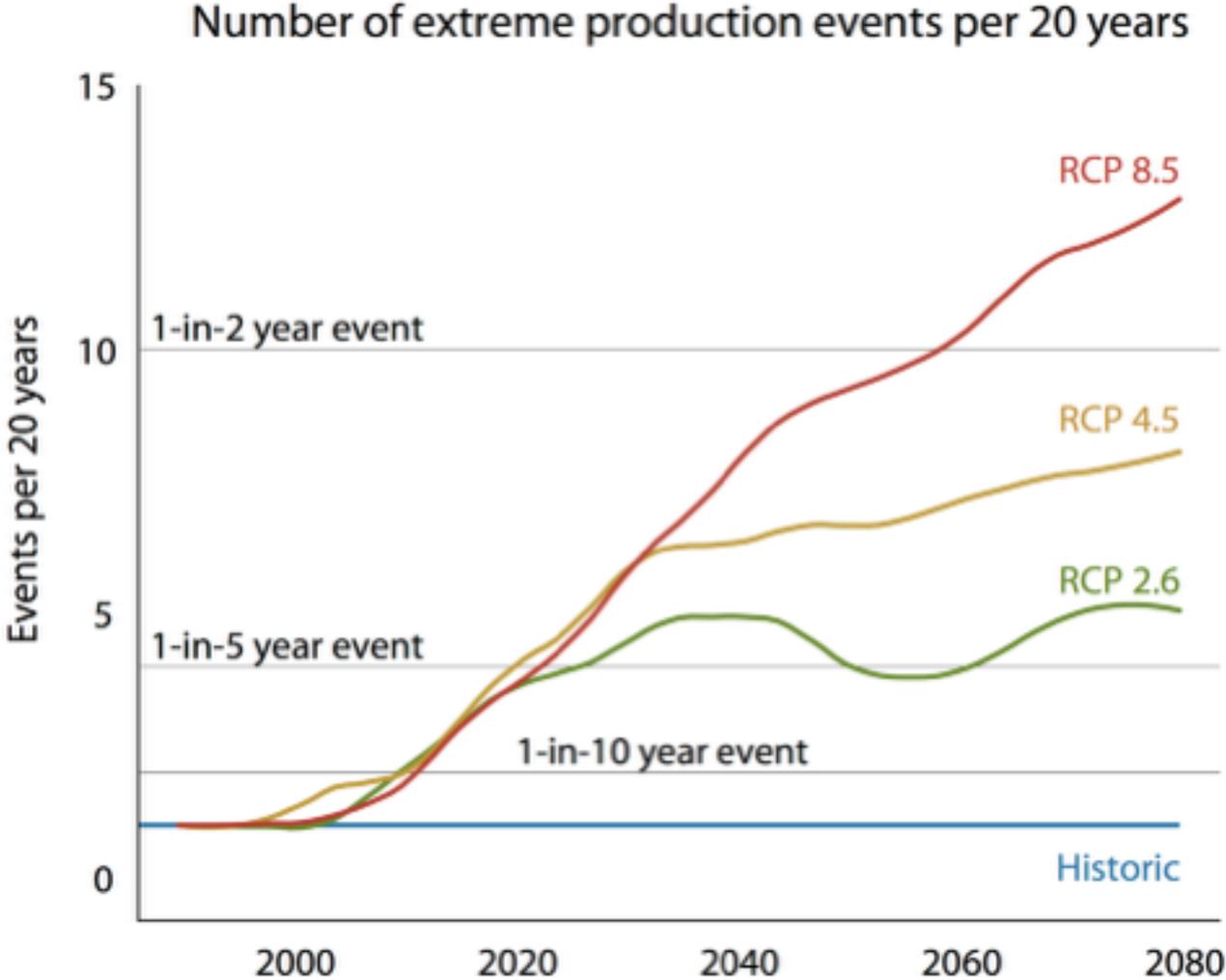


### Absolute Change in Production Thousand metric tons



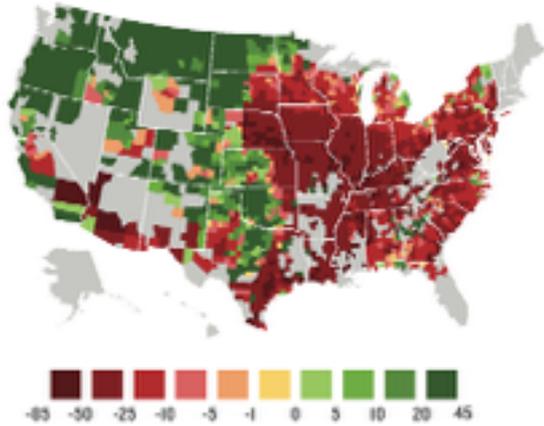
*(assuming current economy)*

# Agriculture example: Expected number of events

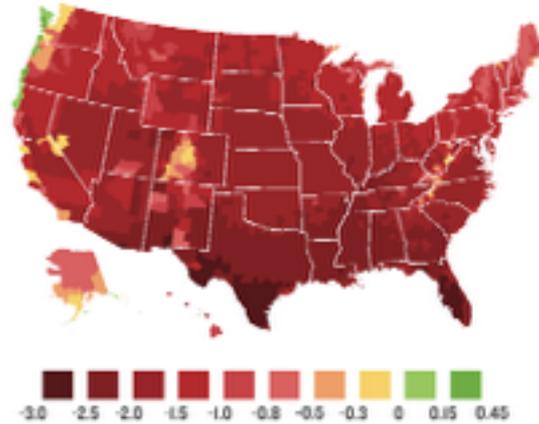


# Distribution of impacts: RCP 8.5, median

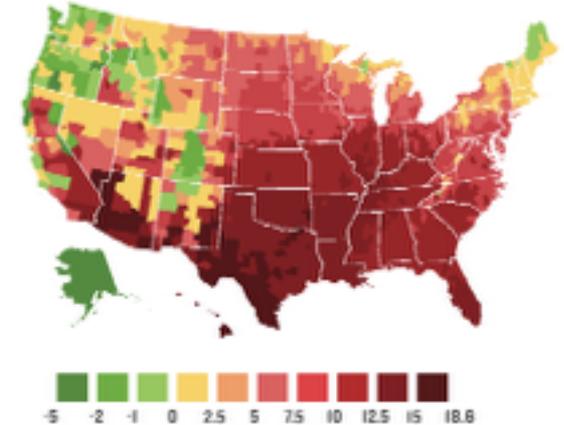
Relative Change in Yields  
Percent



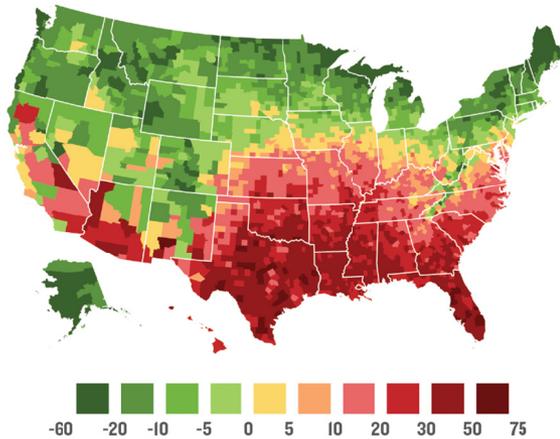
Relative Change in Labor Productivity  
Percent



