

# Model Output and Evaluation

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# Outline

1. How to evaluate a model with observations (ideas)
  - What is a measurement?
  - Comparing observations to models (apples and oranges).
  - Satellite simulators (COSP): Concept and implementation in CAM.
2. Diagnostic packages (show examples)
  - Mean: AMWG
  - Climate Variability: CVDP
  - Variability/Processes: MDTF
3. Model output
  - 'Standard output': how to get it out of the model
  - Look in the user guide: 'fancy' output (sathist, single point, regional).
  - How to output a new variable (needs to be a cam interface level, copy existing)

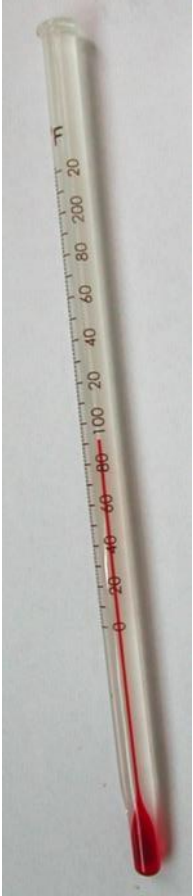
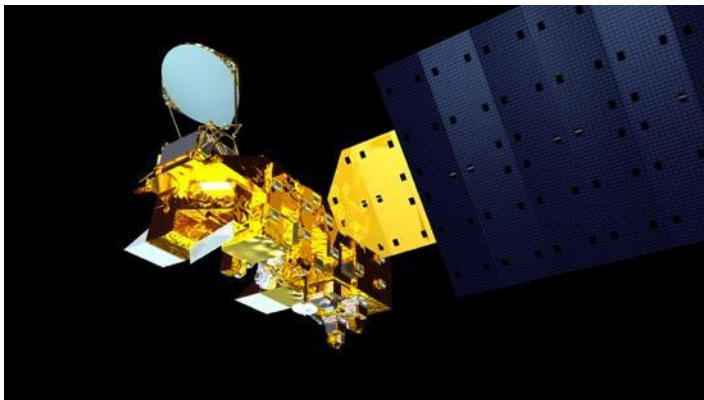


**Observations: What is a cloud?  
How do we compare models and observations?**

# Example: Measuring temperature

What do the following REALLY measure?

- Liquid (alcohol, red) thermometer?
- Thermocouple (digital thermometer)
- Infrared camera
- Satellite infrared imager



# What is the error in a temperature measurement

- Image: ocean surface with a ship.
  - Liquid Thermometer
  - Thermocouple
  - Infrared camera
  - Satellite instrument
- What is the error? What are they measuring?

# Measuring the 'Same' Temperature in a Scene

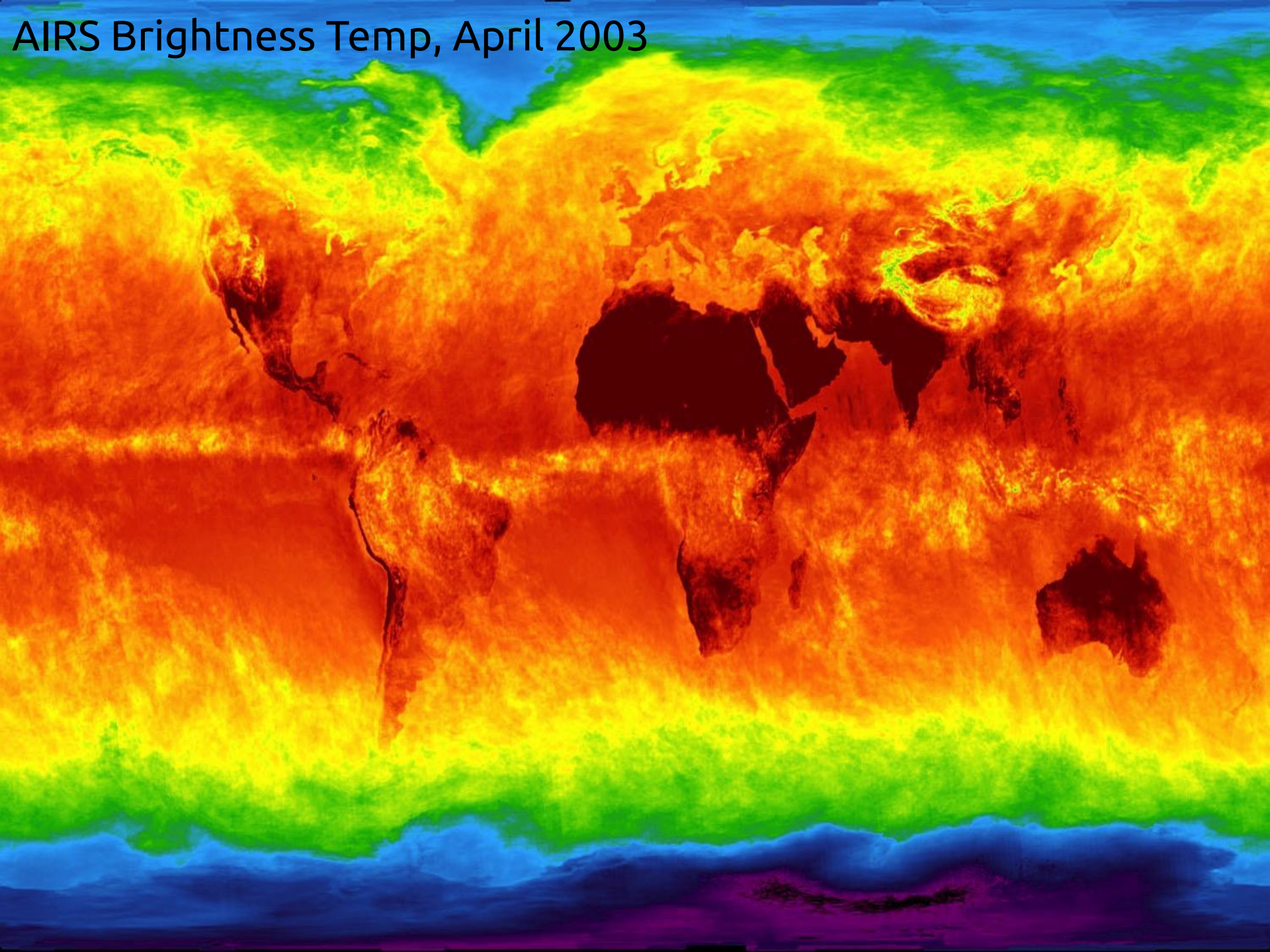
What is the error in the measurement? Uncertainty?



# Issues

- Instrument bias (the instrument model is wrong)
- Instrument noise (unbiased, but scatter)
- Sampling volume: what is being compared.
- Bottom line: mostly we have apples and oranges. As long as we are still dealing with fruit and we know this, maybe okay.
- Almost all observations have a model embedded in them!

