

Improving the Parameterization of Cloud and Rain Microphysics in E3SM using a Novel Observationally-Constrained Bayesian Approach

Marcus van Lier-Walqui¹, Hugh Morrison², Greg
Elsaesser¹, Peter Caldwell³, Po-Lun Ma⁴

¹CCSR Columbia University and NASA GISS

²National Center for Atmospheric Research

³Lawrence Livermore National Laboratory

⁴Pacific Northwestern National Laboratory



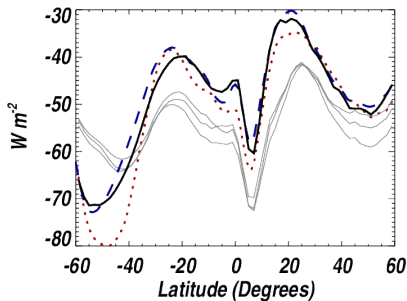
 COLUMBIA UNIVERSITY
IN THE CITY OF NEW YORK

October 27 2020

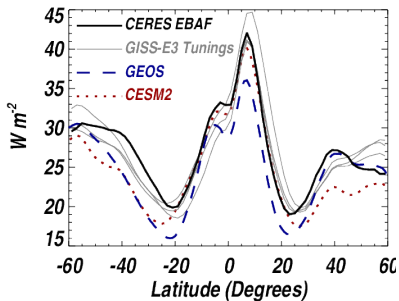
Motivation

There is considerable disagreement among climate models

SW Cloud Radiative Effect

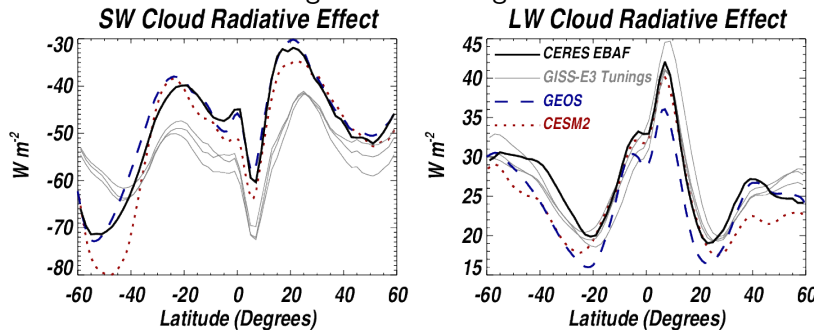


LW Cloud Radiative Effect



Motivation

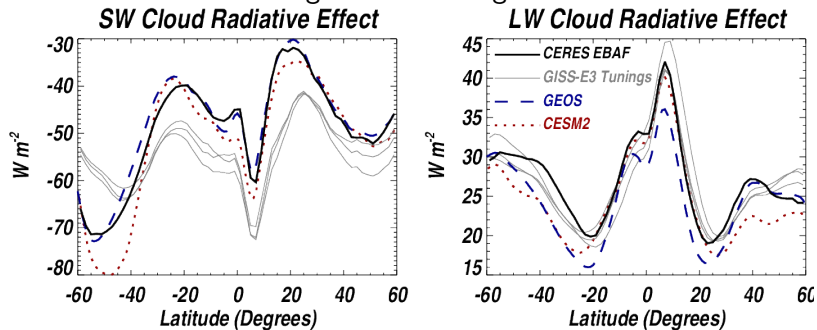
There is considerable disagreement among climate models



Codes that simulate unresolved processes, called *parameterizations*, are a leading source of uncertainty that results in bias

Motivation

There is considerable disagreement among climate models

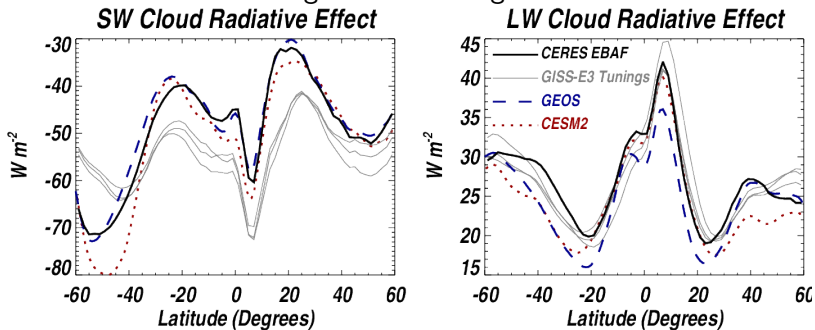


Codes that simulate unresolved processes, called *parameterizations*, are a leading source of uncertainty that results in bias

E.g. clouds, cloud microphysics, radiation, land surface, etc.