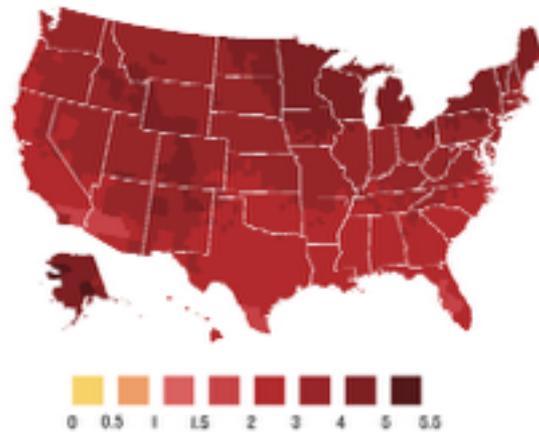
Relative Change in Electricity Demand  
Percent



Change in Mortality Rate  
Deaths per 100,000 People



Change in Violent Crime Rates  
Percent



# Coastal impacts

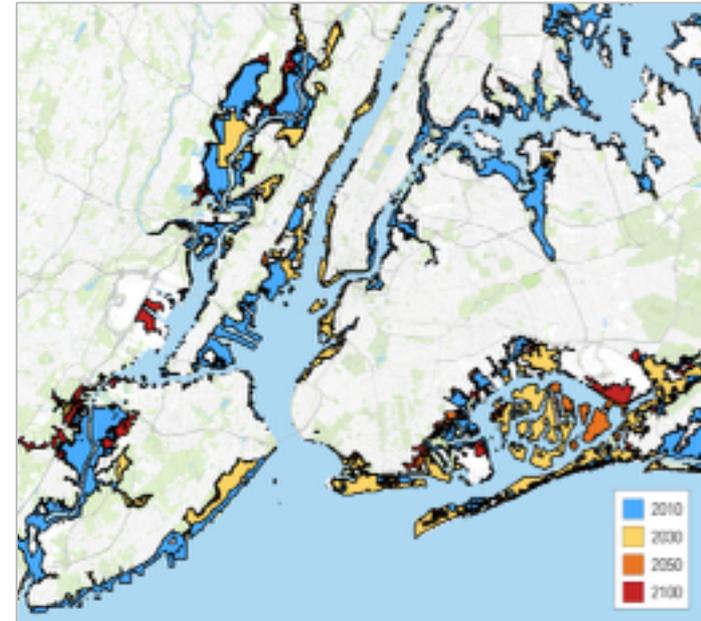
Tropical Cyclone Activity



Sea-Level Rise



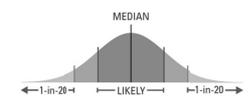
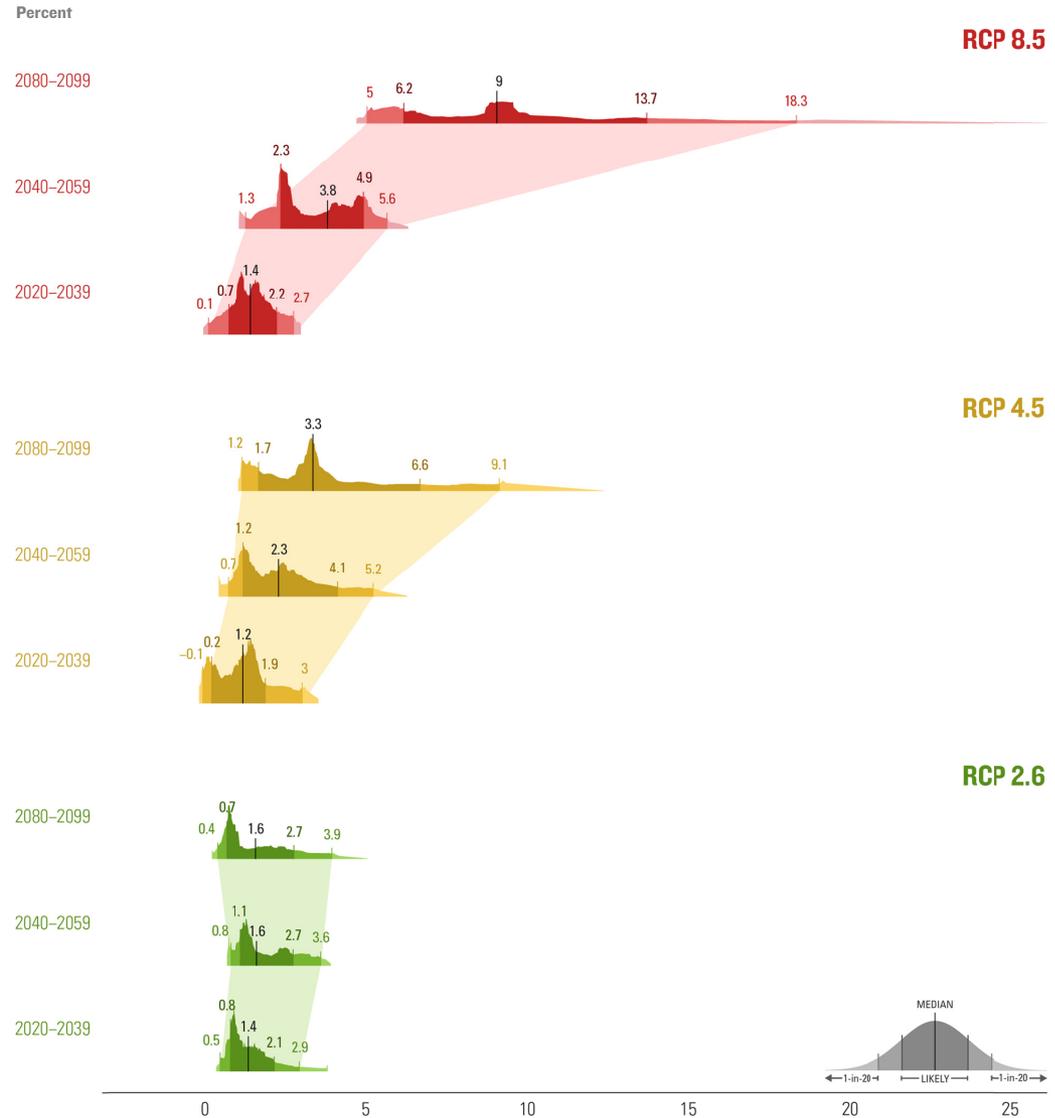
RMS North Atlantic Hurricane Model



