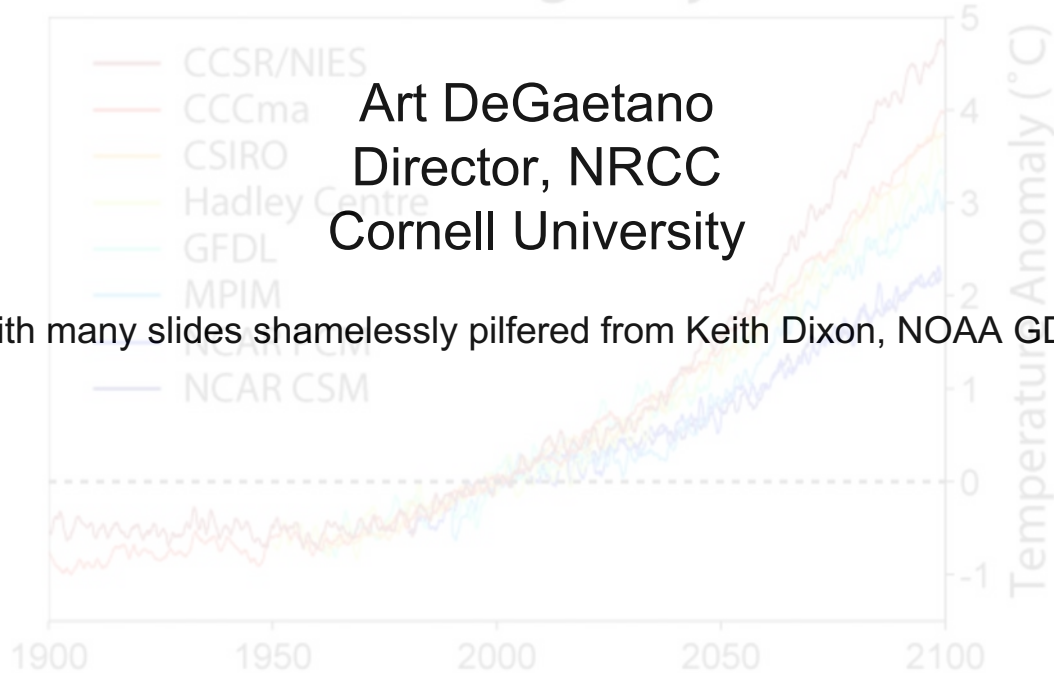


A Quick Overview of Climate Projections

Global Warming Projections



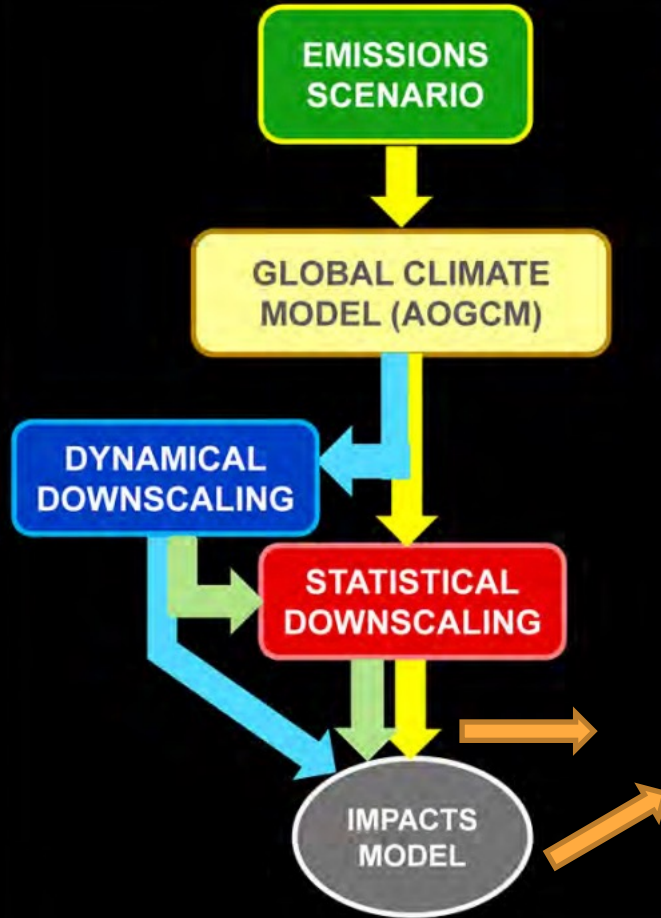
Art DeGaetano
Director, NRCC
Cornell University

With many slides shamelessly pilfered from Keith Dixon, NOAA GFDL



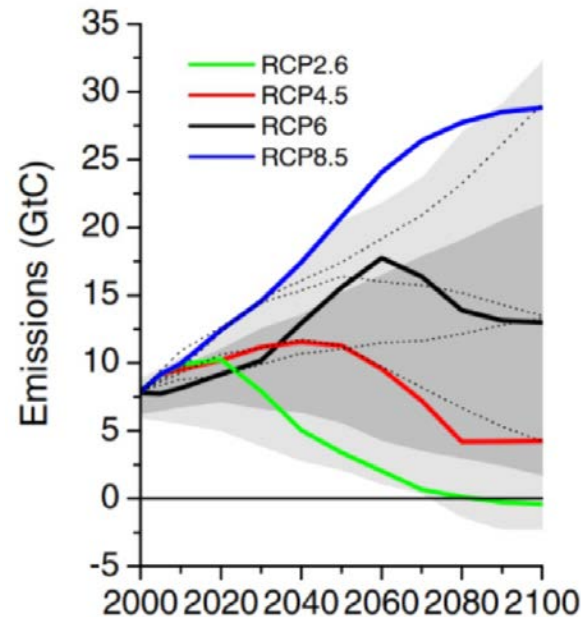
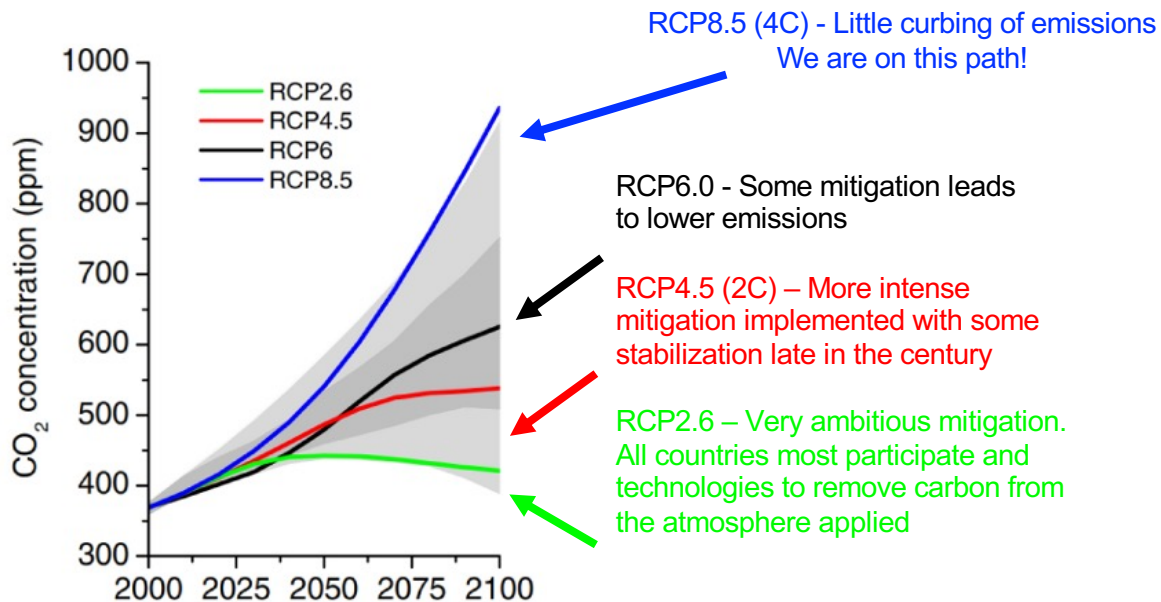
A simplified view of the process...