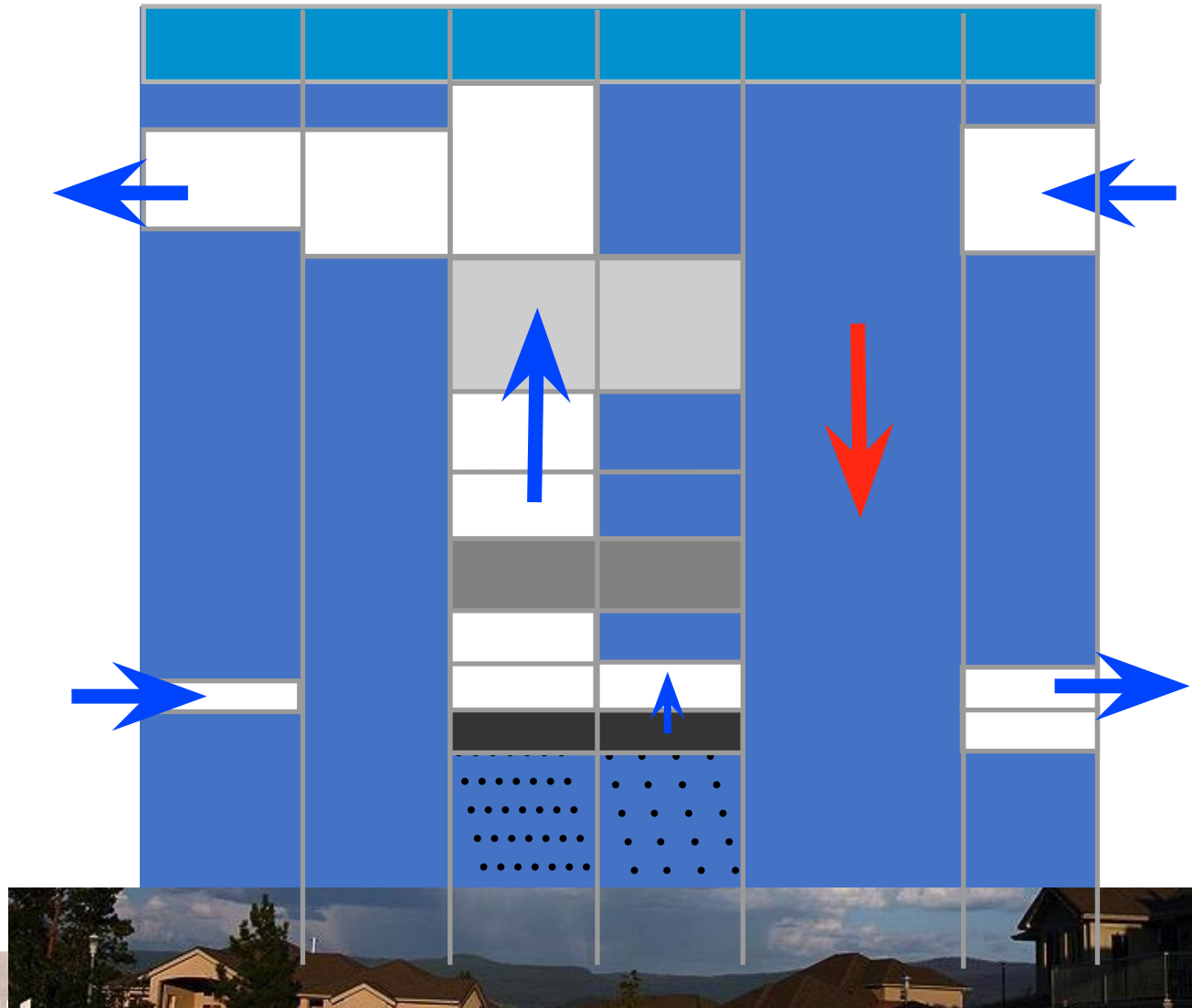
AIRS Brightness Temp, April 2003



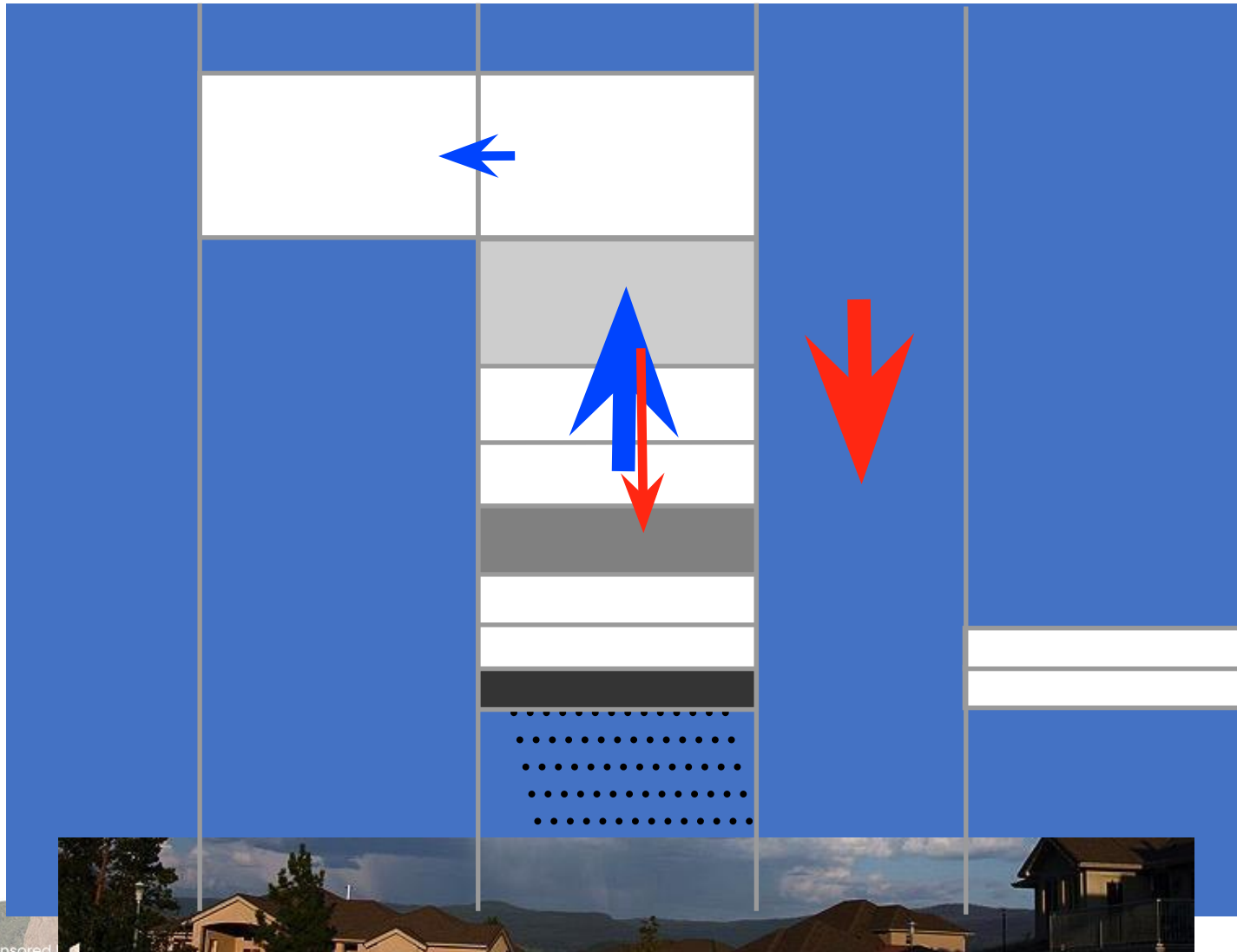
# How does a model see clouds?



