# Physical Process Realism and Model Hierarchies

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Idealized Models (Aquaplanets, RCE)

#### **Process Models**

(DNS, LES, CRMs for radiative transfer, turbulence, boundary layers, convection, cloud microphysics & macrophysics)



Reduced Set of Physical Processes, Simplified BCs

Idealized Models (Aquaplanets, RCE)

#### **Process Models**

(DNS, LES, CRMs for radiative transfer, turbulence, boundary layers, convection, cloud microphysics & macrophysics)



Physical Basis from which **Complexity Emerges** 

#### **Process Models**

(DNS, LES, CRMs for radiative transfer, turbulence, boundary layers, convection, cloud microphysics & macrophysics)

**Idealized Models** 

(Aquaplanets, RCE)

#### **Evaluation and Metrics**

Complex Models (ESMs, NWP)

Simulated
Large-Scale Fields,
Means & Variability

Idealized Models (Aquaplanets, RCE)

#### **Process Models**

(DNS, LES, CRMs for radiative transfer, turbulence, boundary layers, convection, cloud microphysics & macrophysics)

#### **Evaluation and Metrics**

Complex Models (ESMs, NWP)

Idealized Models (Aquaplanets, RCE)

#### **Process Models**

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Conceptual Understanding, Interpretations of Complex Simulations

# Evaluation and Metrics lex Models

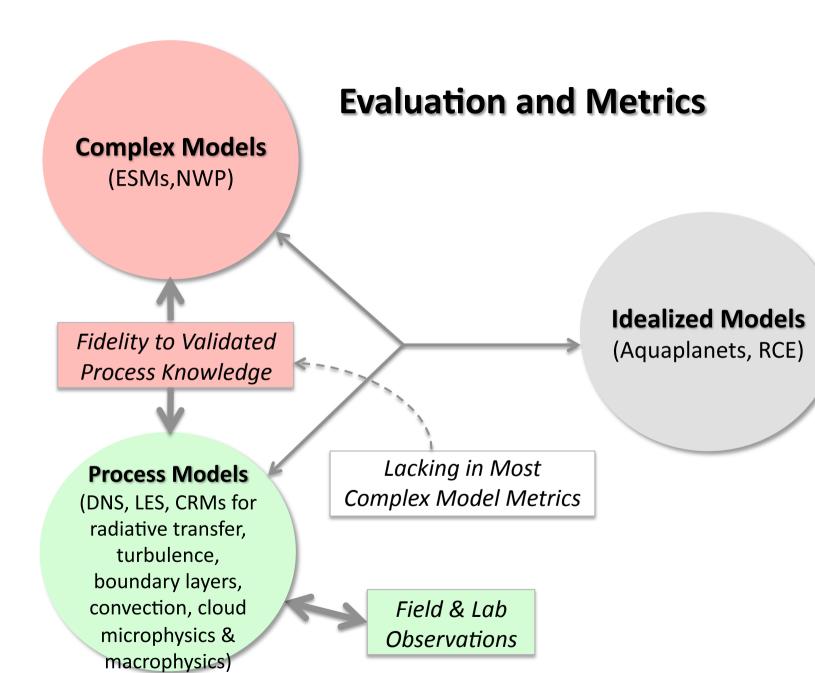
Complex Models (ESMs, NWP)

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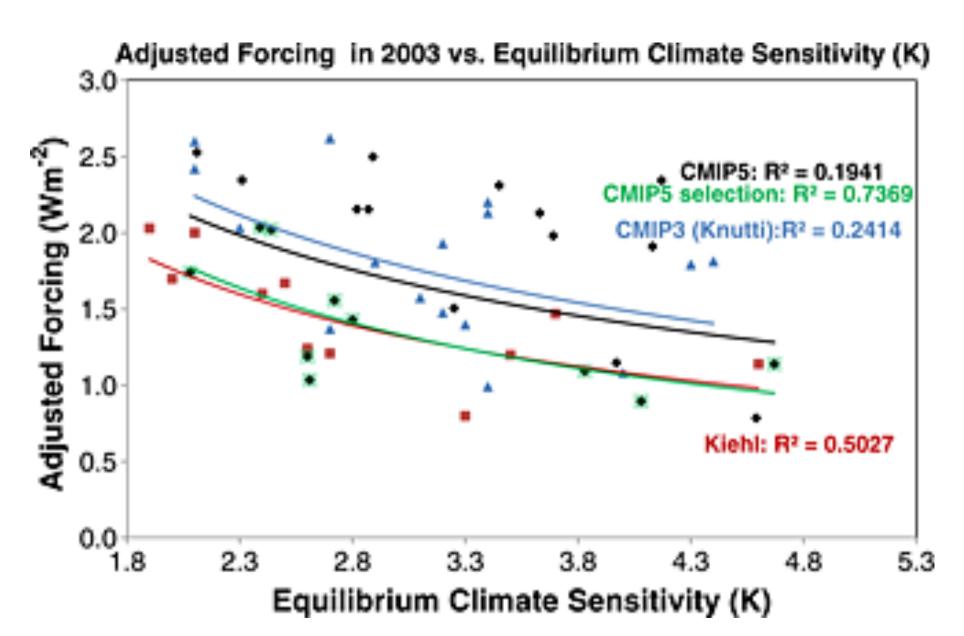
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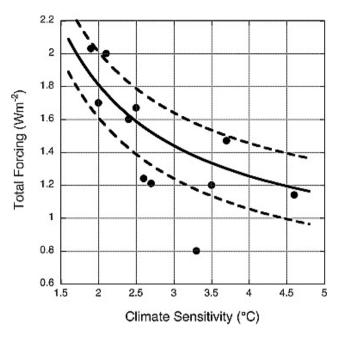
Field & Lab
Observations