...mindful that
“uncertainties”
are introduced
in each step of
the process



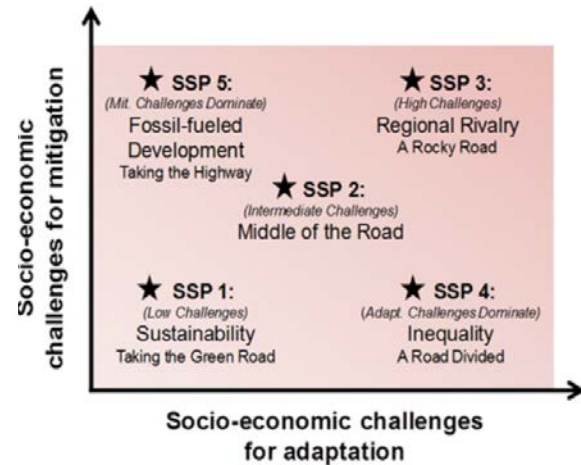
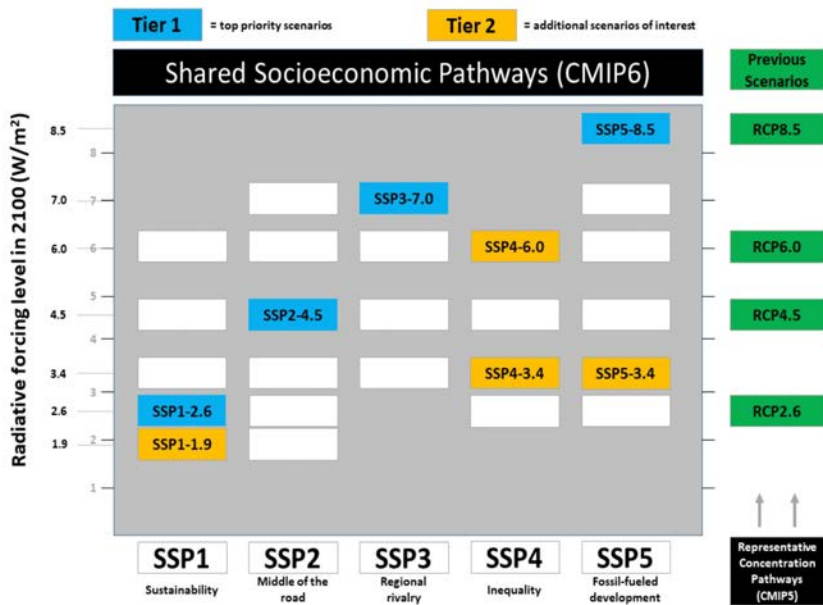
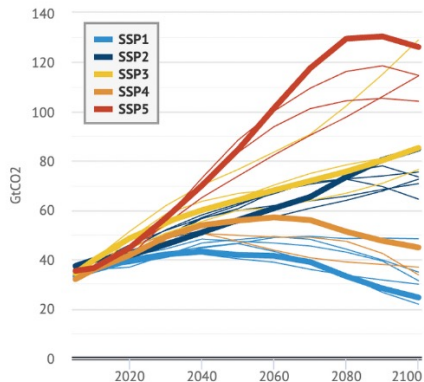
Projections!!!!

Understanding Societies Priorities and Pathways



Understanding Societies Priorities and Pathways

CO2 emissions for SSP baselines



Anatomy of a Climate Model

Equations (all models)

Conservation of momentum:

$$\frac{\partial \vec{V}}{\partial t} = -(\vec{V} \cdot \nabla) \vec{V} - \frac{1}{\rho} \nabla p - \vec{g} - 2\vec{\Omega} \times \vec{V} + \nabla \cdot (k \nabla \vec{V}) - \vec{F}_d$$

Conservation of energy:

$$\rho c_p \frac{\partial T}{\partial t} = -\rho c_p (\vec{V} \cdot \nabla) T - \nabla \cdot \vec{R} + \nabla \cdot (k \nabla T) + C + S$$

Conservation of mass:

$$\frac{\partial \rho}{\partial t} = -(\vec{V} \cdot \nabla) \rho - \rho (\nabla \cdot \vec{V})$$

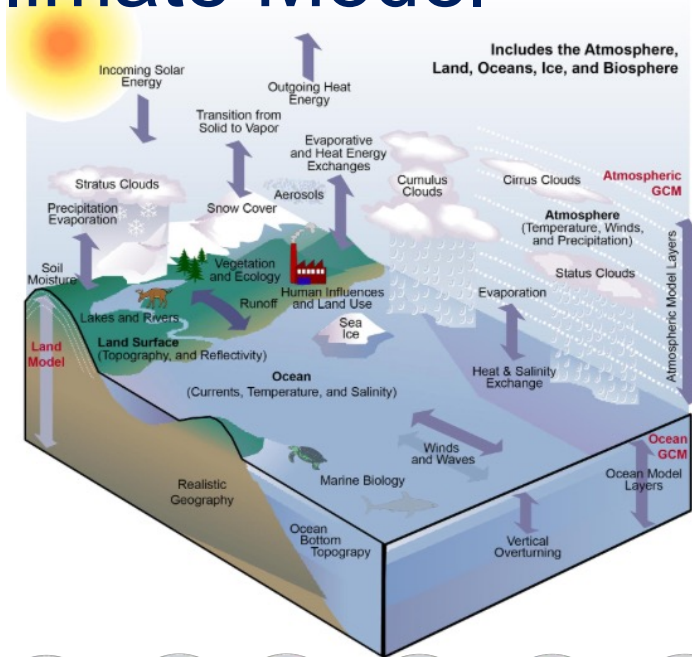
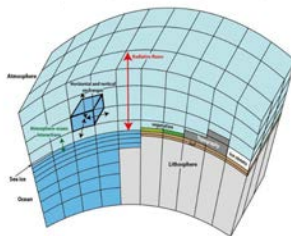
Conservation of H₂O (vapor, liquid, solid):

$$\frac{\partial q}{\partial t} = -(\vec{V} \cdot \nabla) q + \nabla \cdot (k \nabla q) + S_q + E$$