# Energy demand

% increase in annual residential + commercial energy expenditures

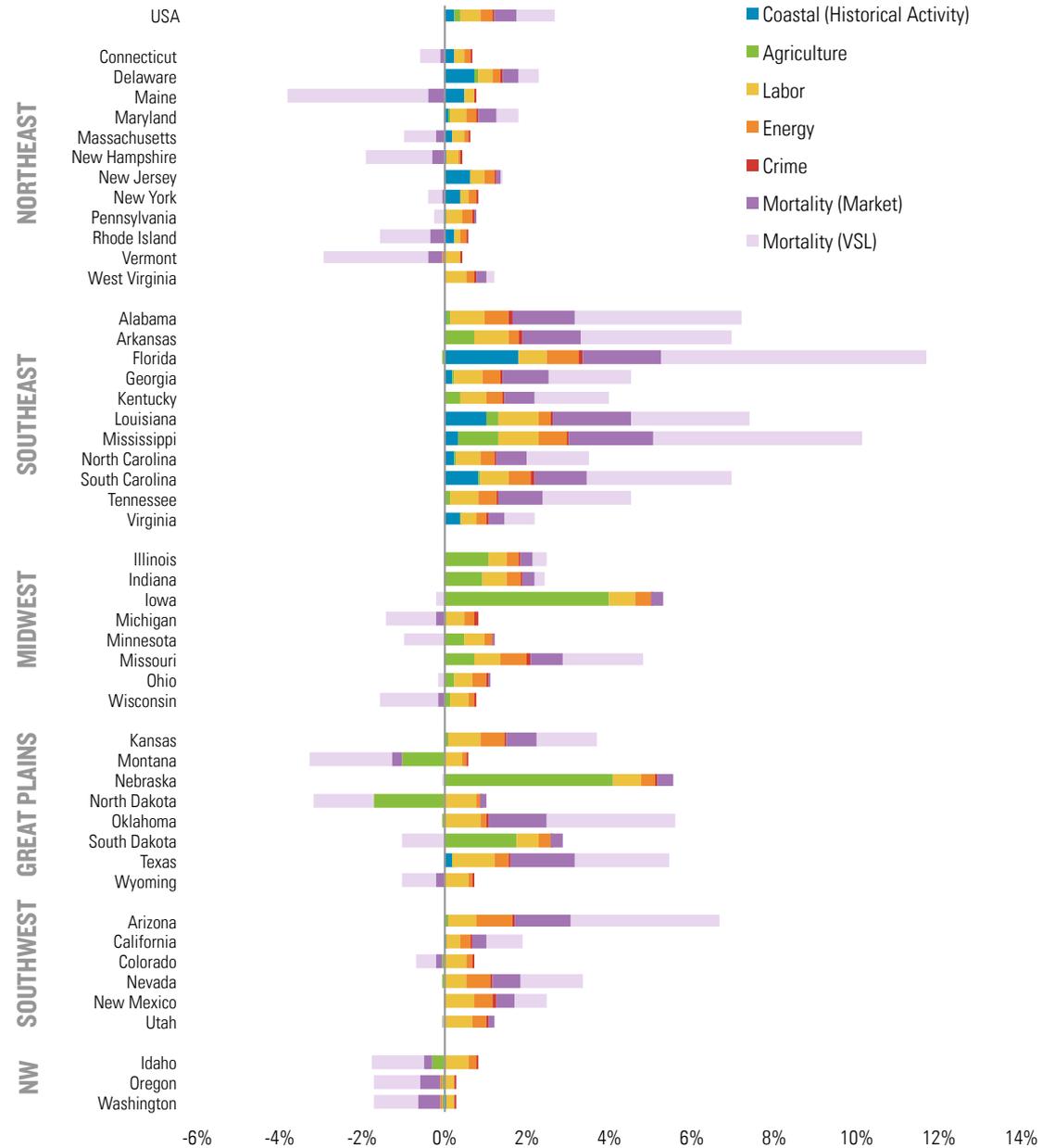
Impact function calibrated against RHG-  
National Energy Modeling System



# National aggregate impacts

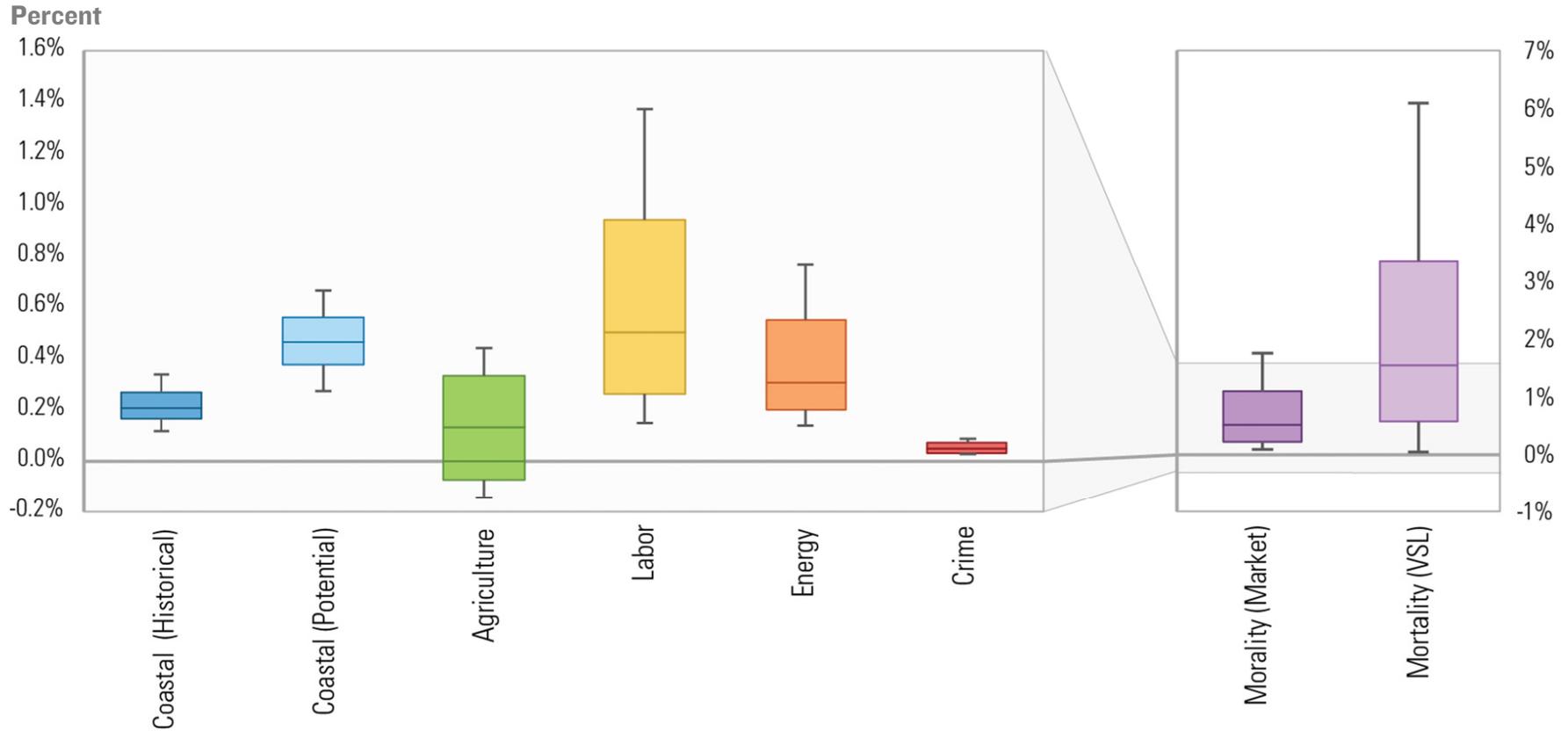
# Total cost and sectoral breakdown differ by region