Countervailing Metrics: The Example of an Emergent System Property (Historical Global Temperature Change) versus a Fundamental Process Property (Rain Production Rate)



From Forster et al. in **Journal of Geophysical Research: Atmospheres**<u>Volume 118, Issue 3, pages 1139-1150, 6 FEB 2013 DOI: 10.1002/jgrd.50174</u>
<a href="http://onlinelibrary.wiley.com/doi/10.1002/jgrd.50174/full#jgrd50174-fig-0007">http://onlinelibrary.wiley.com/doi/10.1002/jgrd.50174/full#jgrd50174-fig-0007</a>

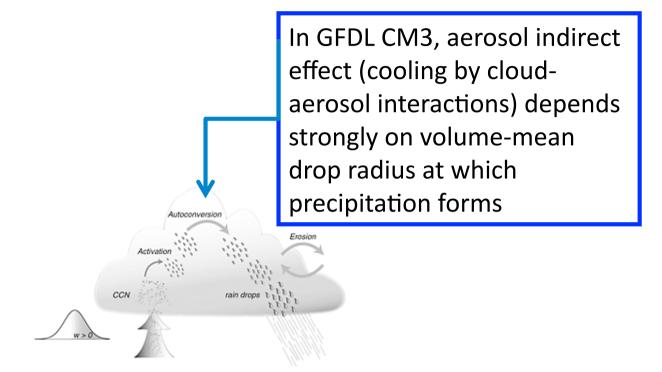


Kiehl (2007, Geophys. Res. Lett.)

How did the 20<sup>th</sup>
Century warm? High
forcing/low
sensitivity or low
forcing/high
sensitivity? Why is it
important?

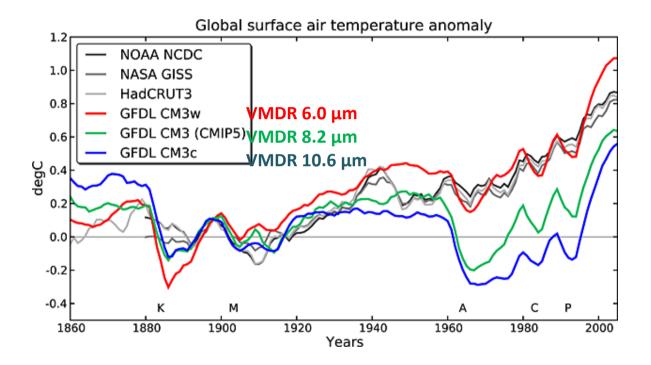
Future climate change will be driven more by greenhouse gases than aerosols, as aerosols have shorter lifetime than dominant anthropogenic greenhouse gases and aerosols likely to be regulated by air-pollution policy. "Masking" by aerosols will be less. Projecting warming requires knowledge of sensitivity.

#### Cloud tuning in a coupled climate model: Impact on 20th century warming



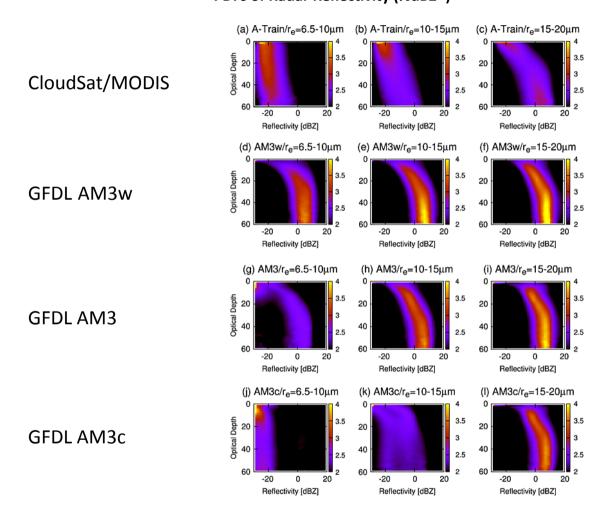
Observations show volume-mean drop radius for precipitation formation around 10-12 µm (Gerber, 1996, *J. Atmos. Sci.;* Pawlowska and Brenguier, 2003, *J. Geophys. Res.;* Boers *et al.*, 2006, *QJRMS*)