Equation of state:

$$p = \rho R_d T$$

Grid



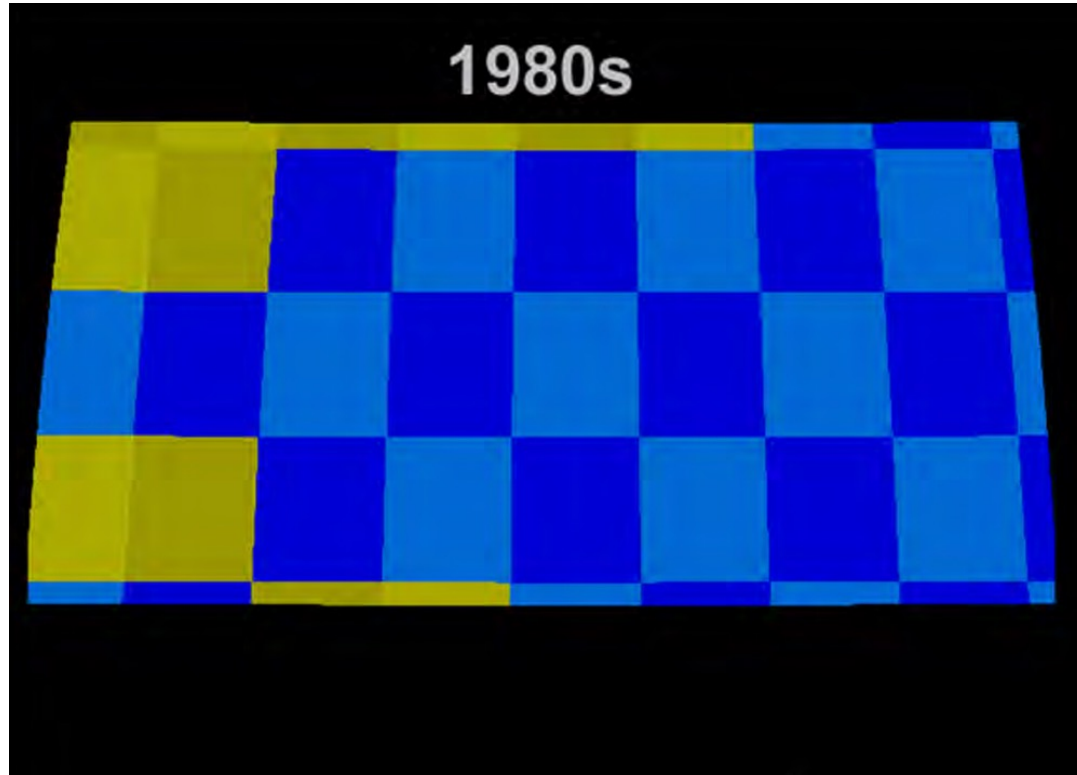
Components Interconnected to allow for Feedbacks

Parameterizations (differ by model)

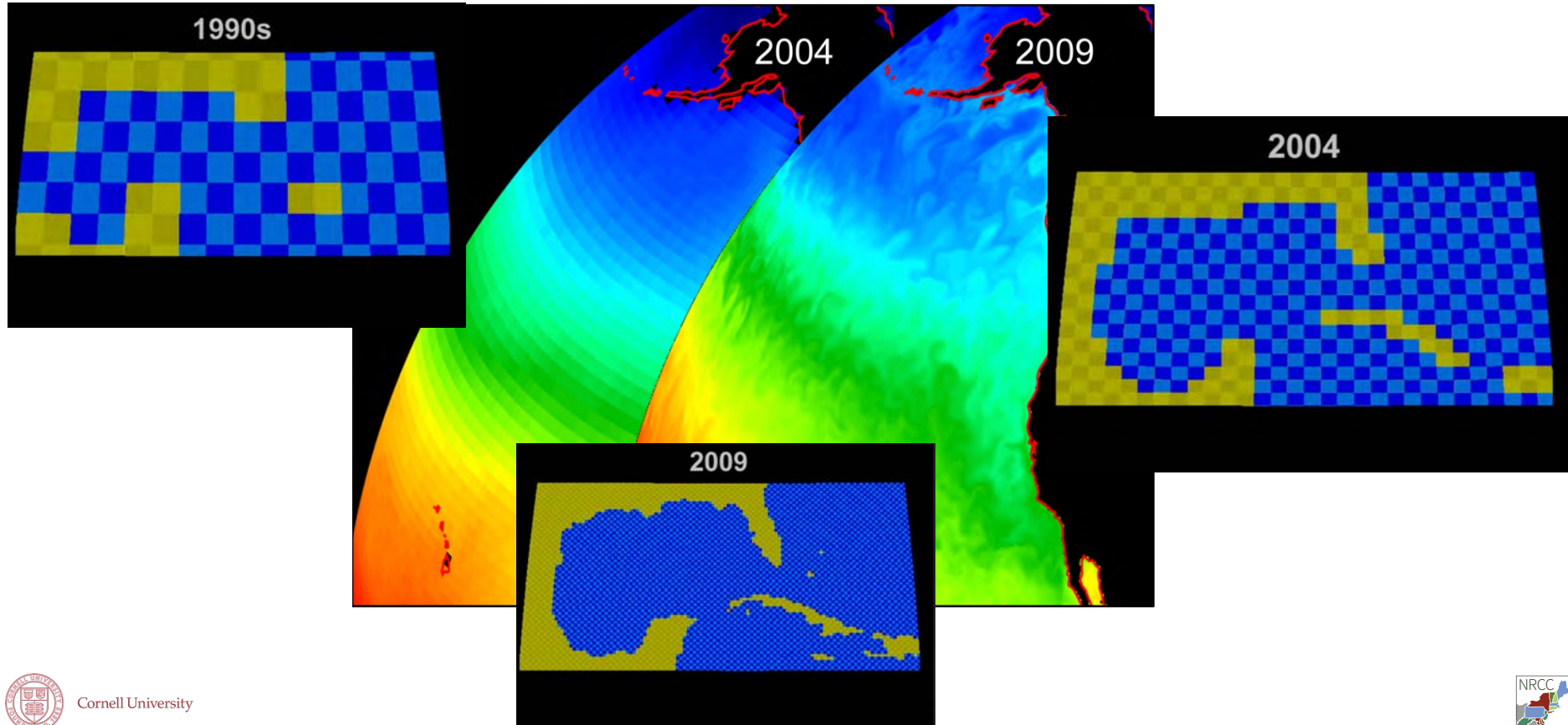


Energy Balance Models Atmosphere-Ocean General Circulation Models Earth System Models