RCP 8.5, median case, 2080-2099

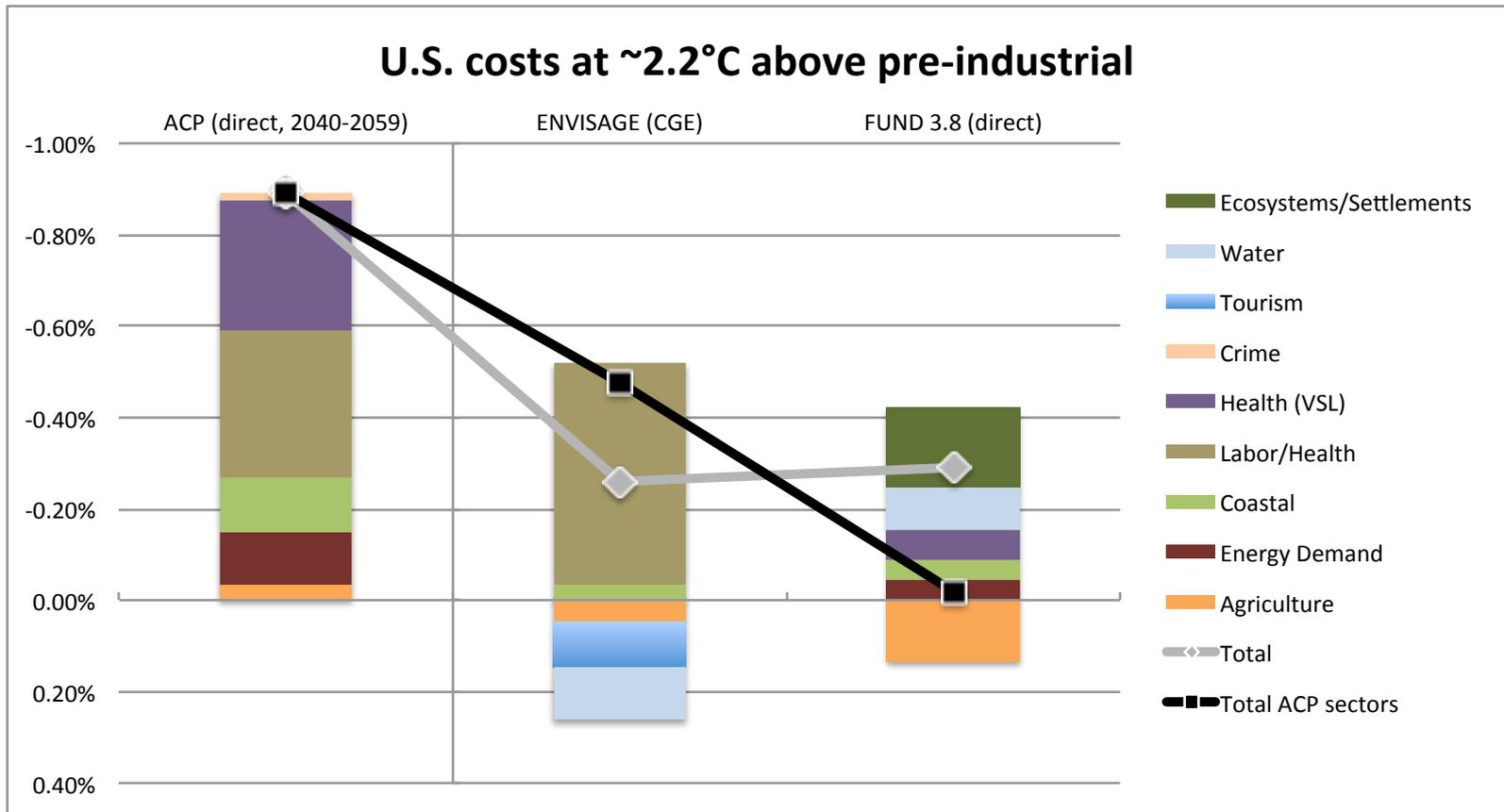


# Direct costs and benefits

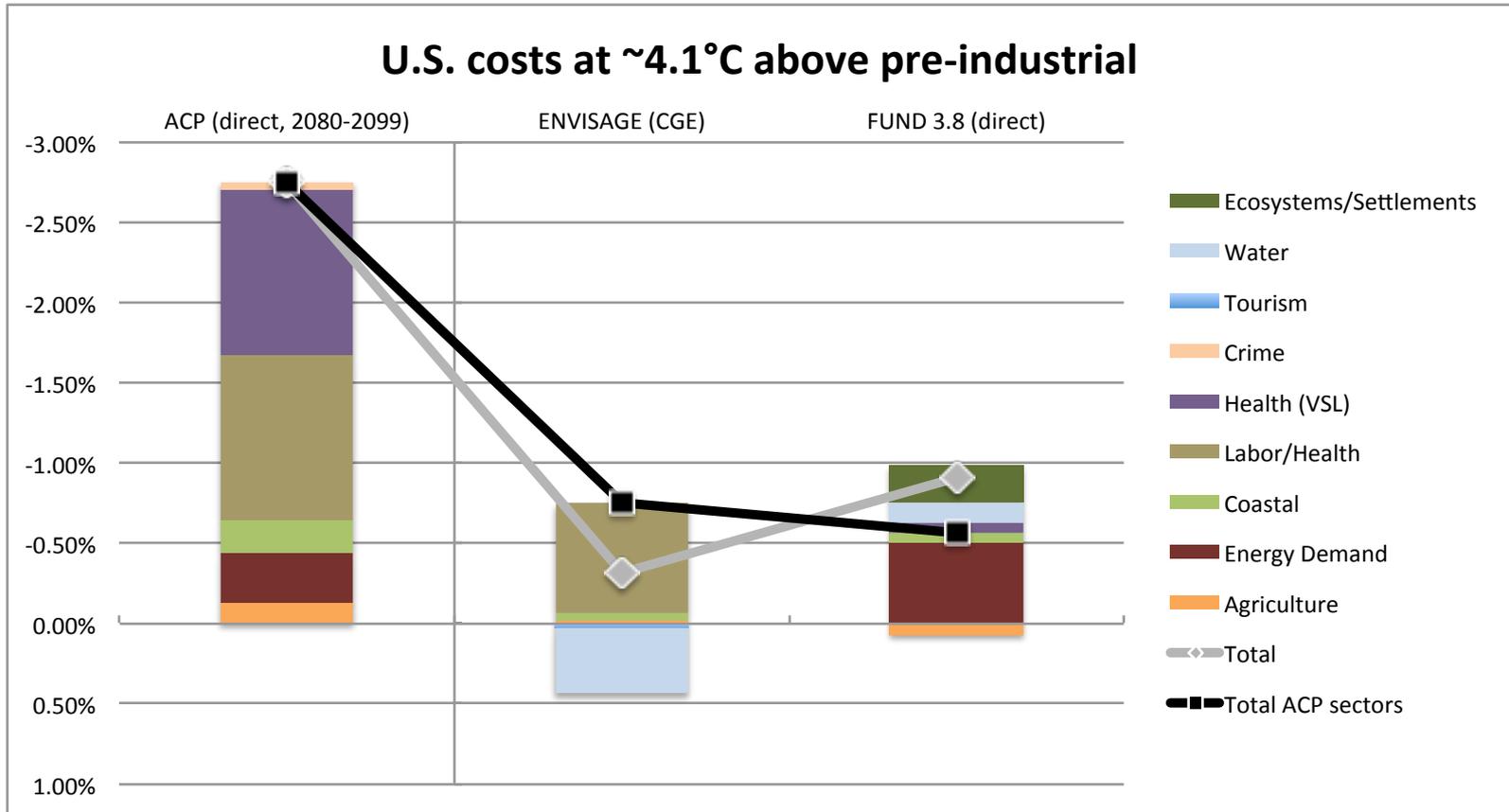
% of GDP, RCP 8.5, 2080-2099



# Comparison to benefit-cost integrated assessment