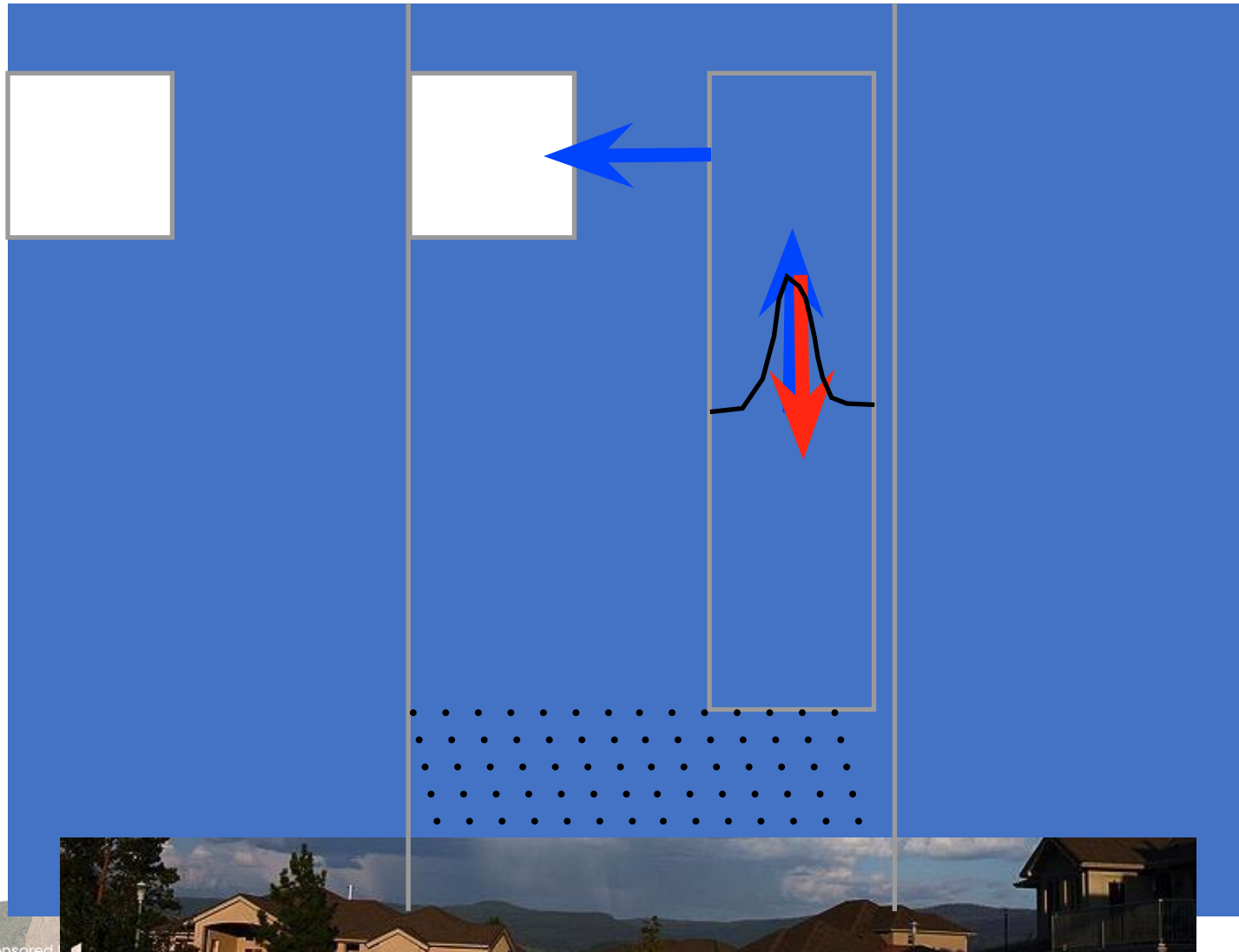
# CRM/LES Convective Cloud



# Mesoscale Convective Cloud



# GCM Convective Cloud

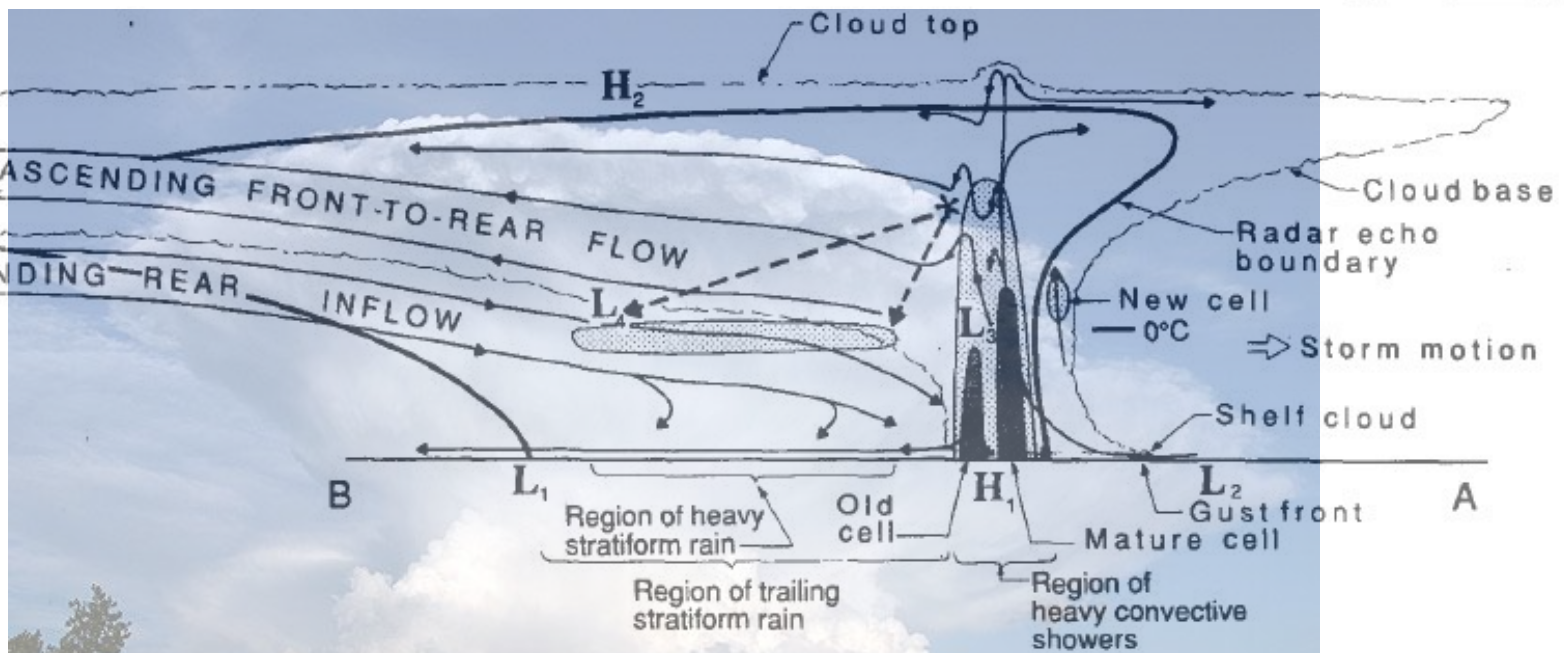


# This is a (picture of a) Cloud

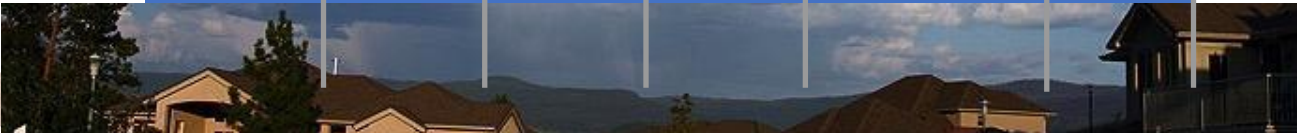
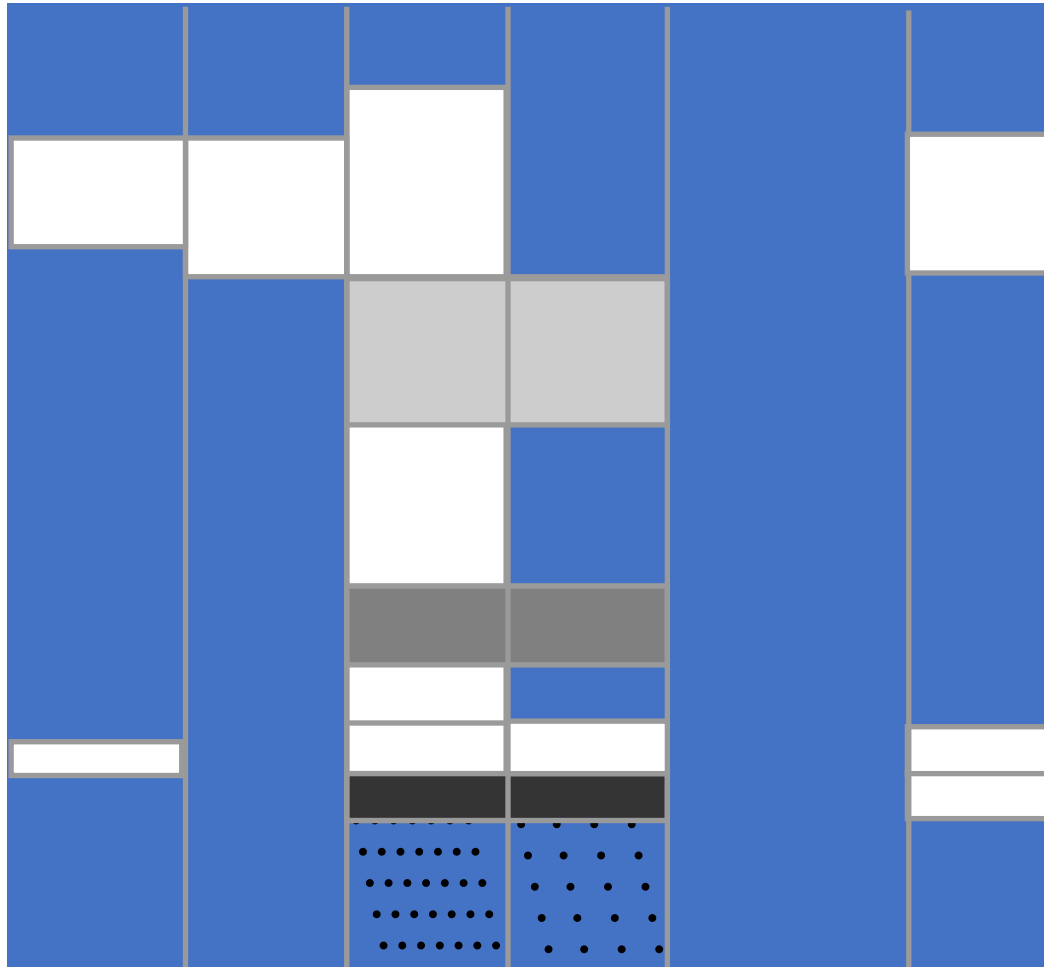
How do we observe  
it?



# Conceptual Picture



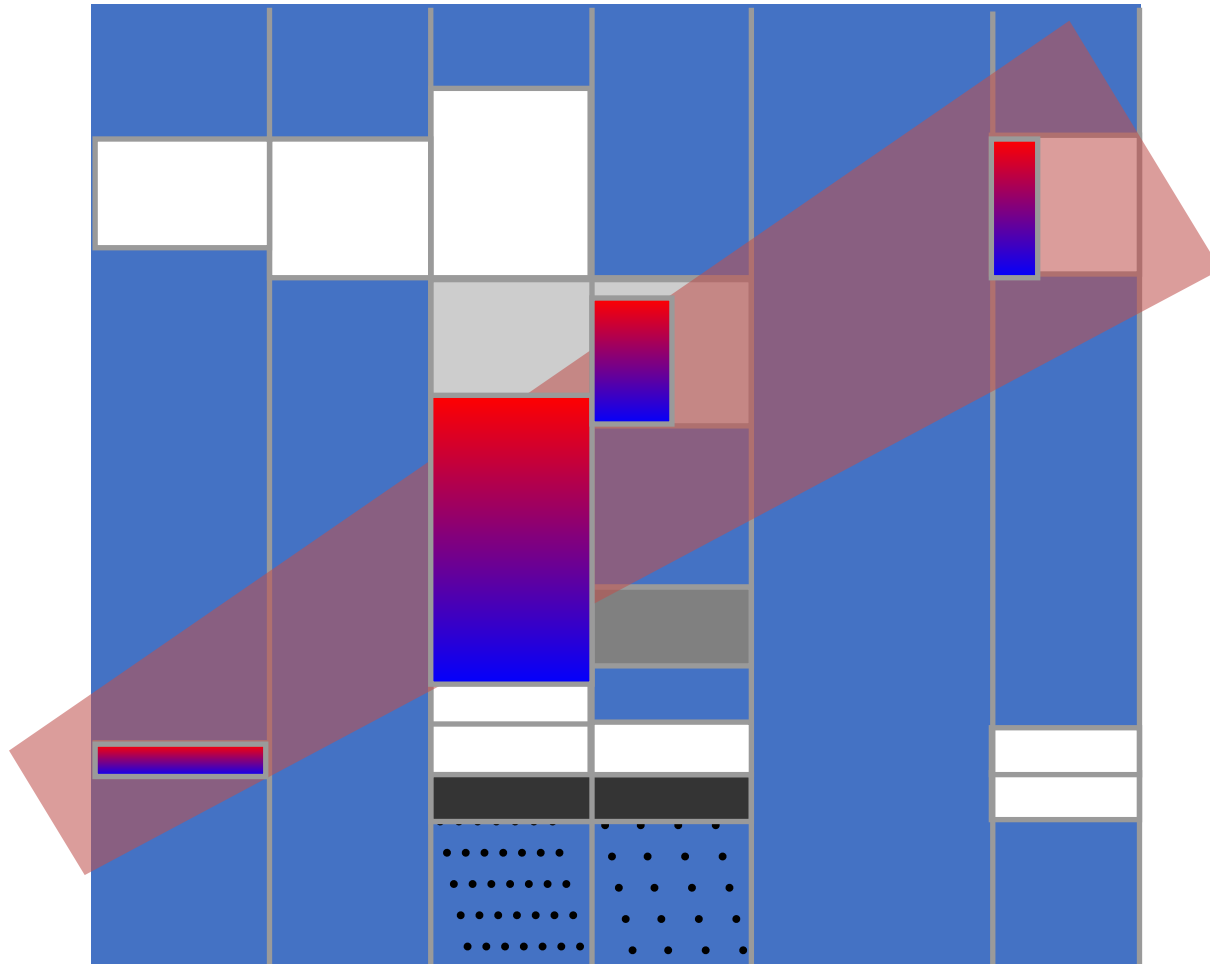
# Convective Cloud (=Truth for now)