Anatomy of a Climate Model



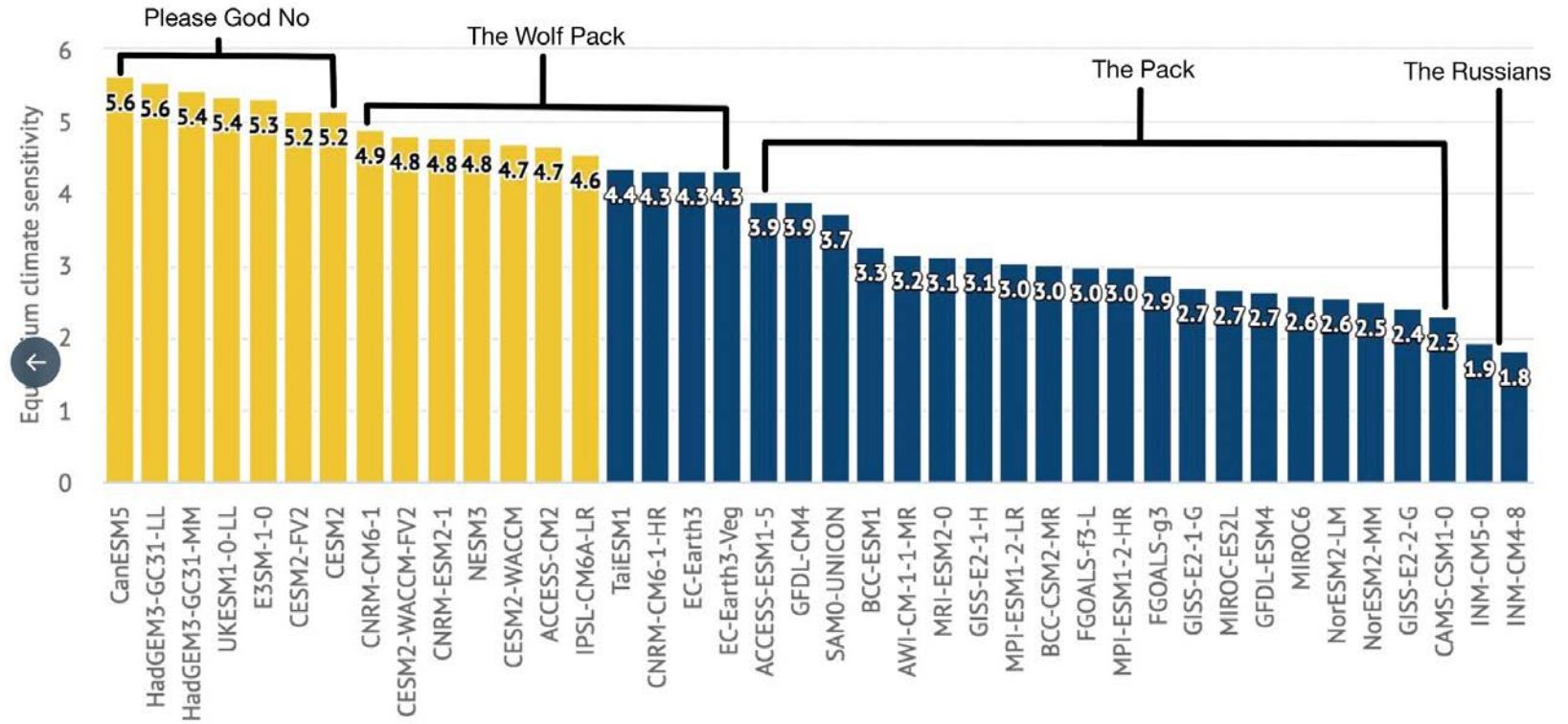
Anatomy of a Climate Model



amount of global warming over hundreds of years after a doubling of the atmospheric CO₂ concentration.

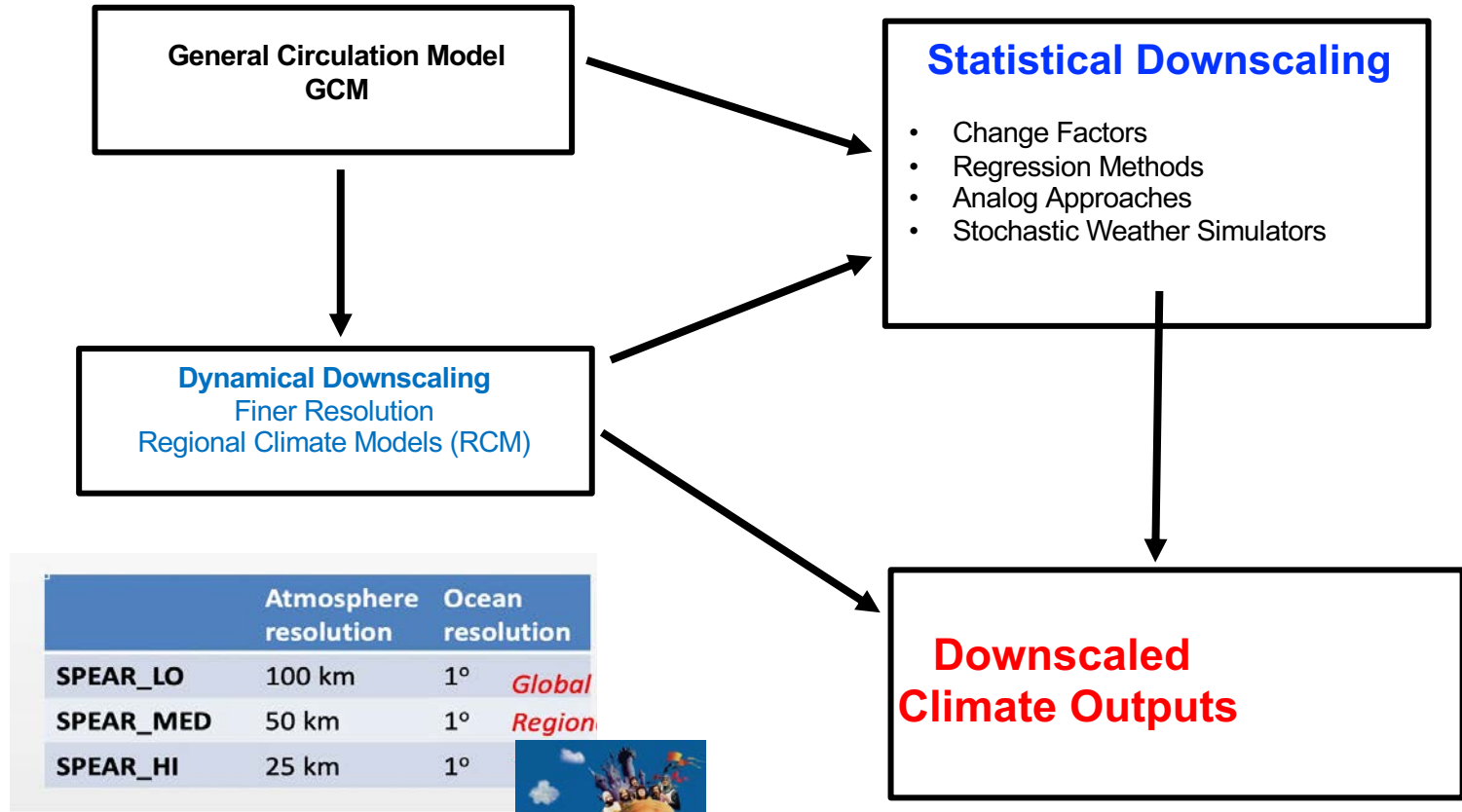
Anatomy of a Climate Model

Climate sensitivity in CMIP6 models



Source: Zeke Hausfather <https://twitter.com/hausfath>

Impacts Occur at Finer Scales than GCM Resolution

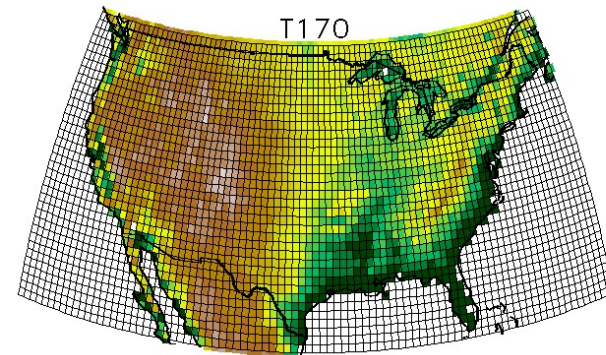
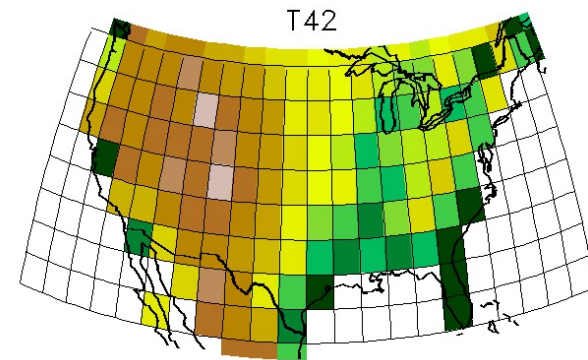


What is Downscaling?