models



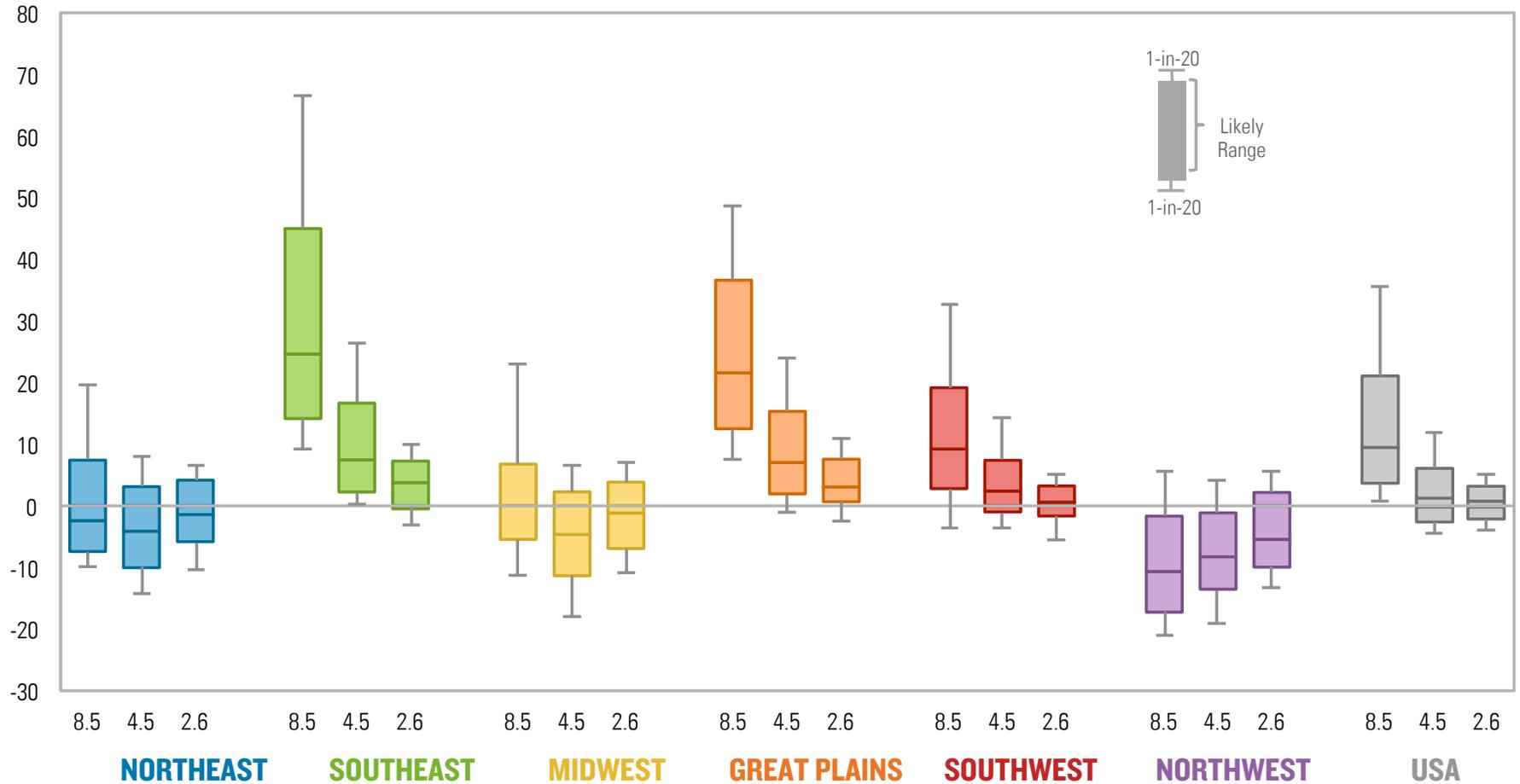
# Comparison to benefit-cost integrated assessment models



# Considering mitigation and adaptation

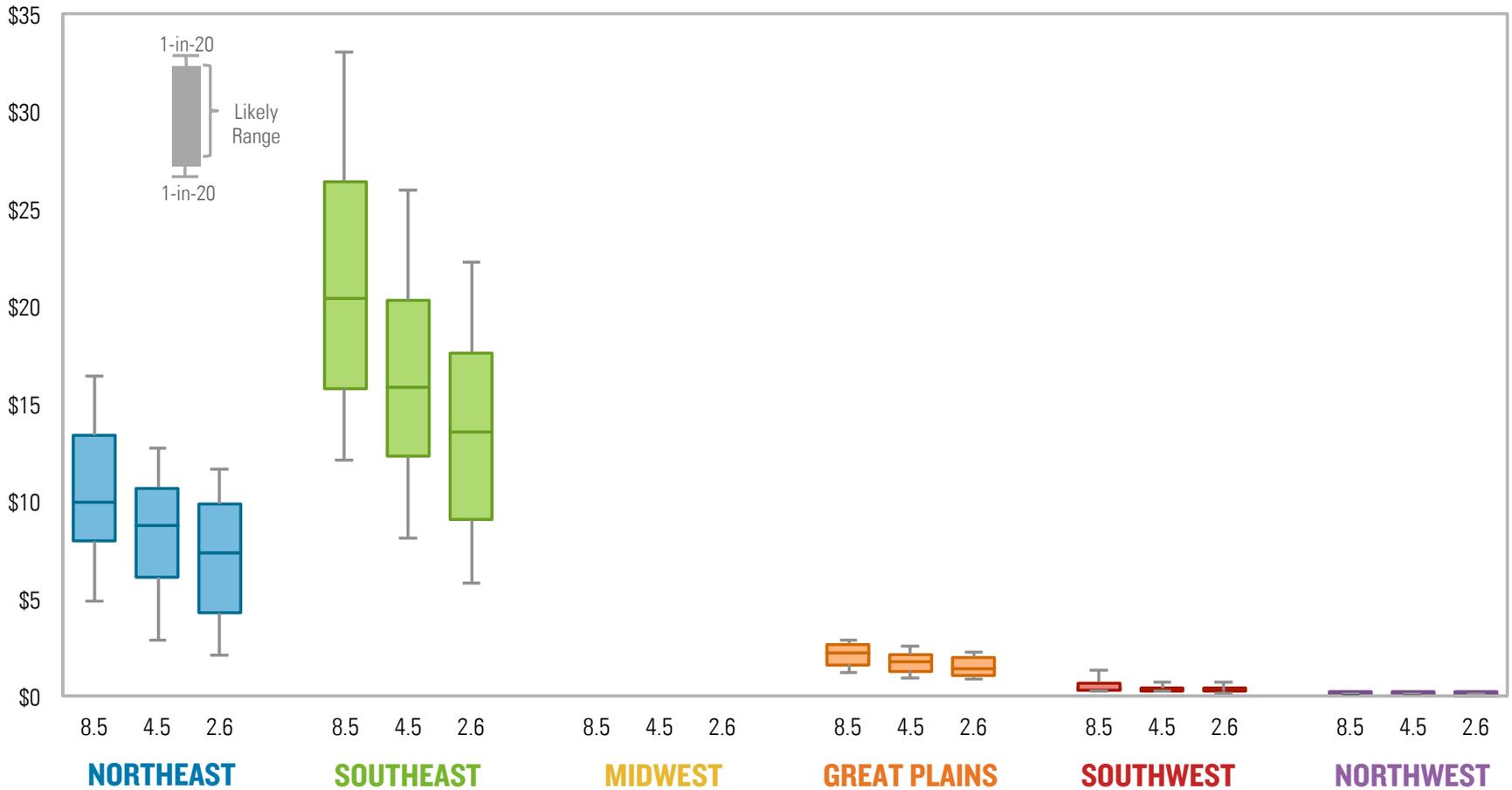
# Large mitigation benefit for mortality