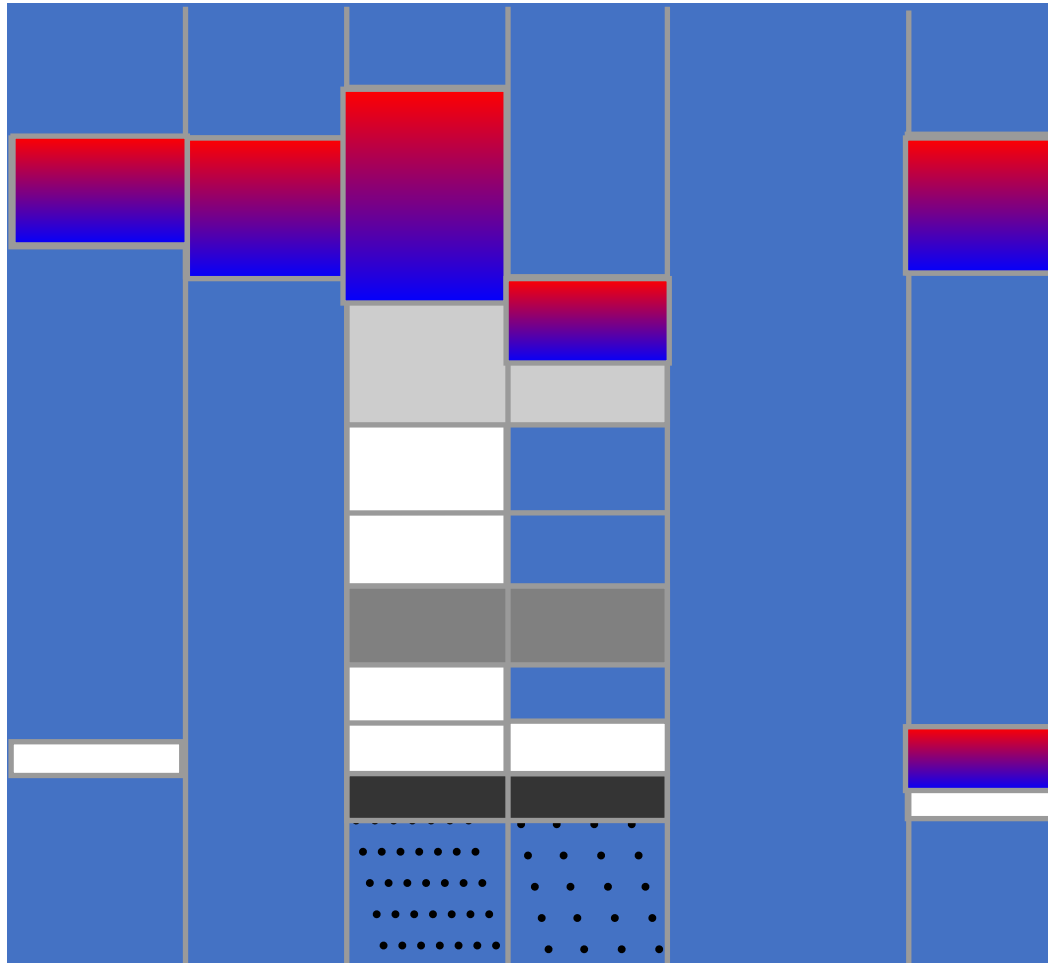
# Aircraft Convective Cloud



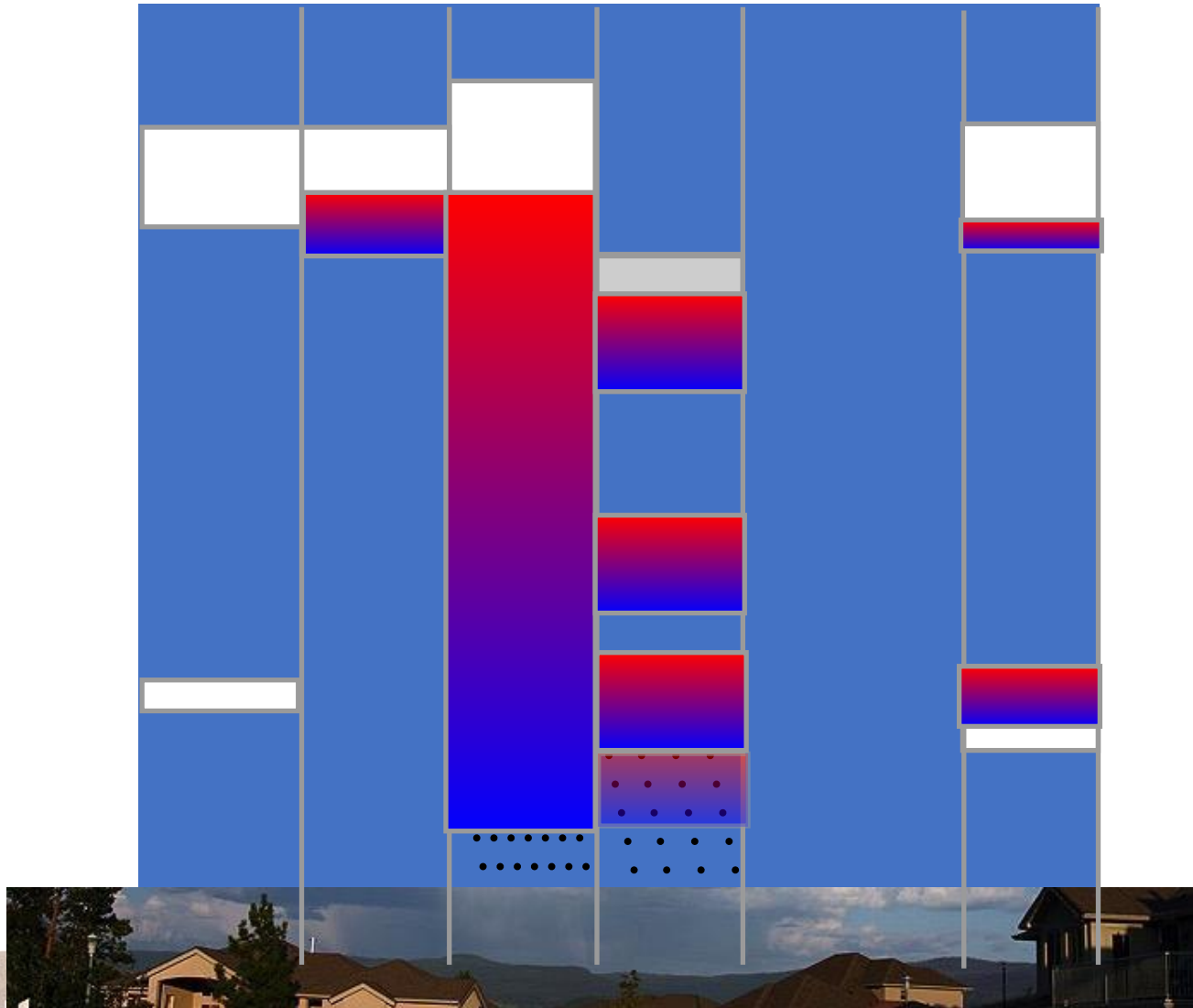
# Radar Convective Cloud



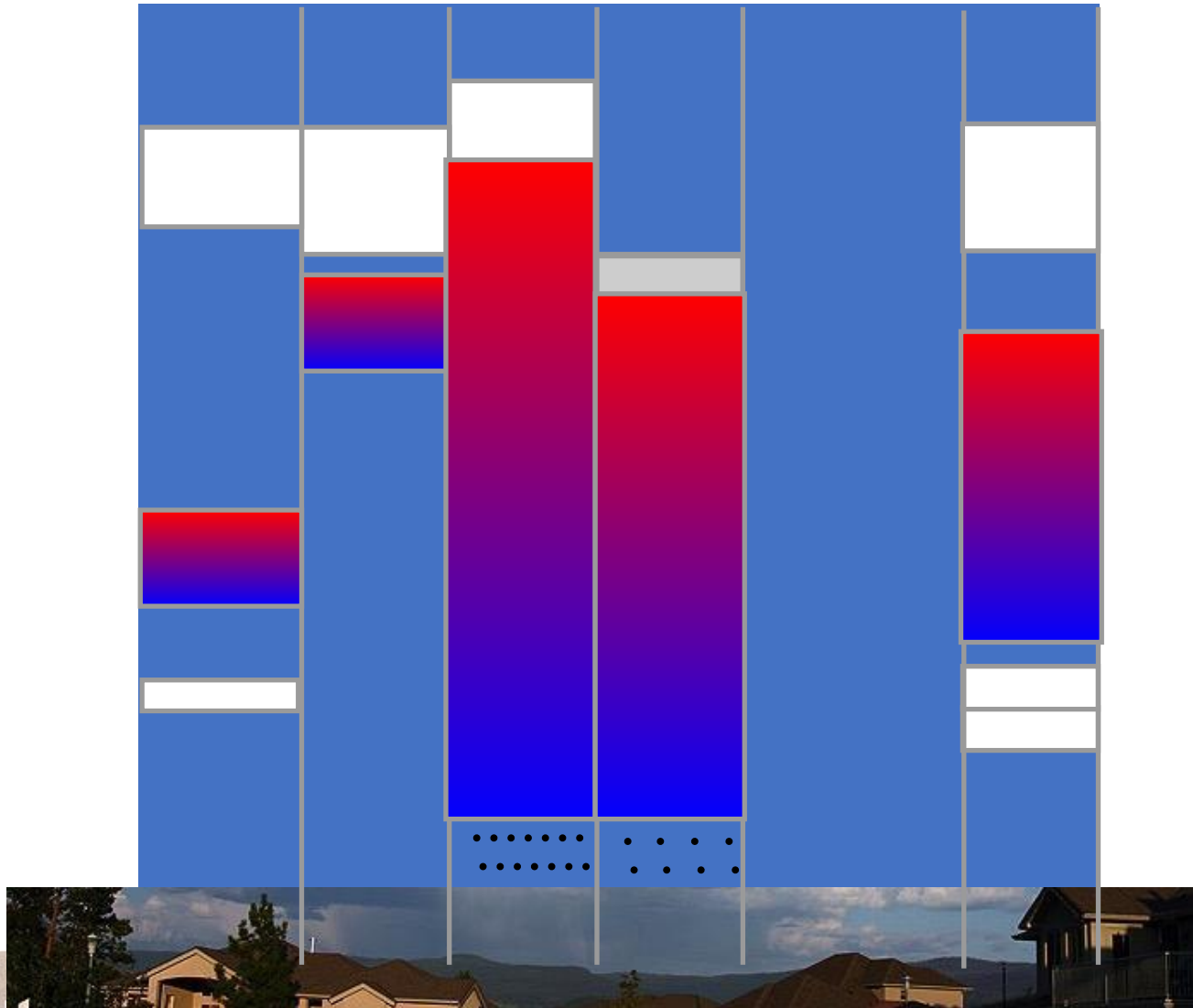
# Lidar (CALIPSO) Convective Cloud



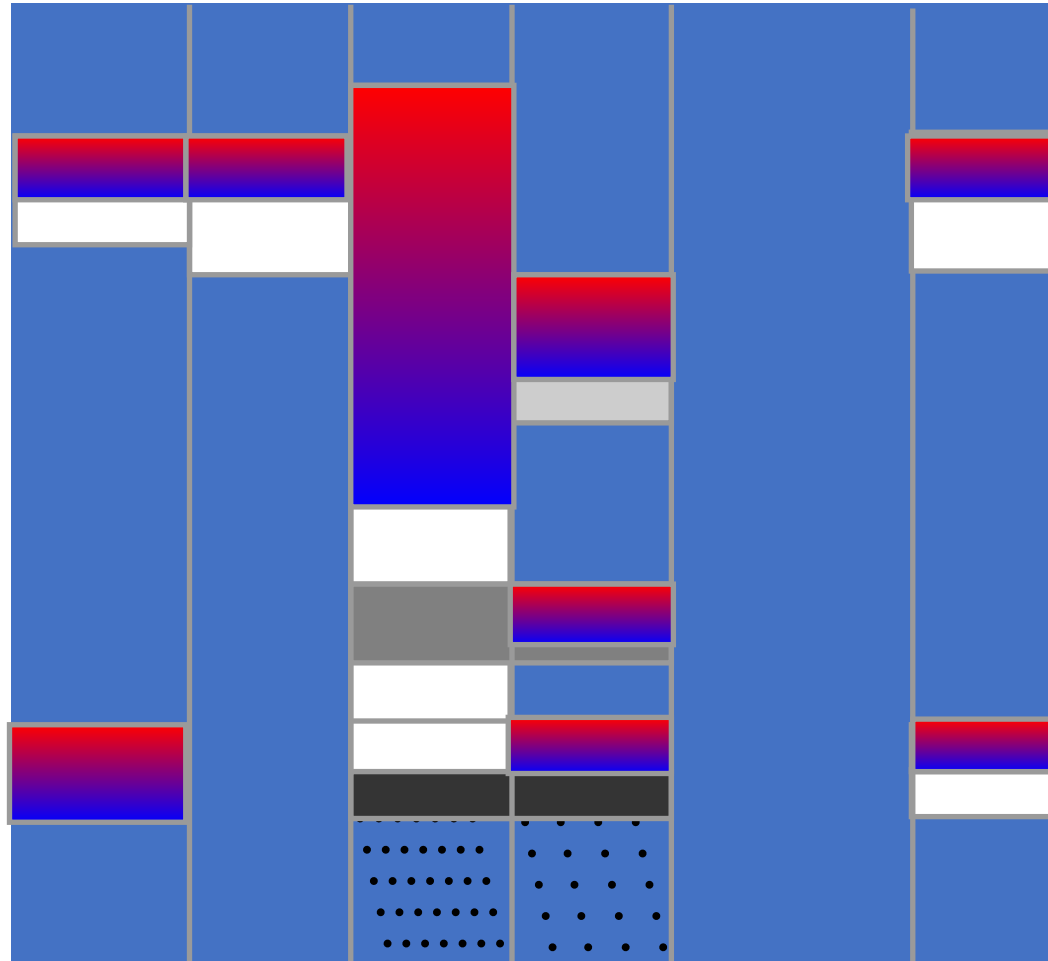
# Radar (CloudSat) Convective Cloud



# IR (MODIS) Convective Cloud



# Vis/IR (MISR) Convective Cloud



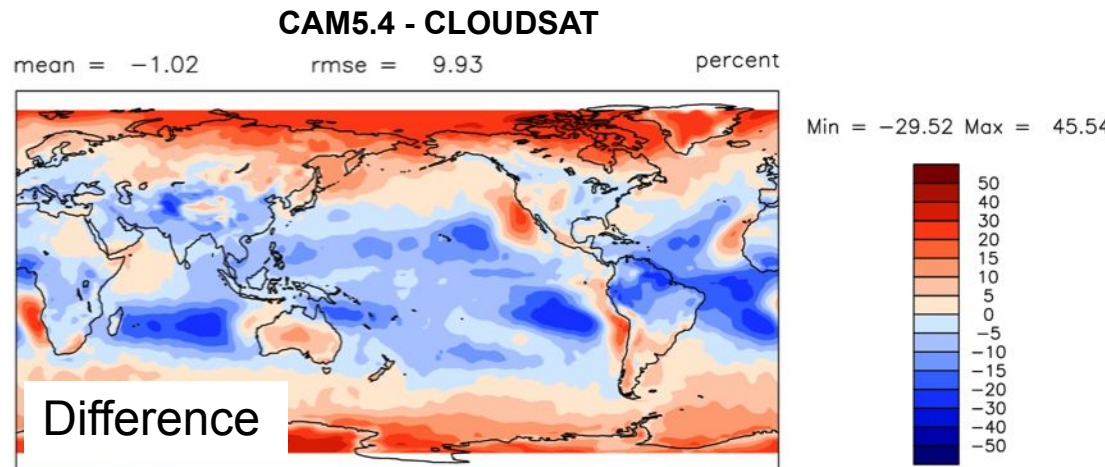
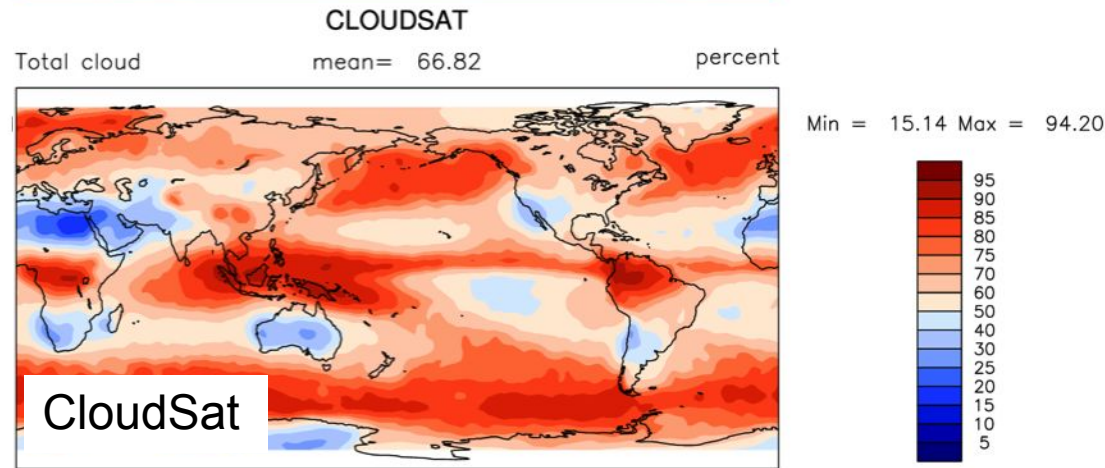
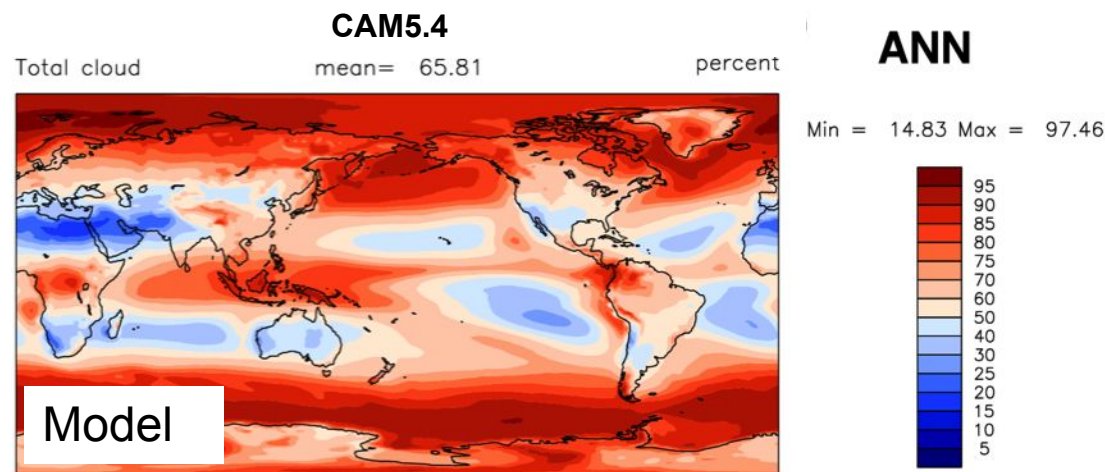
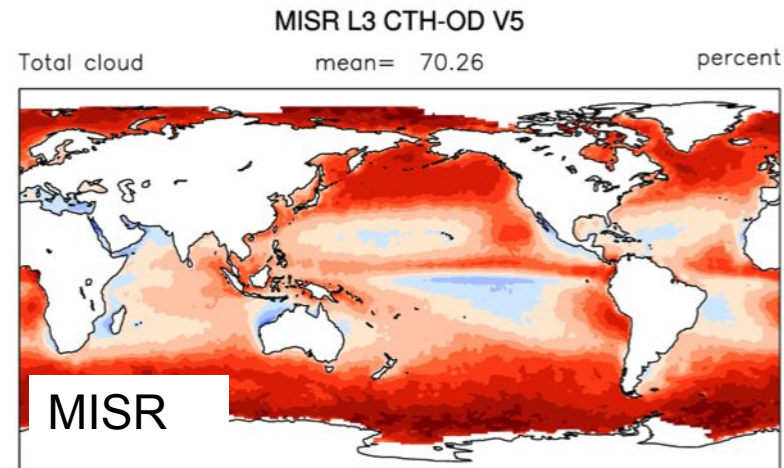