GOALS WHEN PRODUCING STATISTICAL DOWNSCALED CLIMATE MODEL PROJECTIONS:

A refinement of dynamical model
results, informed by observations

- 1) Account for GCM biases relative to observations
- 2) Add spatial detail or localized info not present in coarse resolution GCMS
- 3) Do Not markedly distort the GCM's climate change signal





FAQ

Which is the best downscaling method?

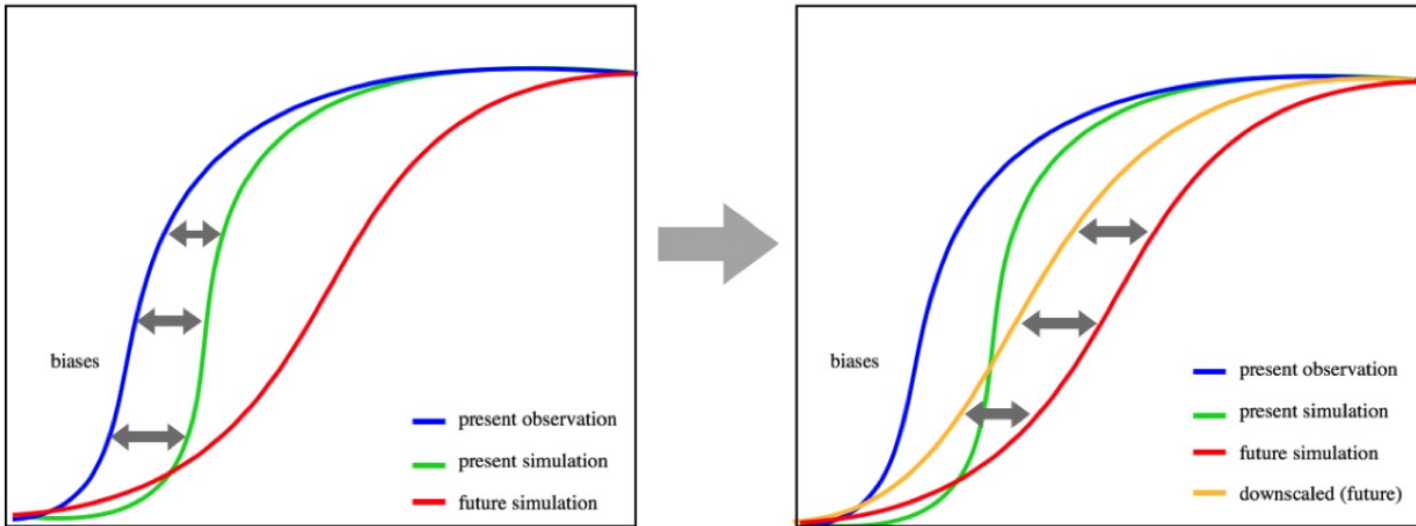
**A: It depends on several factors,
including what is the intended
end use (application).**

- Time & spatial scales of interest.
- Climate variables of interest.
- Sensitivity to central tendencies vs. extremes or spells.
- Whether ensembles are to be considered.

etc.

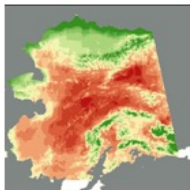


Delta Method Downscaling Bias Correction



Bias Correction and Spatial Downscaling

Observed Data Anomalies
High Spatial Resolution (1/8°)

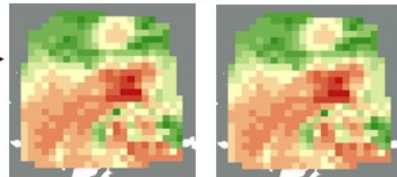


Aggregate to GCM Scale (1°)

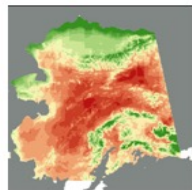


Compute function between Aggregated
Observed and GCM Historical Anomalies

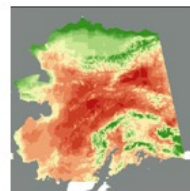
$$\text{Historical GCM} = a(\text{Observation}) + b$$