Change in temperature-related deaths per 100,000 individuals, 2080-2099



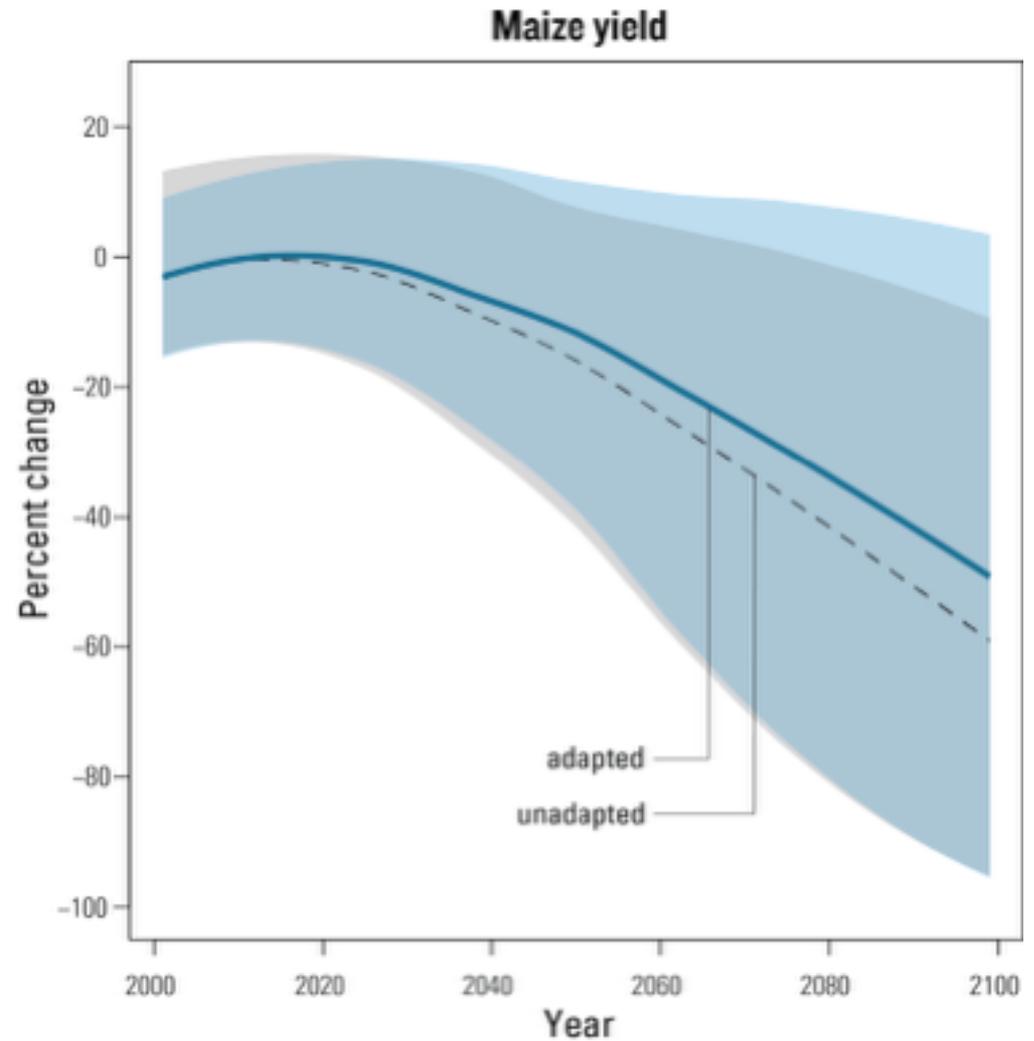
# Smaller mitigation benefit for coastal impacts in 21st century

Changes in average annual hurricane and coastal flood damage, 2080-2099  
Billion 2011 USD

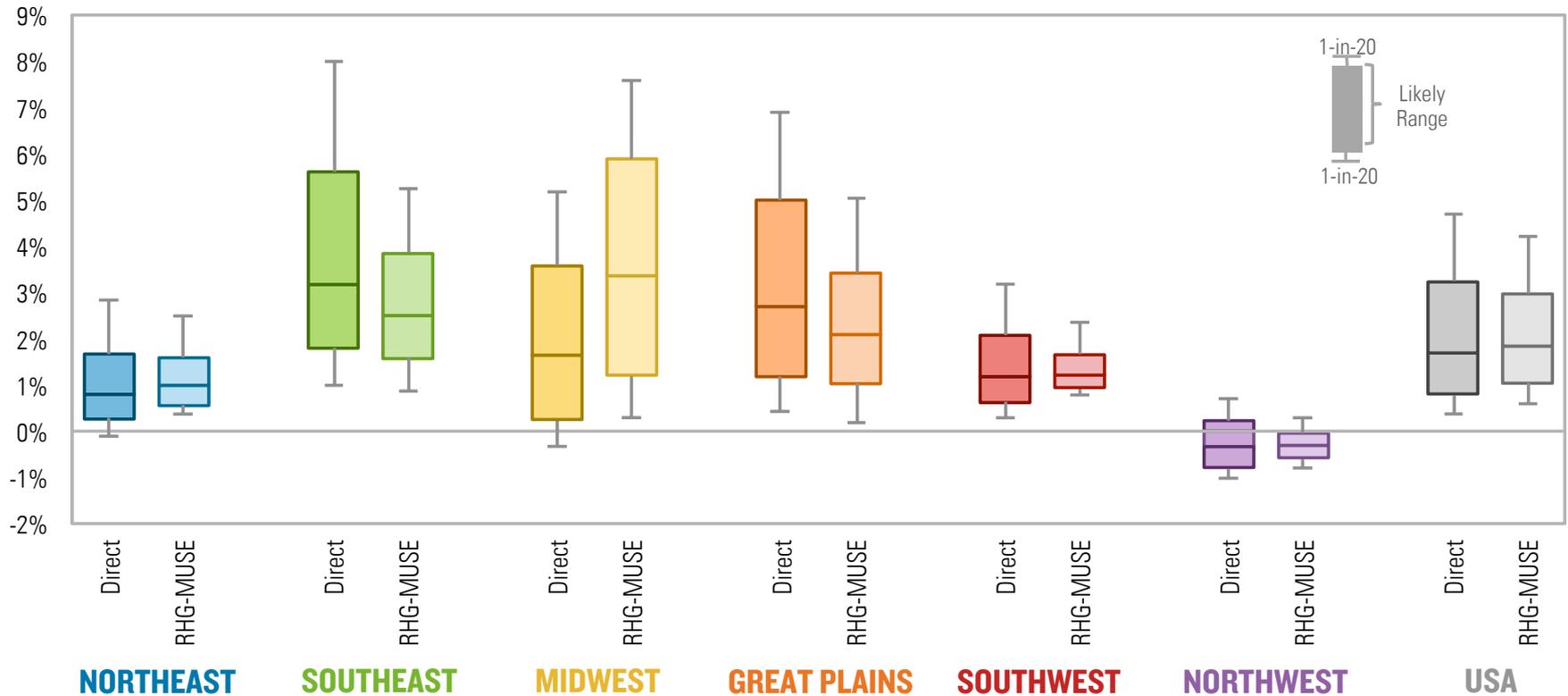


# Thought experiment: benefits of adaptation

What if T/P-yield relationship in the East evolves toward that in the West?