Motivation

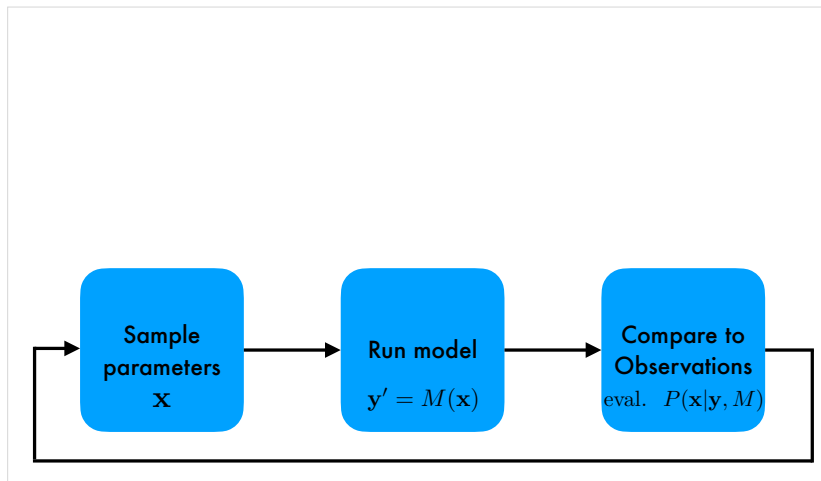
There is considerable disagreement among climate models



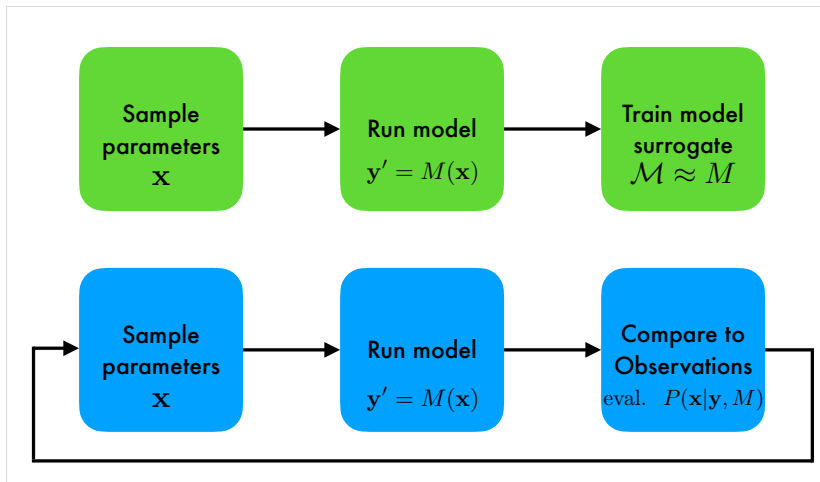
Codes that simulate unresolved processes, called *parameterizations*, are a leading source of uncertainty that results in bias

E.g. clouds, [cloud microphysics](#), radiation, land surface, etc.

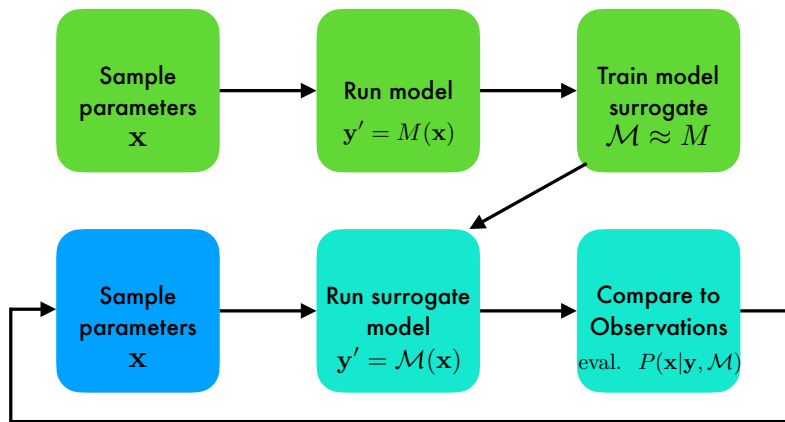
Big Picture 1/3: Surrogate-accelerated inference