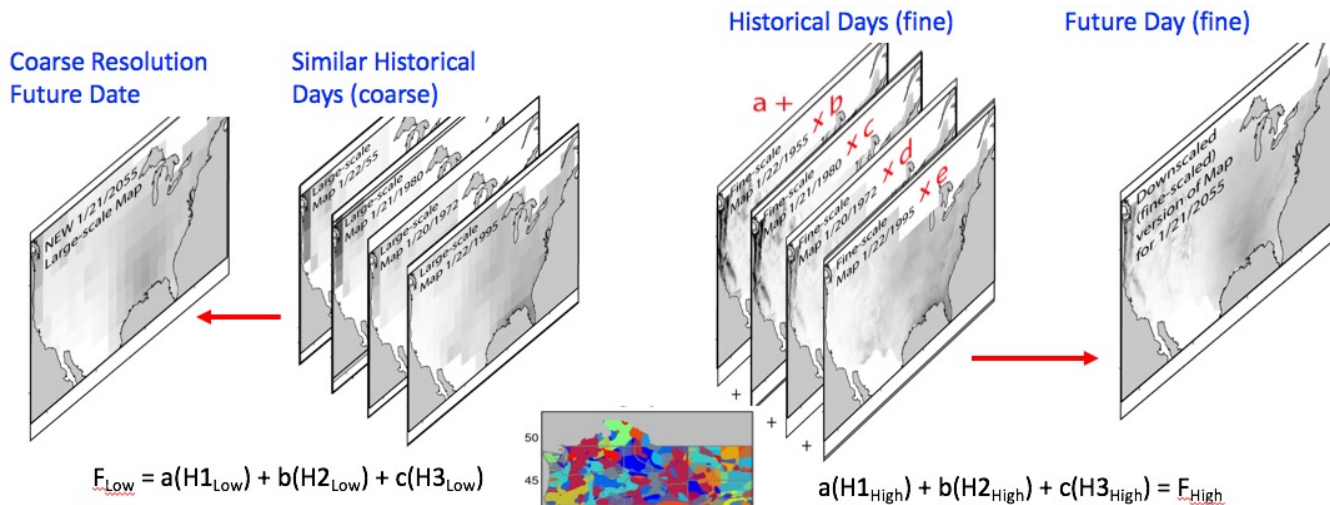
Apply function to Future GCM values



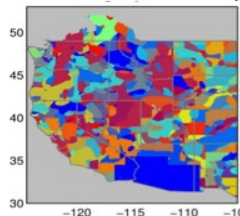
Interpolate factors to fine grid



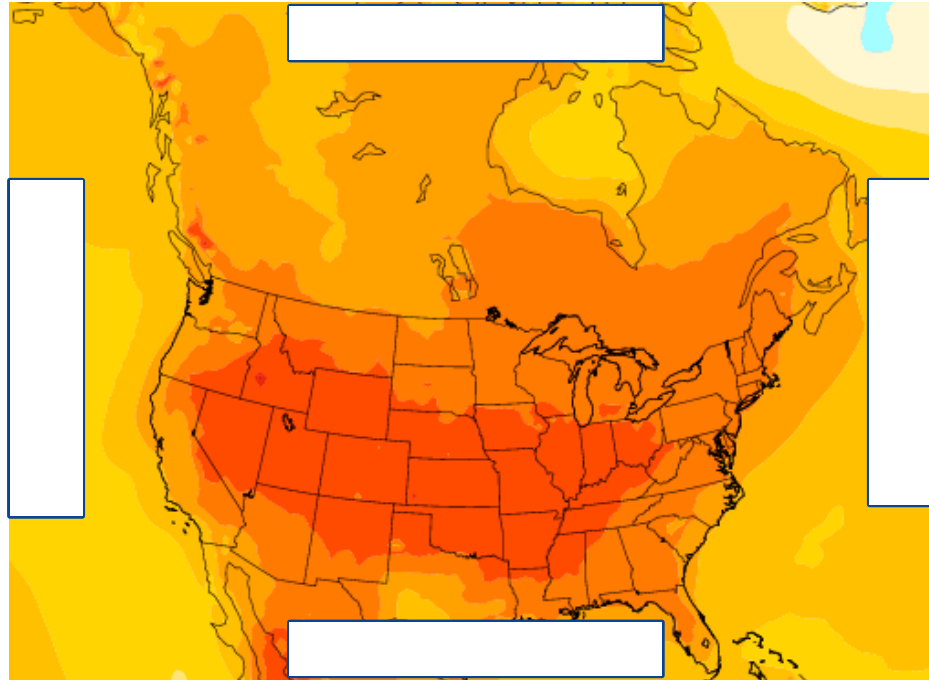
Localized Constructed Analogs



NATIONAL CENTERS FOR ENVIRONMENTAL INFORMATION

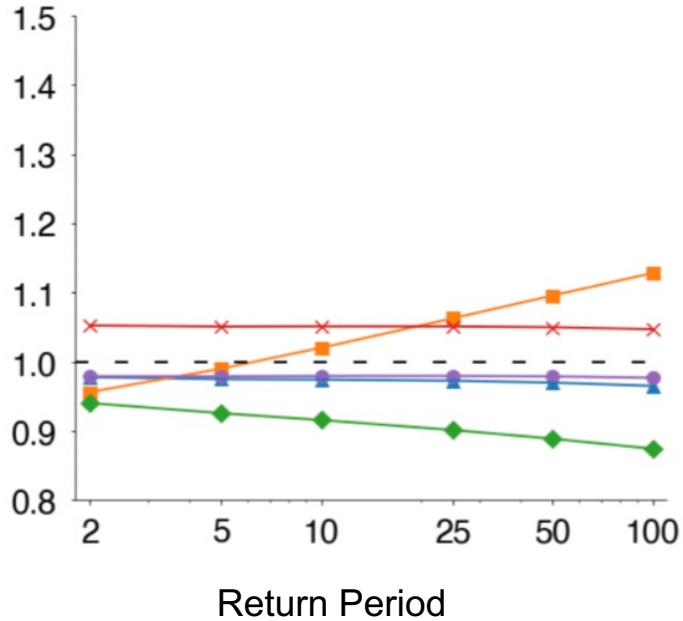


Dynamic Downscaling



Distort GCM signal

Downscaled Change Factor / CanESM native CF



Northeast



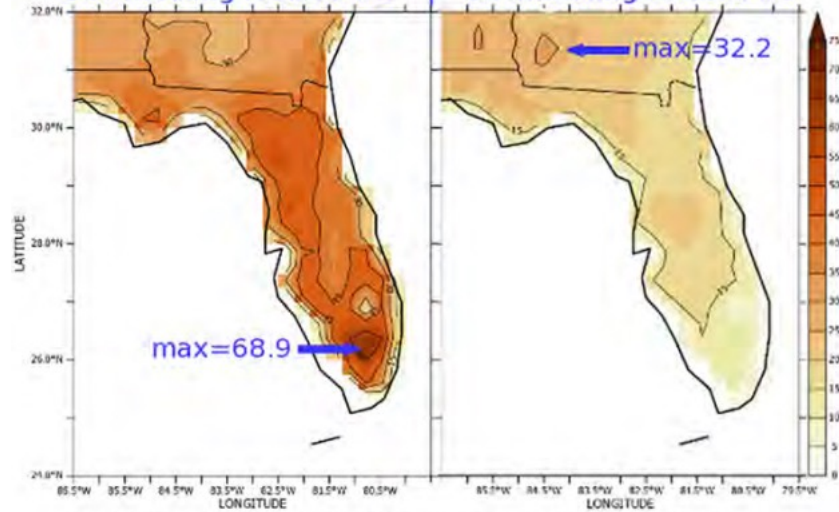
What Method is best?

Avg # of Days per Year >100F
(RCP8.5 scenario; 2086-2095)

Identical inputs expect for downscaling method

Downscaling Method #1

Downscaling Method #2

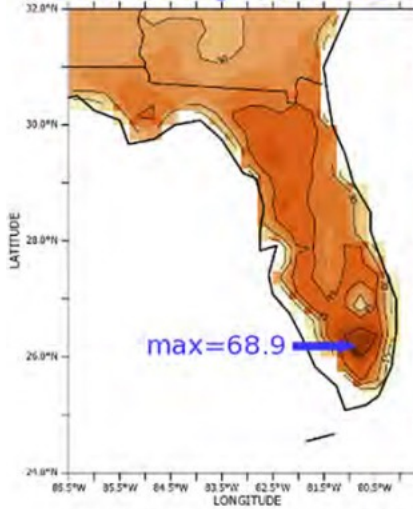


What Method is best?

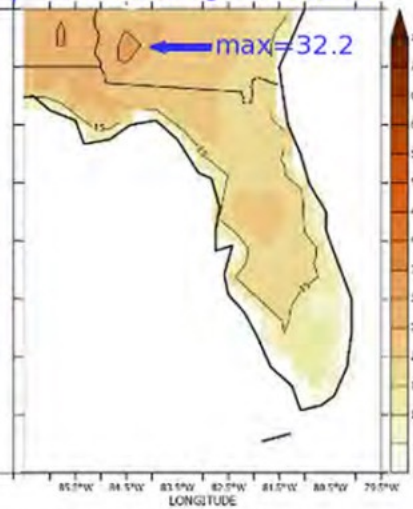
Avg # of Days per Year >100F
(RCP8.5 scenario; 2086-2095)