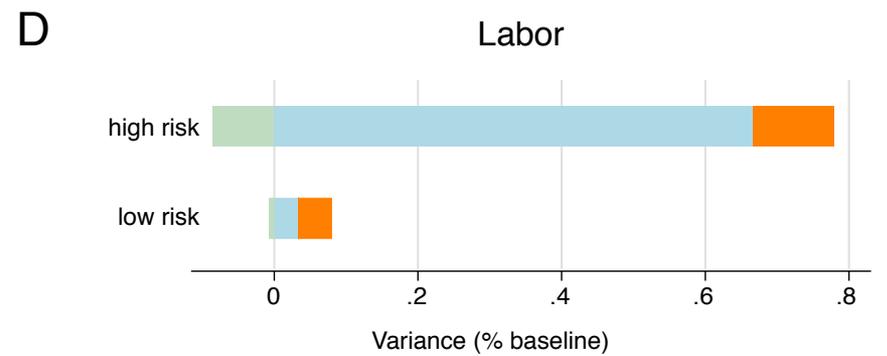
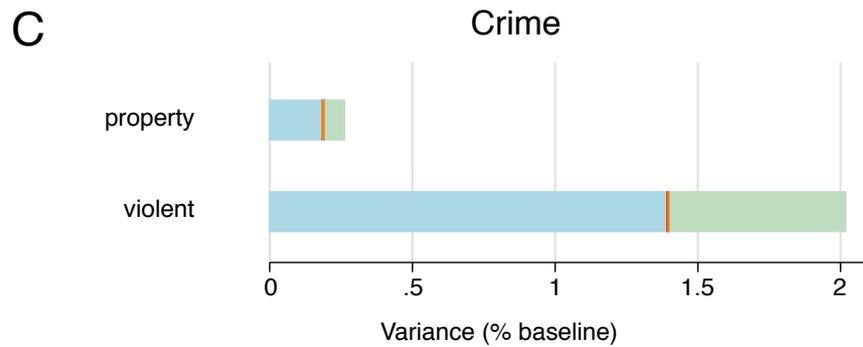
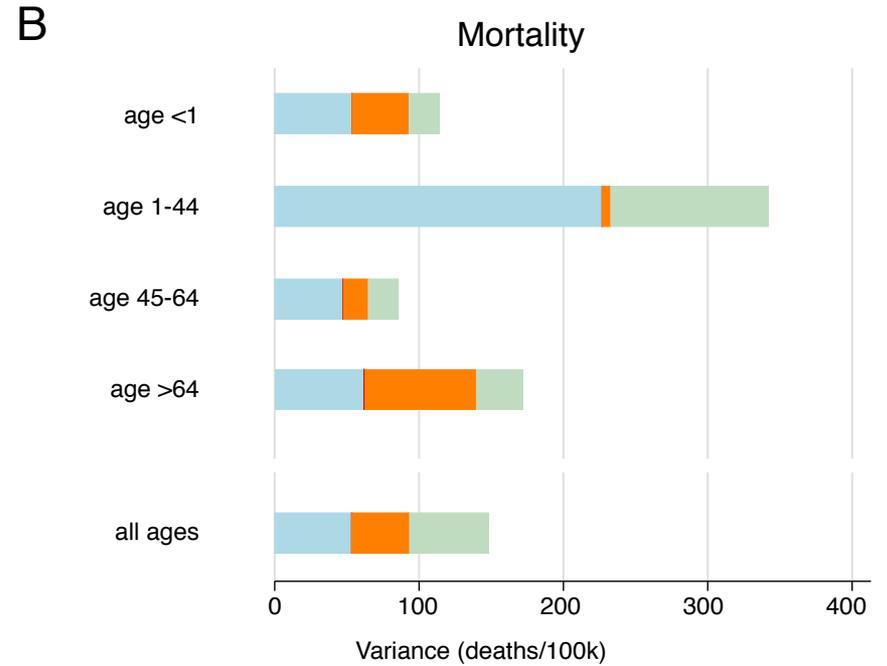
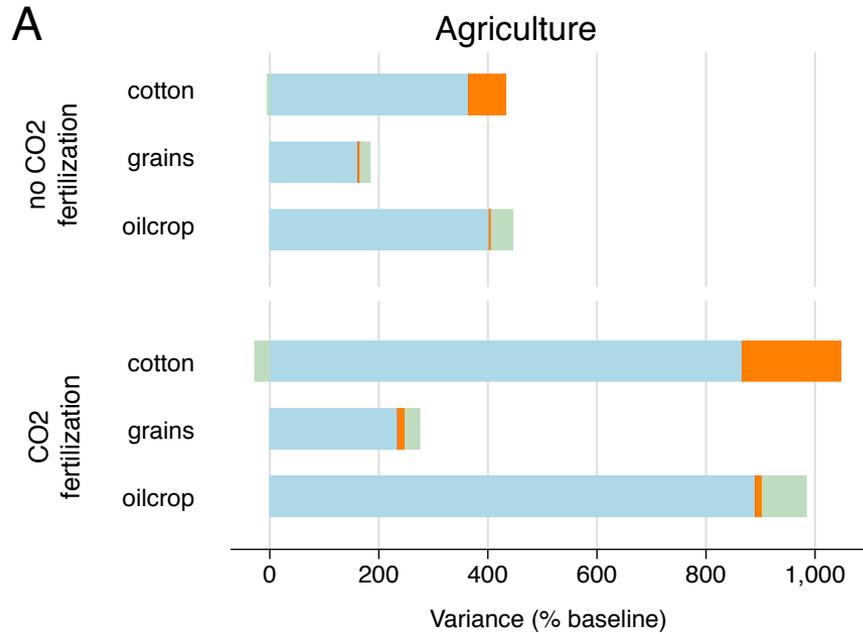