This is just  
cloud fraction:  
try it with LWP,  
Re, etc !

# Traditional Evaluation Methods

- Climate Evaluation
  - Use the easy variables: Cloud Fraction
  - Means or Climatology
- Weather Evaluation
  - Forecast: Looks okay
  - Composites
  - Forecast Skill Metrics

# Climate Evaluation: Cloud fraction...

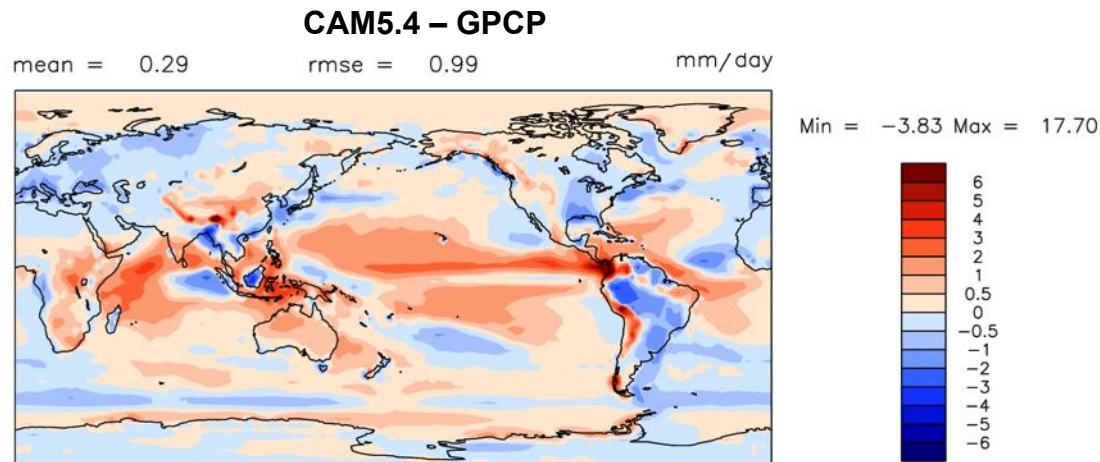
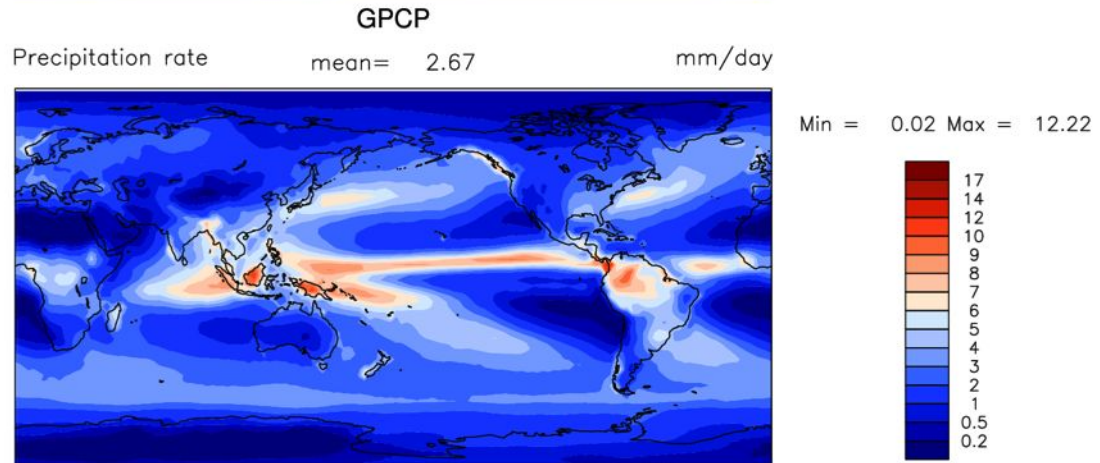
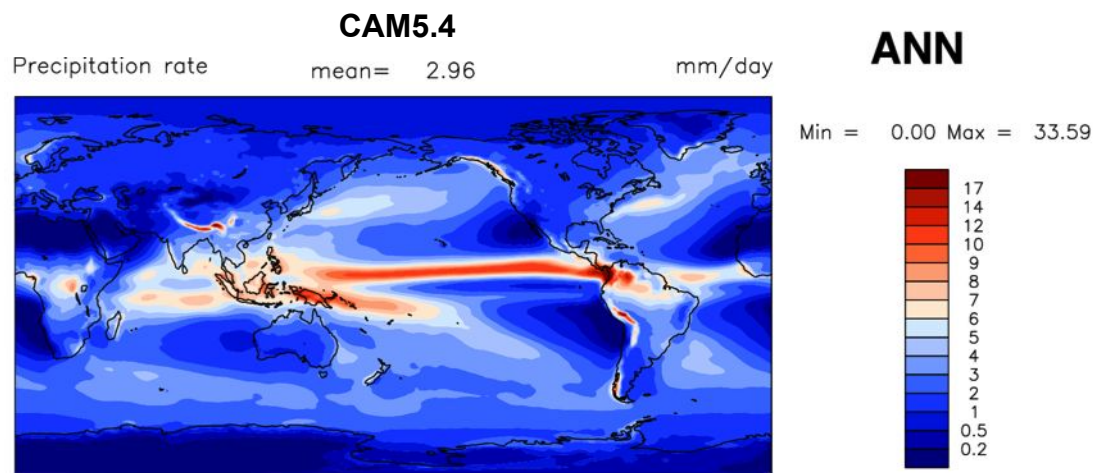
Traditional View: Model –  
CloudSat (Radar+ Lidar)