Identical inputs expect for downscaling method

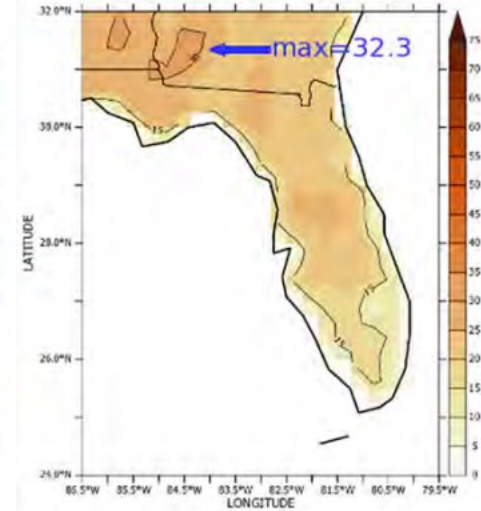
Downscaling Method #1



Downscaling Method #2



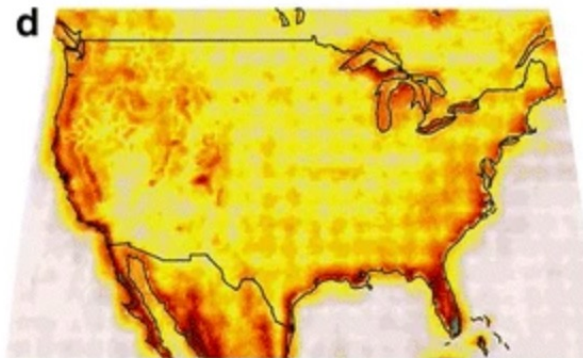
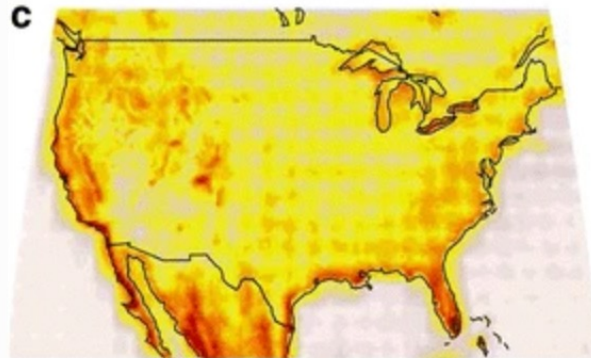
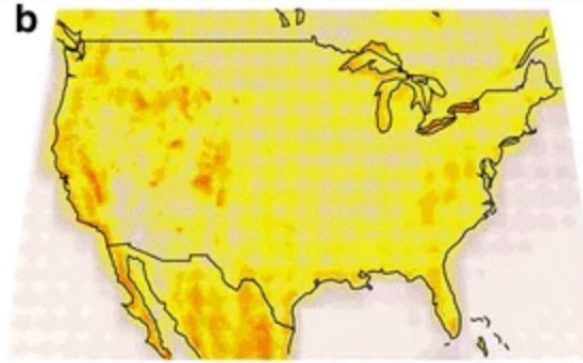
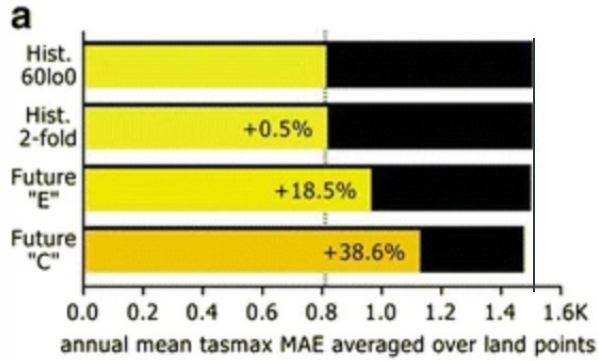
"TRUTH"



What Method is best?

Black bars
Coarsened –
high res (no
downscaling)

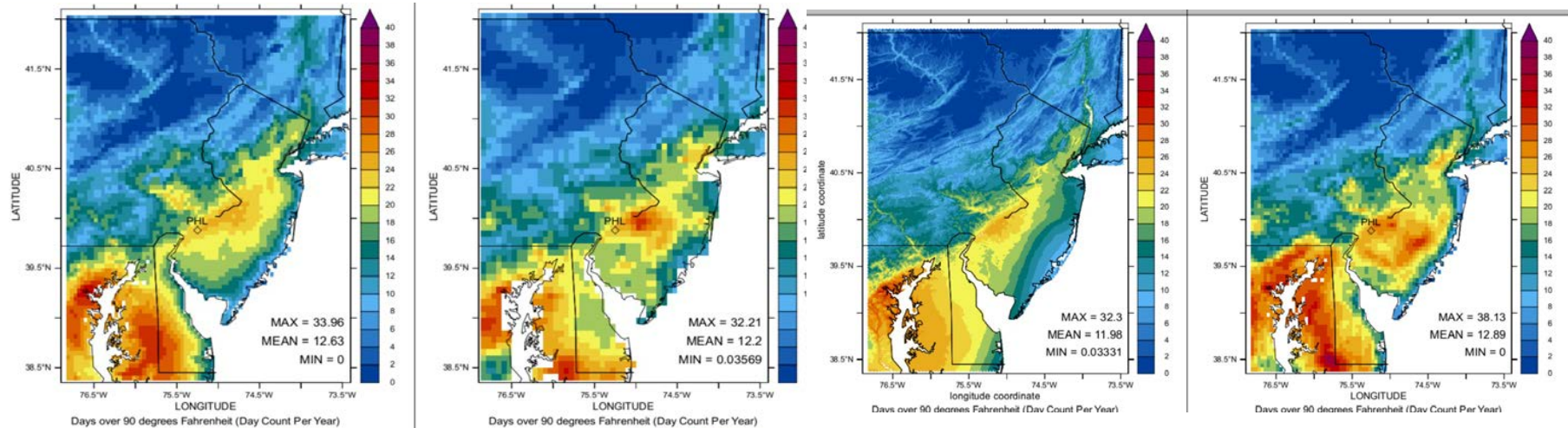
Yellow Bars
Downscaled
coarsened – high
res



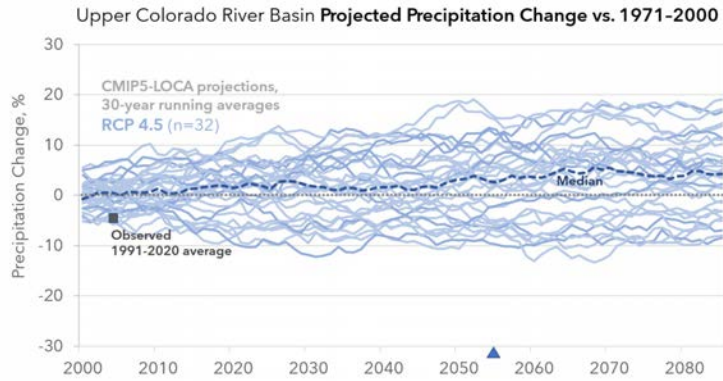
Dixon, K.W., Lanzante, J.R., Nath, M.J., Hayhoe, K., Stoner, A., Radhakrishnan, A., Balaji, V. and Gaitán, C.F., 2016. Evaluating the stationarity assumption in statistically downscaled climate projections: is past performance an indicator of future results?. *Climatic Change*, 135, pp.395-408.

Even Observed Data Set Matters!