# Market adaptation via general equilibrium effects



# Uncertainty and inequality

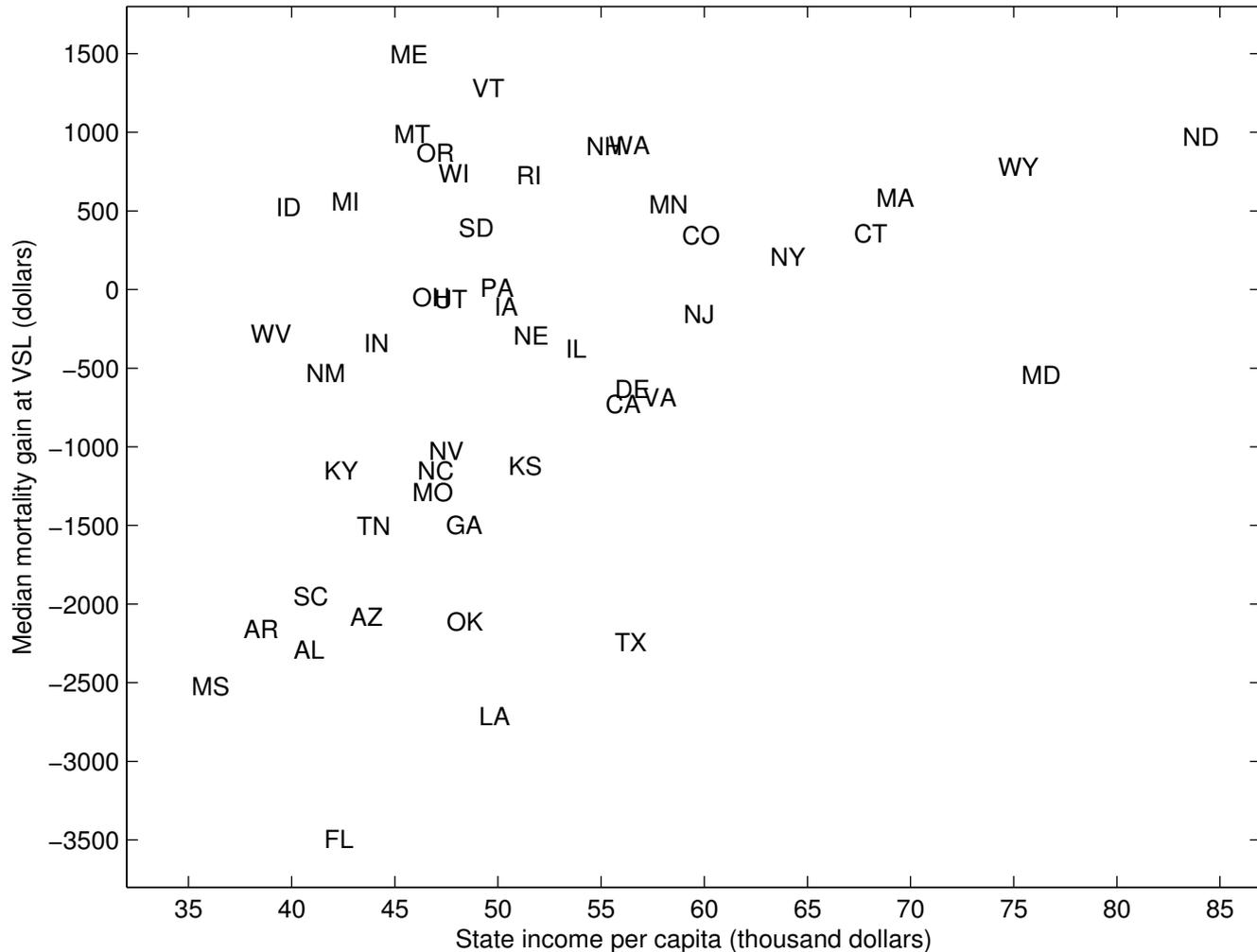
# Sources of uncertainty



■ model 
 ■ weather 
 ■ statistical 
 ■ interaction

# Death is the most unequal impact

RCP 8.5 median, 2080-2099, valued at VSL



**Average of 10 wealthiest states:**  
\$75 gain/capita  
(-1.1 deaths per 100,000)

**Average of 10 poorest states:**  
\$1900 loss/capita  
(27 deaths per 100,000)

# Valuing risk and inequality

certainty-equivalent consumption  $c_i^*$  given by 
$$c_i^* = \left( \sum_j p_j c_{i,j}^{1-\eta} \right)^{1/(1-\eta)}.$$

inequality-neutral equivalent % loss  $f$  given by 
$$\sum_i N_i v(c_i^*) = \sum_i N_i v(c_i^0 (1 - f))$$

**Table 15.1: Combined inequality-risk premiums for agricultural impacts, 2080-2099**

RCP 8.5, Premium as percentage of expected losses for maize, wheat, cotton, and soy output

		RRA						
		<-- Low risk aversion			High risk aversion -->			
		0	2	4	6	8	10	
C/A	High inequality tolerance -->	0	0%	0%	1%	1%	2%	2%
		2	13%	13%	14%	14%	15%	15%
		4	21%	21%	22%	22%	23%	23%
		6	26%	27%	27%	28%	28%	29%
		8	32%	33%	33%	34%	34%	35%
		10	40%	41%	41%	42%	42%	42%
	<-- Low tolerance for inequality							

**Table 15.2: Combined inequality-risk premiums for mortality impacts, 2080-2099**

RCP 8.5, Premium as percentage of expected losses, applying value of a statistical life

		RRA						
		<-- Low risk aversion			High risk aversion -->			
		0	2	4	6	8	10	
C/A	High inequality tolerance -->	0	0%	3%	6%	10%	14%	18%
		2	28%	32%	36%	41%	46%	52%
		4	58%	64%	70%	76%	83%	91%
		6	90%	97%	104%	113%	121%	131%
		8	121%	129%	138%	148%	159%	170%
		10	150%	159%	170%	181%	193%	206%
	<-- Low tolerance for inequality							

## Take-aways

- By 2020-2039, median projected average summer temperatures in DC match and the expected number of dangerously humid summer days exceed those of Mississippi today.
- By 2080-2099 under RCP 8.5, the NE, SE, and MW south of the Mason-Dixon lines have median projected summer T hotter than Louisiana today, and even north of M-D line have more expected dangerously humid days than Louisiana.
- Mortality & Labor are largest costs in the US; Energy & Coastal impacts sizable.
- Median projected increase in deaths under RCP 8.5, 2080-2099, is about 10 per 100,000, similar to current traffic death rate.
- Cost of Crime  $\approx$  cost to Agriculture (small in \$).

# Take-aways

- Largest sources of uncertainty continues to be driven by physical climate.
- Nonlinear response functions → South and Midwest lose most in the sectors we quantified.
- Inequality impact on welfare is large (likely exceeds risk effect), with mortality the largest source of inequality (RCP 8.5 2080-2099 median, mortality in wealthiest 10 states falls ~1/100,000 and in poorest 10 states rises ~30/100,000).
- Mitigation benefits largest and most certain for labor, mortality, energy, and crime. Agriculture benefits less clear because of carbon fertilization; coastal because of slow response of the system.

# Potentially generalizable innovations in SEAGLAS

- [DMAS.berkeley.edu](https://DMAS.berkeley.edu) – broadly applicable to statistical analyses, lowers the cost of cross-disciplinary communication through automation.
- Framework for probabilizing GCM projections
- Approach to micro-founding damage functions
- Coupling of downscaled GCMs and impact functions in a coherent probabilistic framework to quantify risk and inequality



# American Climate Prospectus

Economic Risks in the United States

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Climate Change Impacts/Integrated Assessment | Snowmass, CO | July 22, 2014