Biases change with data set.  
May even change sign!

# Mean Metrics

## Example: Precipitation



# How do we do this better?

Issue: these evaluations are loosely related to specific processes.

- Evaluation of Variability, Climate Modes
- Process based evaluation (weather, climate)
- Hindcast experiments for Weather, Climate Models
  - Multiple forecasts and forecast increments
  - Case studies in particular regions
- Satellite simulators

# Variability: Diurnal Cycle of Precip

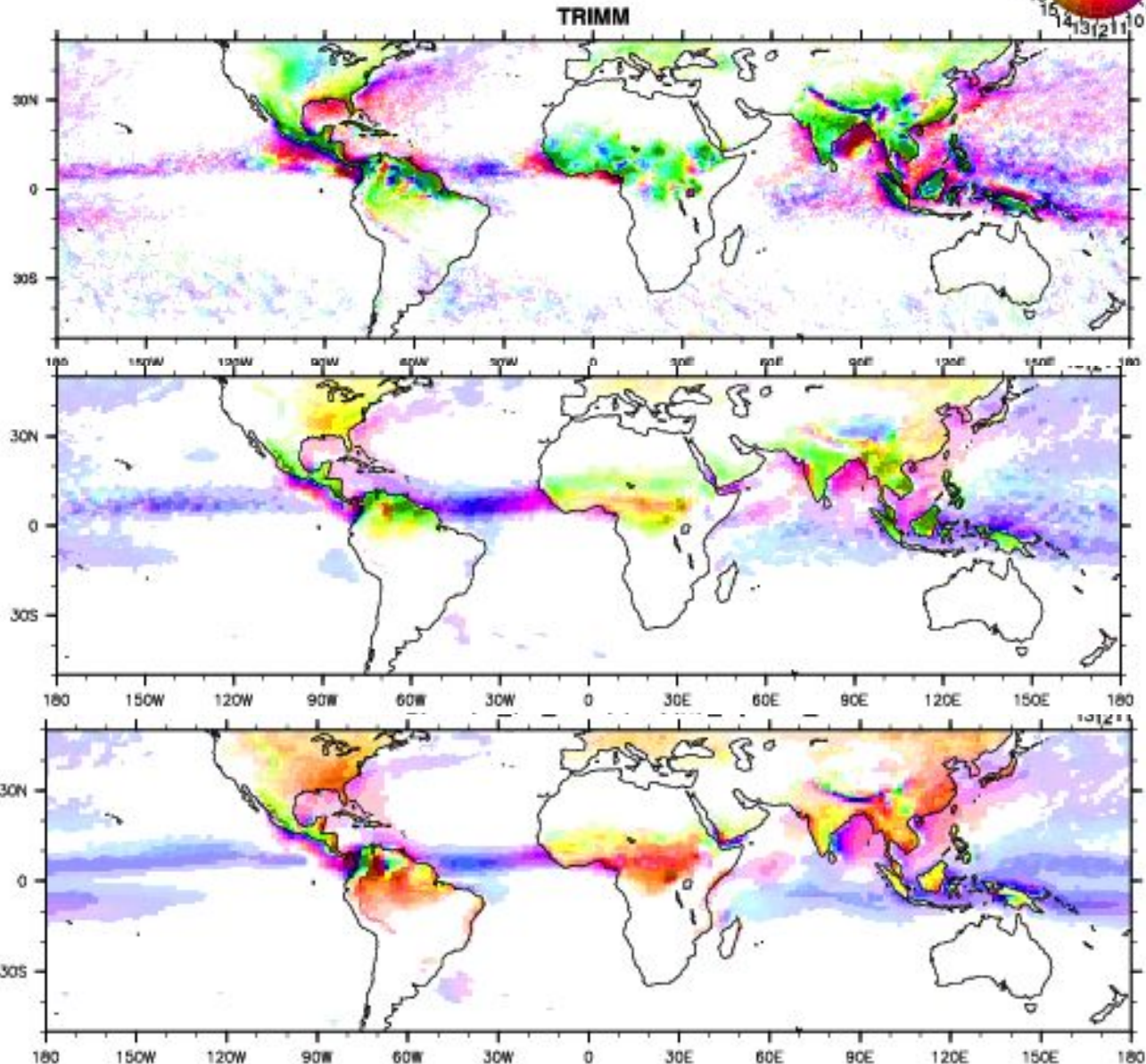
JJA Precipitation Rate

TRMM:

OBS

CAM5.5  
(new)

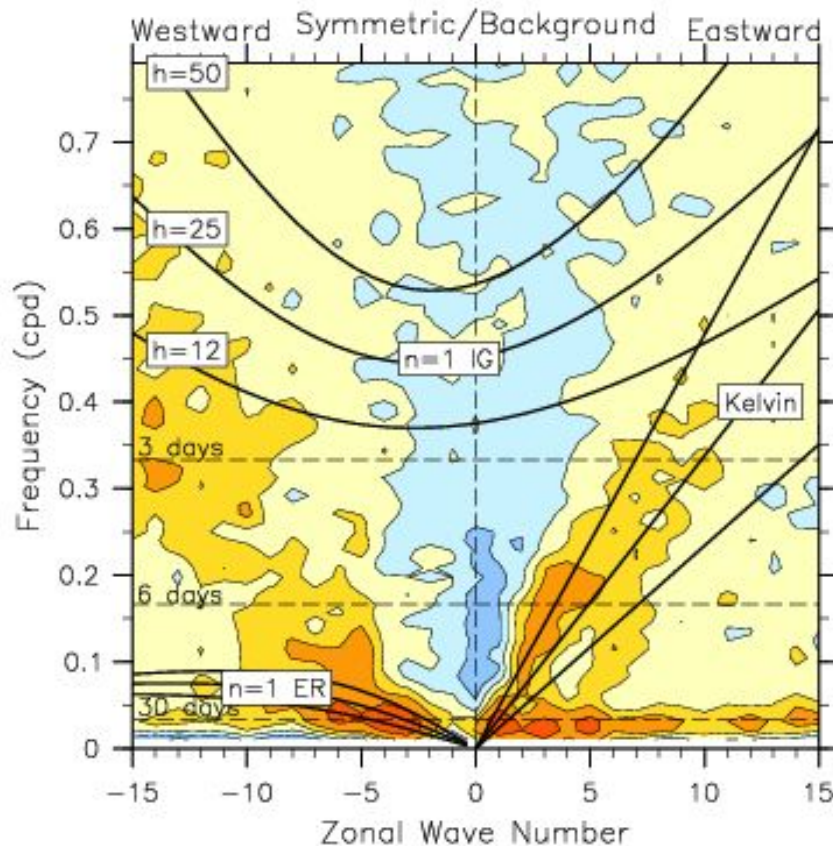
CAM5.3  
(Old)