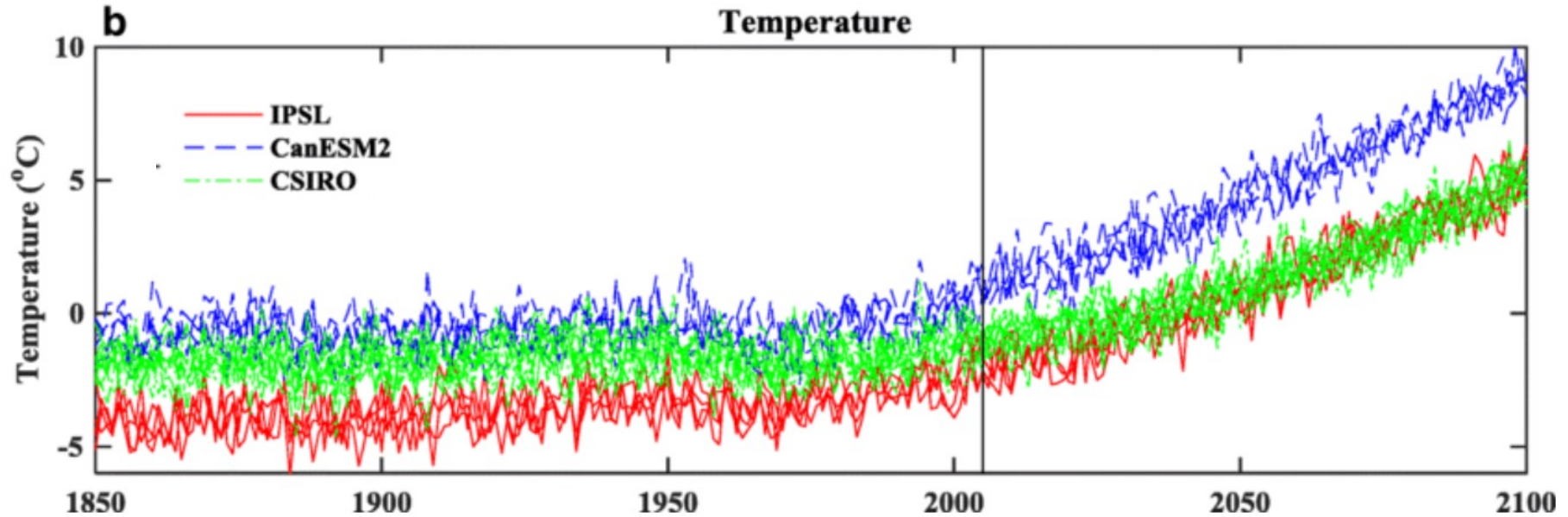
Annual Number of Days > 90



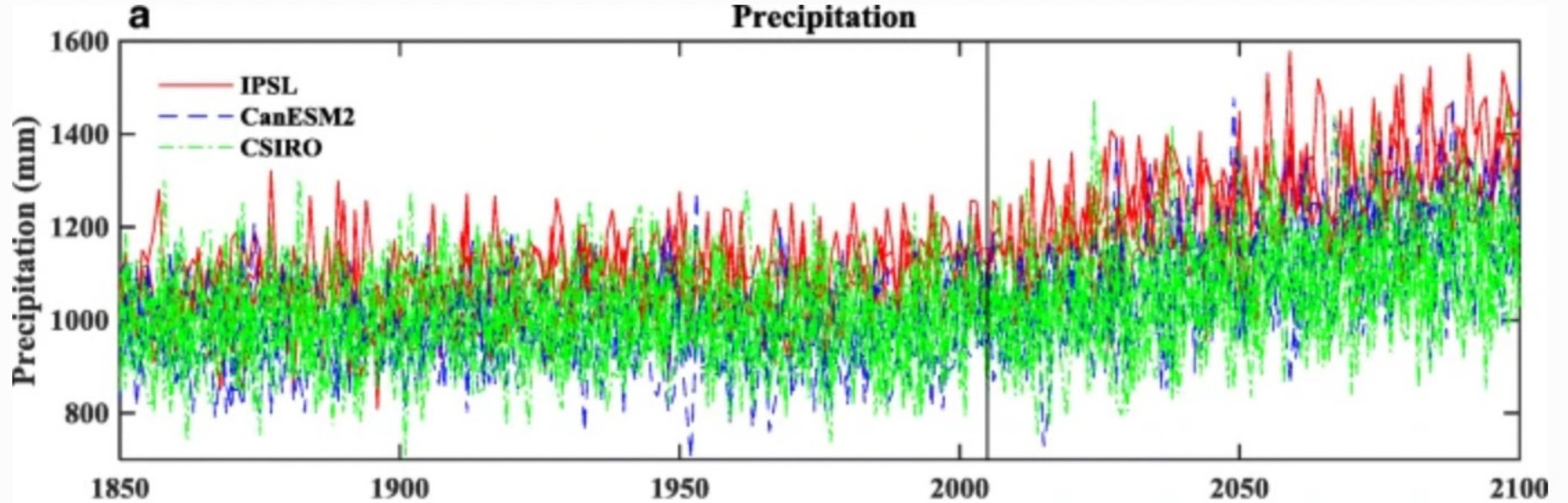
Multi Model Ensembles



Single Model Ensembles



Single Model Ensembles



CMIP5 vs CMIP6

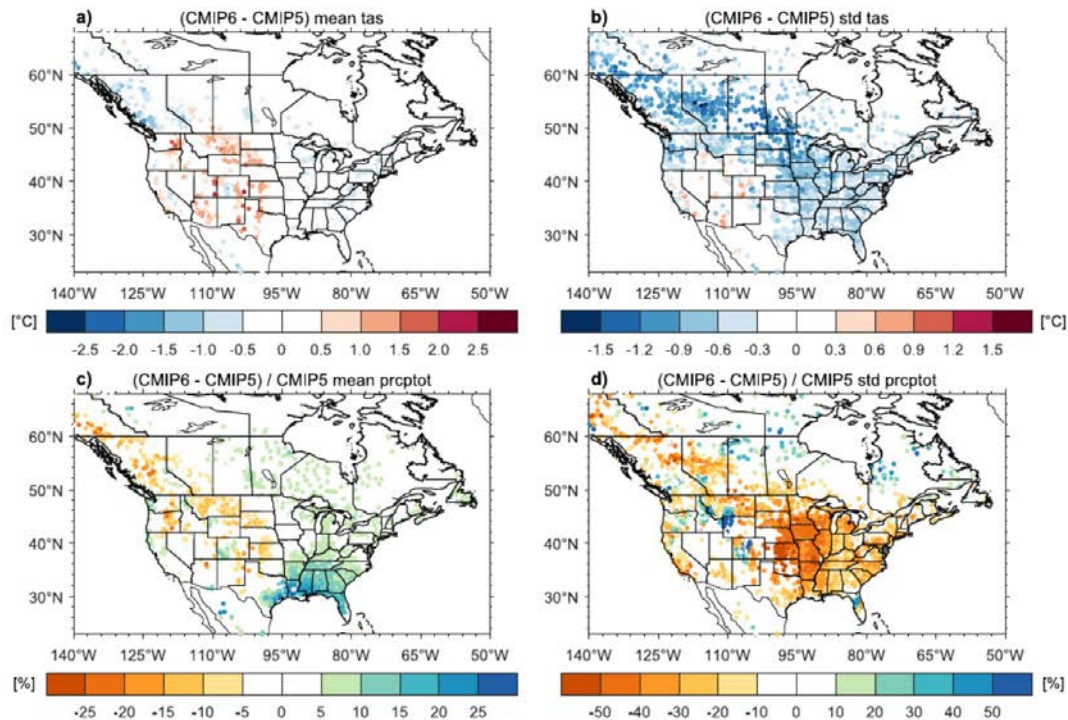


Figure 1. Differences between both raw Coupled Model Intercomparison Project Phase 6 (CMIP6) and Coupled Model Intercomparison Project version 5 (CMIP5) multimodel means (a and c) and inter-model standard deviation (b and d) over the 2070–2099 period for mean annual temperature (tas; a and b) and total precipitation (prcptot; c and d).

Martel, J.L., Brissette, F., Troin, M., Arsenault, R., Chen, J., Su, T. and Lucas-Picher, P., 2022. CMIP5 and CMIP6 model projection comparison for hydrological impacts over North America. *Geophysical Research Letters*, 49(15), p.e2022GL098364.

Questions?

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atd2@cornell.edu