Big Picture 1/3: Surrogate-accelerated inference



Big Picture 1/3: Surrogate-accelerated inference



Big Picture 2/3: Unite “bottom-up” and “top-down”

Big Picture 2/3: Unite “bottom-up” and “top-down”

Perform bottom-up constraint (with UQ!) using “reference” schemes and detailed case data (e.g. within LES or SCMs)

Big Picture 2/3: Unite “bottom-up” and “top-down”

Perform bottom-up constraint (with UQ!) using “reference” schemes and detailed case data (e.g. within LES or SCMs)

Bottom-up insights are probabilistic, they can be used as *prior distributions* to be further refined by “top-down” global satellite constraint

Big Picture 2/3: Unite “bottom-up” and “top-down”

Perform bottom-up constraint (with UQ!) using “reference” schemes and detailed case data (e.g. within LES or SCMs)

Bottom-up insights are probabilistic, they can be used as *prior distributions* to be further refined by “top-down” global satellite constraint

Final result is probabilistic model tuning that is consistent with detailed process-level knowledge *and* emergent global constraints

Big Picture 2/3: Unite “bottom-up” and “top-down”

Perform bottom-up constraint (with UQ!) using “reference” schemes and detailed case data (e.g. within LES or SCMs)

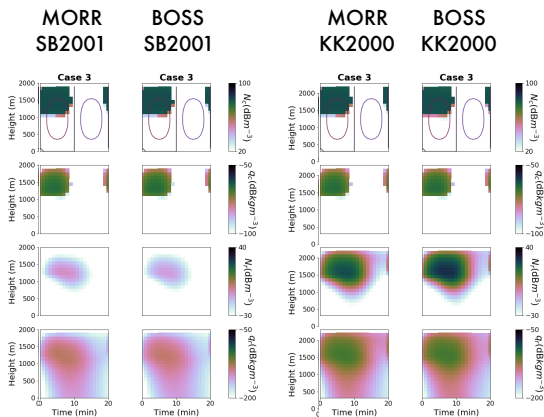
Bottom-up insights are probabilistic, they can be used as *prior distributions* to be further refined by “top-down” global satellite constraint

Final result is probabilistic model tuning that is consistent with detailed process-level knowledge *and* emergent global constraints

This process typically only considers *parameter* estimation — what about *both* structural uncertainties?

Big picture 3/3: Address structural errors using BOSS

The Bayesian Observationally-constrained Statistical-physical Scheme (BOSS) was developed to provide a structurally flexible warm microphysics parameterization scheme (Morrison et al. JAS 2020, van Lier-Walqui et al. JAS 2020)



Big Picture: E3SM BOSS PPE

Surrogate-accelerated PPE inference

Run a large PPE used to train a *fast* surrogate model, perform inference on the fast model using global satellite observations.

Unify bottom-up and top-down constraints

Probabilistically unify detailed bottom-up process-level insights with emergent global constraints from satellite observations.

Address structural uncertainties in microphysics

Use BOSS to relax structural inflexibilities in (MG2/P3) microphysics, move beyond the limitations of the state-of-the-art schemes (Morrison et al. JAMES 2020)

We're looking for a postdoc! Email me mv2525@columbia.edu

Resolving structural uncertainty

Sources of model uncertainty:

- Structural uncertainty: *don't know the equations*
- Initial/boundary condition uncertainty: *don't know the initial state and/or forcing*
- Numerical uncertainty: *don't know how to solve the equations*
- Parametric uncertainty: *don't know the parameter values in the equations*

Using BOSS to address uncertainty

BOSS can systematically explore and quantify *structural uncertainty* via model expansion and parsimony

Using BOSS to address uncertainty

BOSS can systematically explore and quantify *structural uncertainty* via model expansion and parsimony

Model Expansion

systematic addition of complexity to a model equation set that covers all possible formulations (e.g., polynomials for smooth functions)

Using BOSS to address uncertainty

BOSS can systematically explore and quantify *structural uncertainty* via model expansion and parsimony

Model Expansion

systematic addition of complexity to a model equation set that covers all possible formulations (e.g., polynomials for smooth functions)

Parsimony

the idea that a model should only be as complex as needed to adequately represent the phenomenon being simulated, in the absence of theoretical knowledge (i.e., avoid over-fitting)

Overly complicated schematic