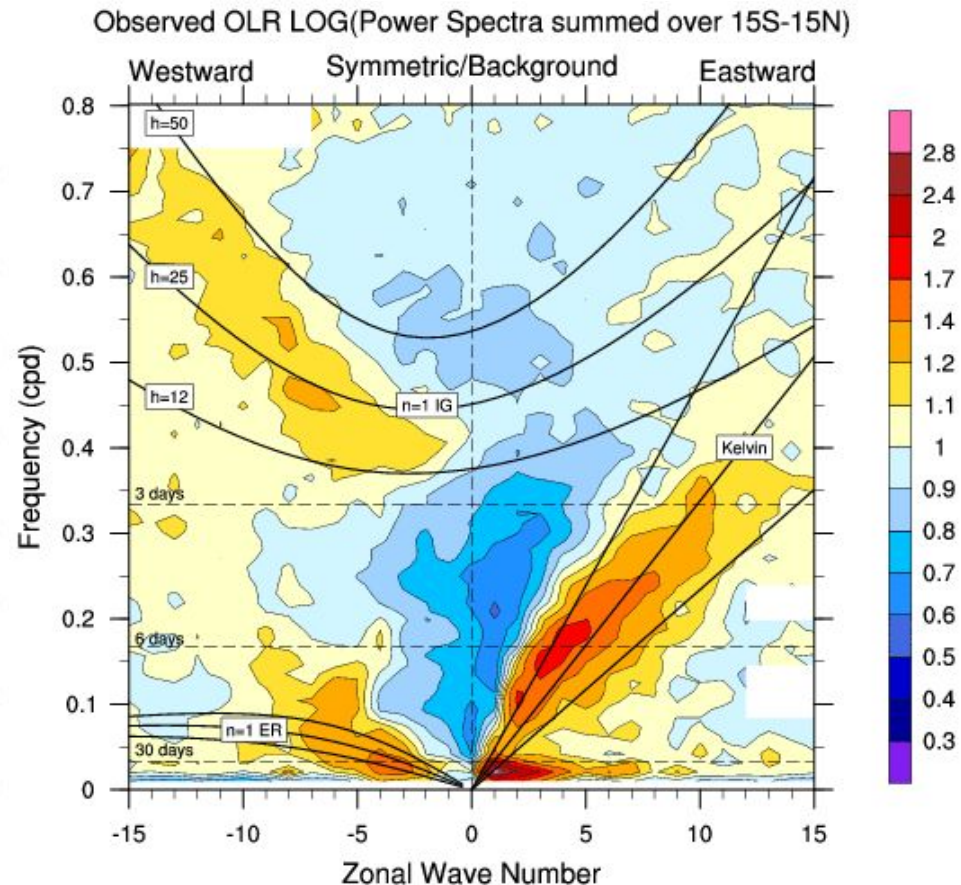
# Variability: Tropical Waves

Symmetric OLR Spectrum (Wheeler & Kiladis, 1999)

CAM5.5



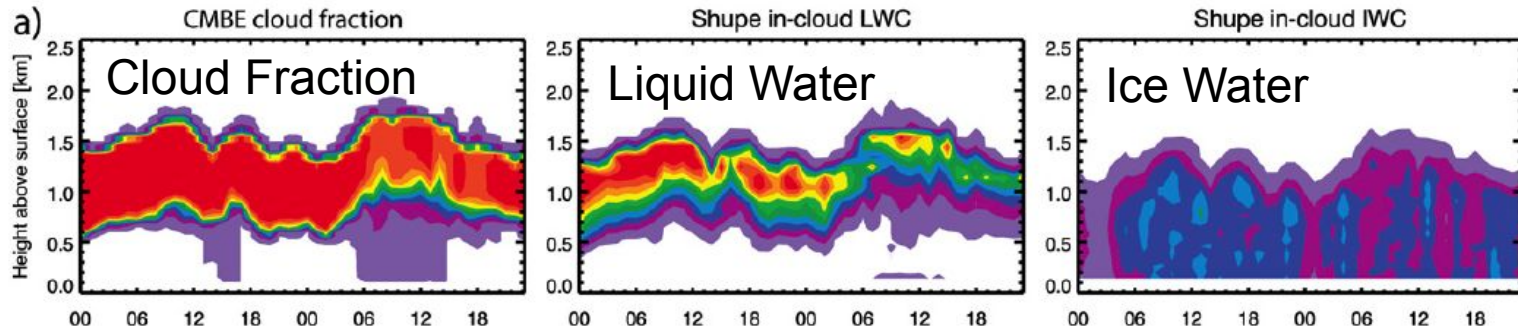
NOAA-OLR



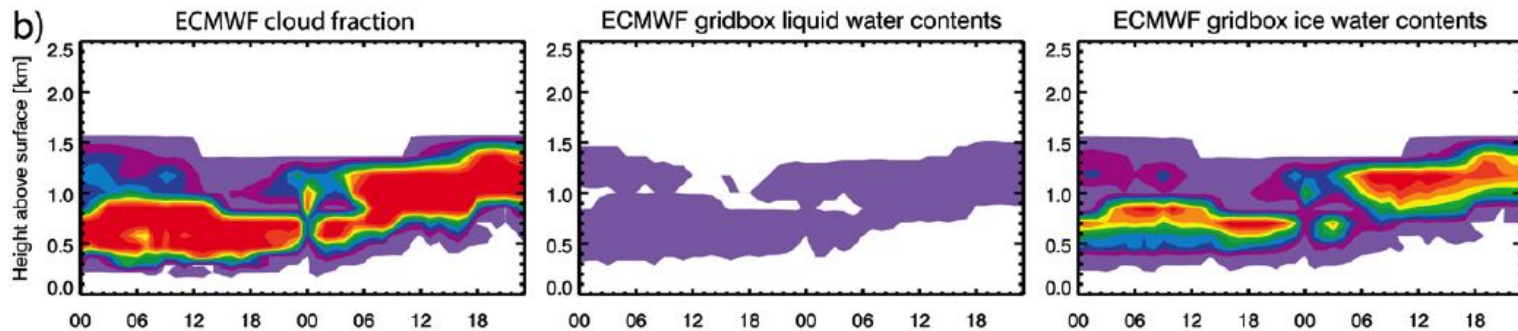
# Weather Model Forecasts: Case Study

## Mixed Phase Arctic Clouds

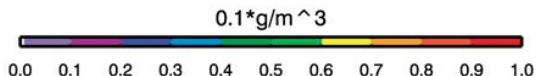
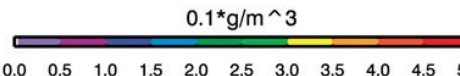
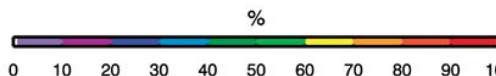
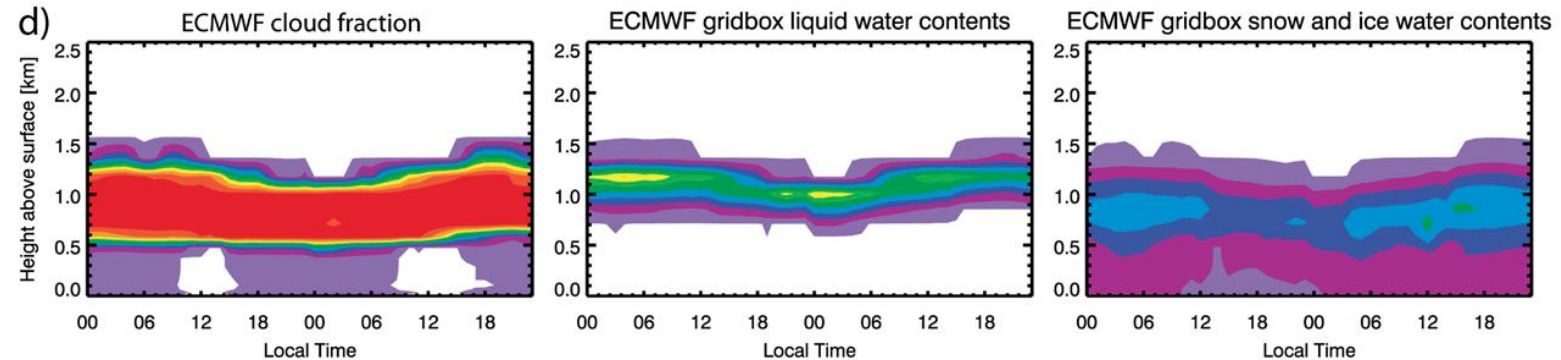
Observations



IFS Original



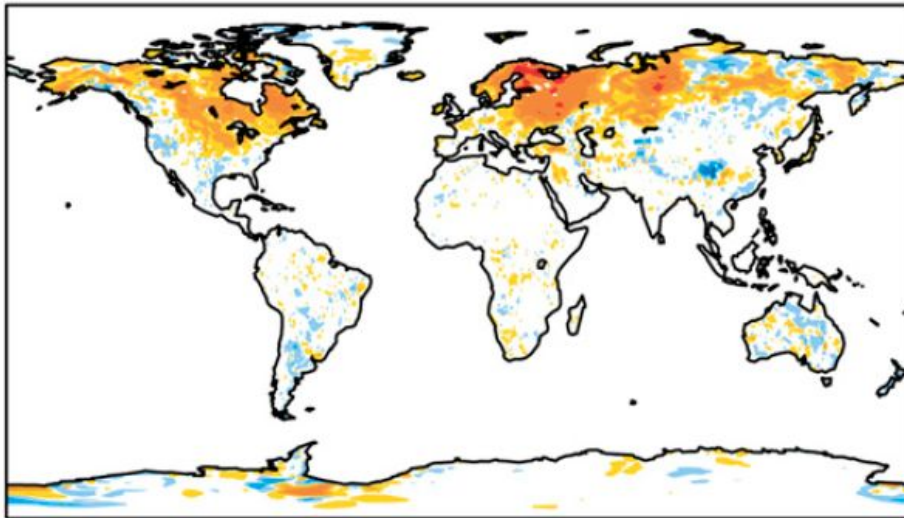
Improved!