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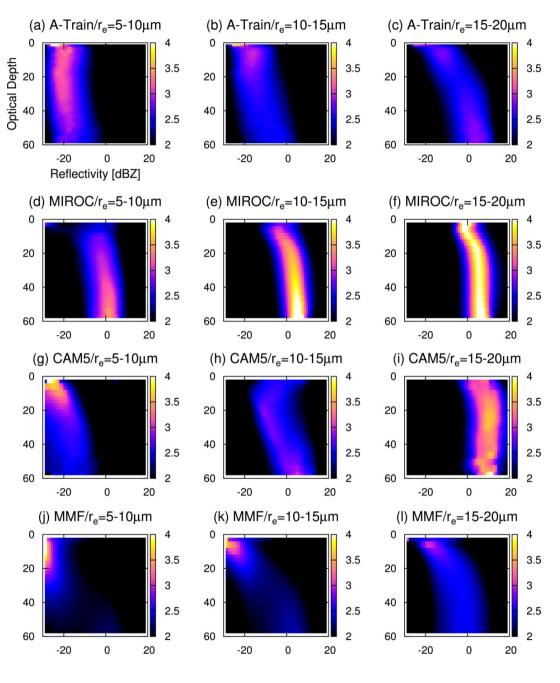


Observed volume-mean drop radius for precipitation initiation (VMDR) 10-12  $\mu m$ 

## Evaluating cloud tuning in a climate model with satellite observations PDFs of Radar Reflectivity (%dBZ-1)



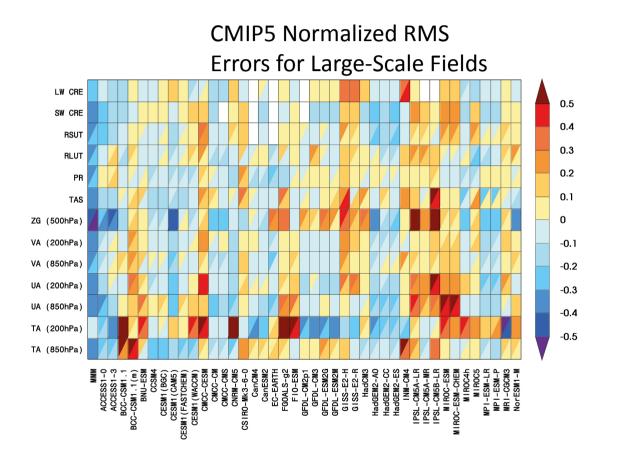
This diagnostic can be used with many formulations for cloud microphysics. **Summary** statistics (bias, RMSE, correlation coefficients against A-Train provide process metrics for climate models.



PDFs of Radar Reflectivity (%dBZ<sup>-1</sup>)

Figure 3 from Suzuki et al 2015: Evaluation of the Warm Rain Formation Process in Global Models with Satellite Observations. J. Atmos. Sci., 72, 3996–4014, doi: 10.1175/JAS-D-14-0265.1.

# Portrait Diagrams for Large-Scale Fields and Process-Level Metrics



A New Portrait

Diagram Based on

Process Evaluation?

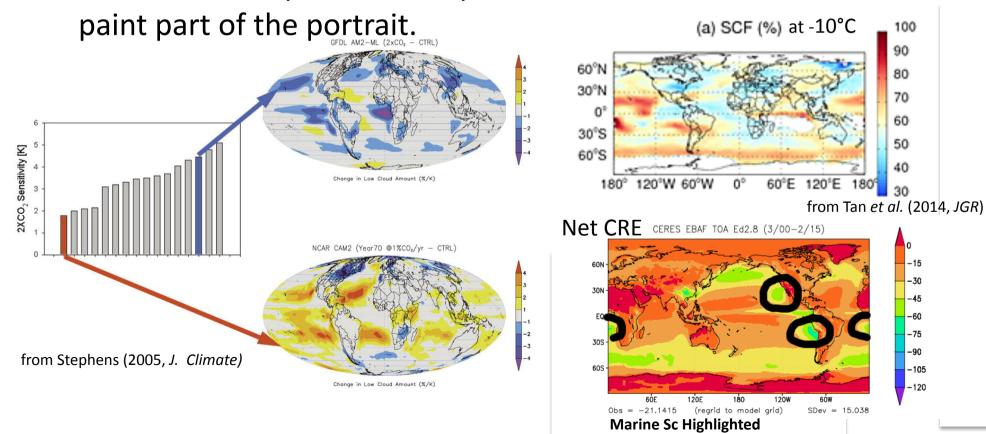
Example:

RMS Errors, Bias,
Correlation
Coefficients for A-Train
vs. Models'
Reflectivity/Optical
Depth PDFs

IPCC AR5 Fig. 9-7 (2013, Cambridge University Press)

## Building a Portrait for Evaluating Physical Processes in Climate Models

Uncertainties in climate sensitivity related to cloud feedbacks are apparent in distinct cloud systems that can be evaluated with LES and CRMs, in turn evaluated against observations. GCM/SCM comparison with process-based skill scores can



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- Once identified, these controls should be evaluated against process models and observations.
- Process realism should join means and variability among the evaluation metrics for complex models.