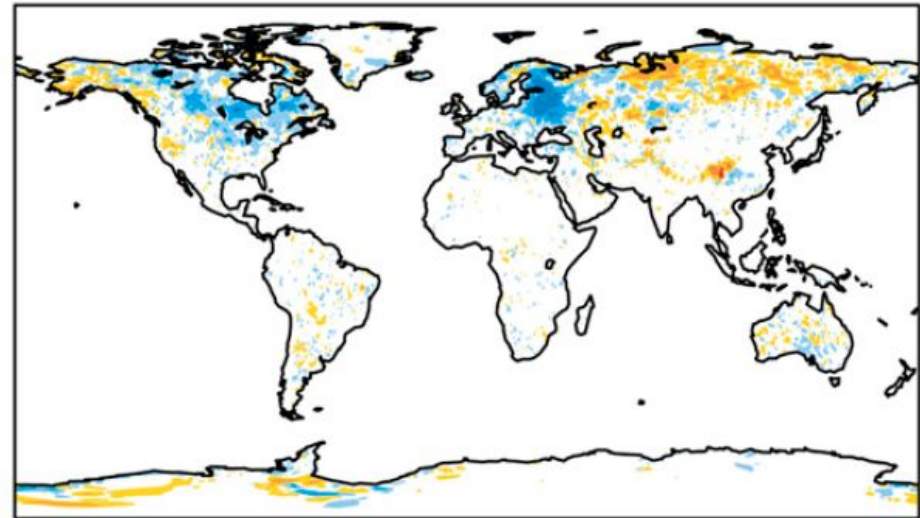
# Show it works globally

Warmer Temps = Reduced Mean Error

**a)** 2 m Temperature: Change in the mean (PROG+ - PROG)



**b)** 2 m Temperature: Change in the mean absolute error



Goal was to reduce a temperature bias

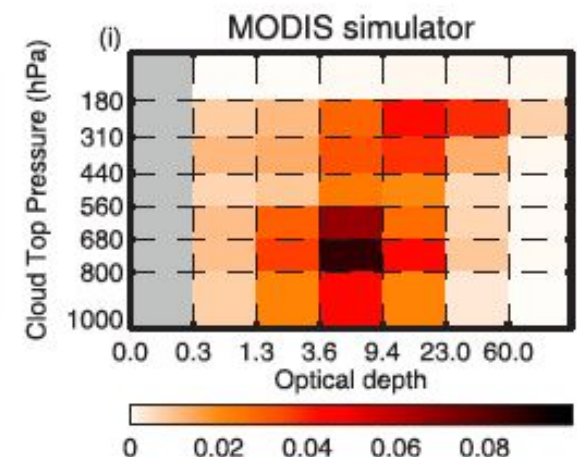
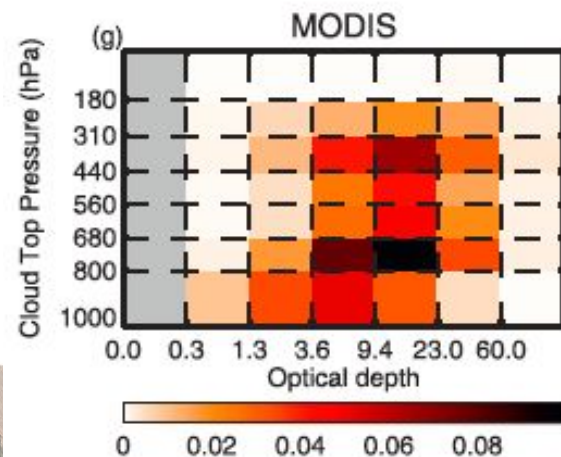
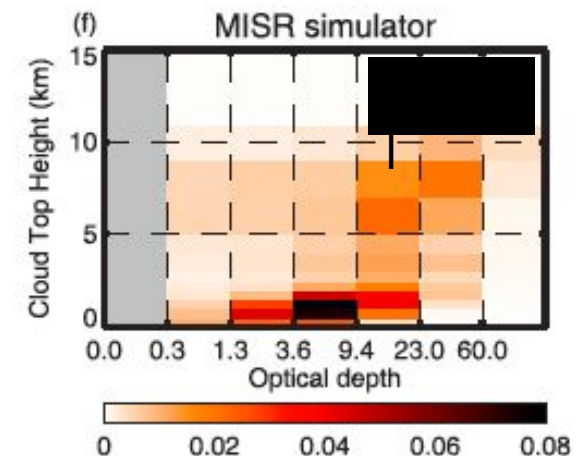
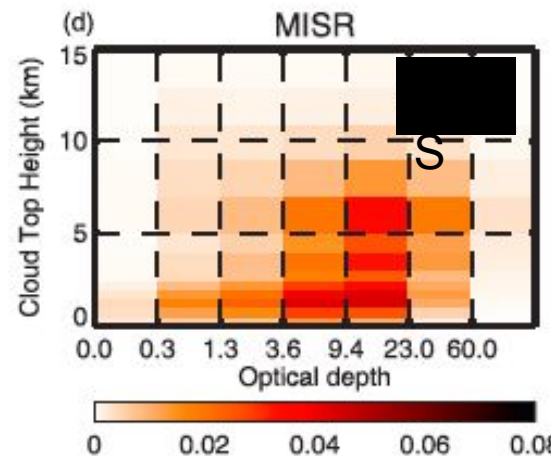
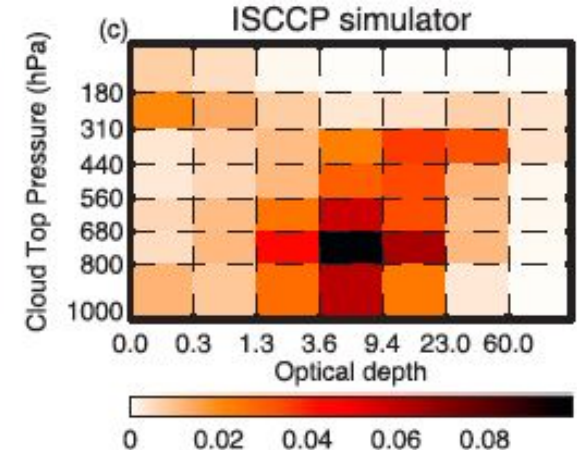
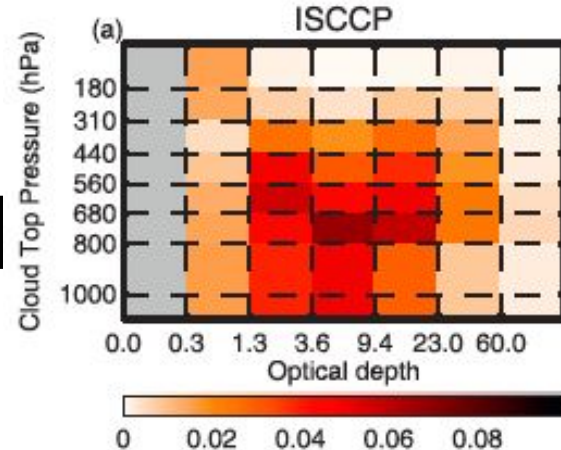
# Instrument Simulators

- Designed to “simulate” retrievals of a variety of satellites: includes MODIS, MISR, CloudSat, CALIPSO, ISCCP
- Why? Better comparisons between models and observations
- Mostly Cloud fraction, but also cloud microphysics from MODIS
- CFMIP Observation Simulator Package (COSP)
  - CFMIP = Cloud Feedback Model Intercomparison Project

# Why?

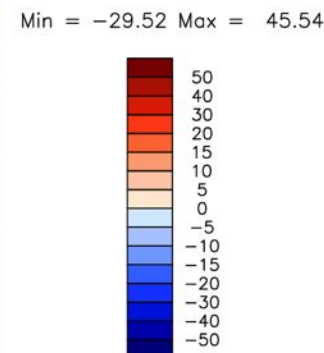
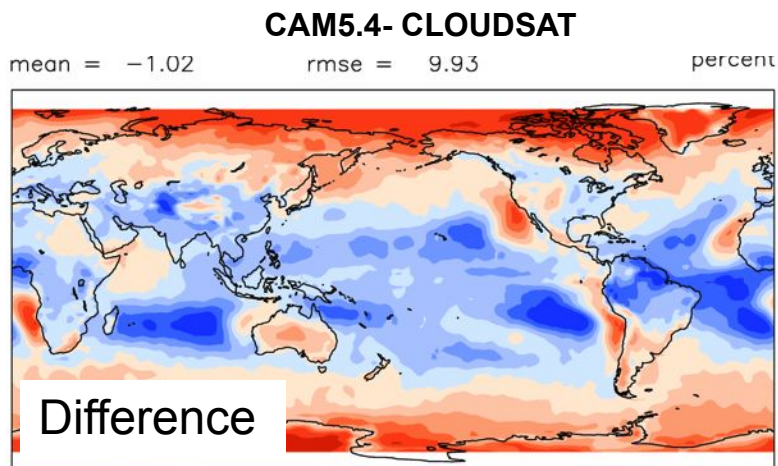
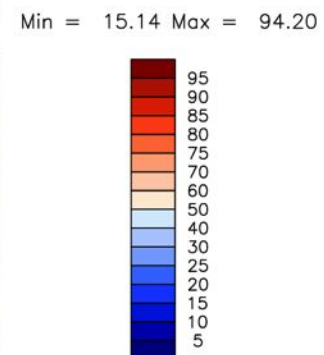
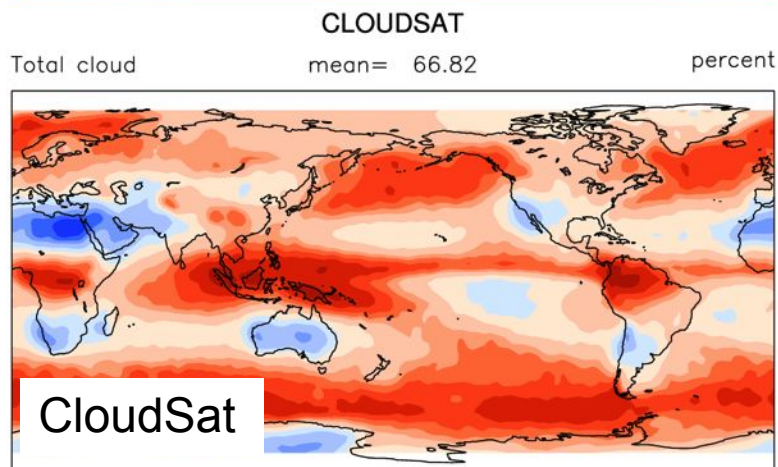
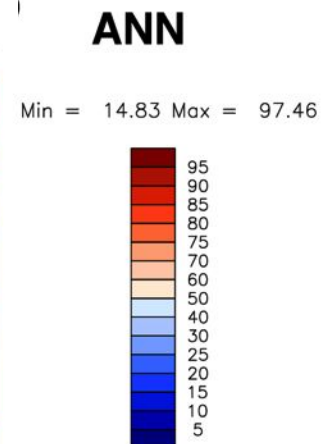
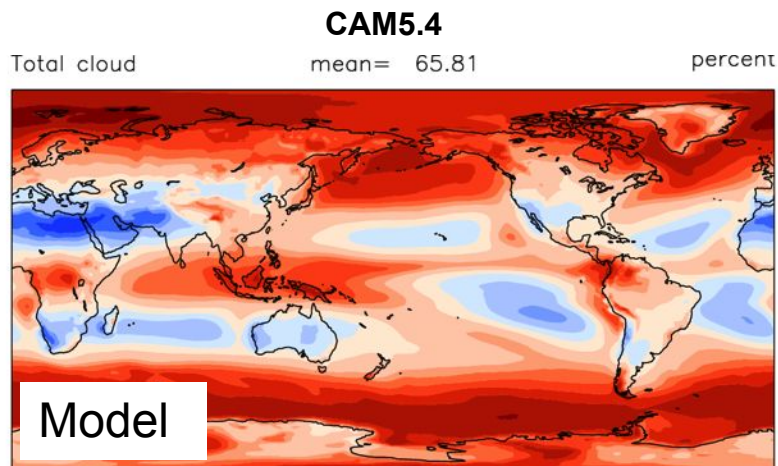
- Different instruments have different sensitivities
- Need to sample the model correctly to compare apples to apples
- Examples:
  - Cloud Fraction
  - Liquid Water Path

# Simulator Examples: UM



# Cloud fraction...

Traditional View: Model – CloudSat (Radar+ Lidar)

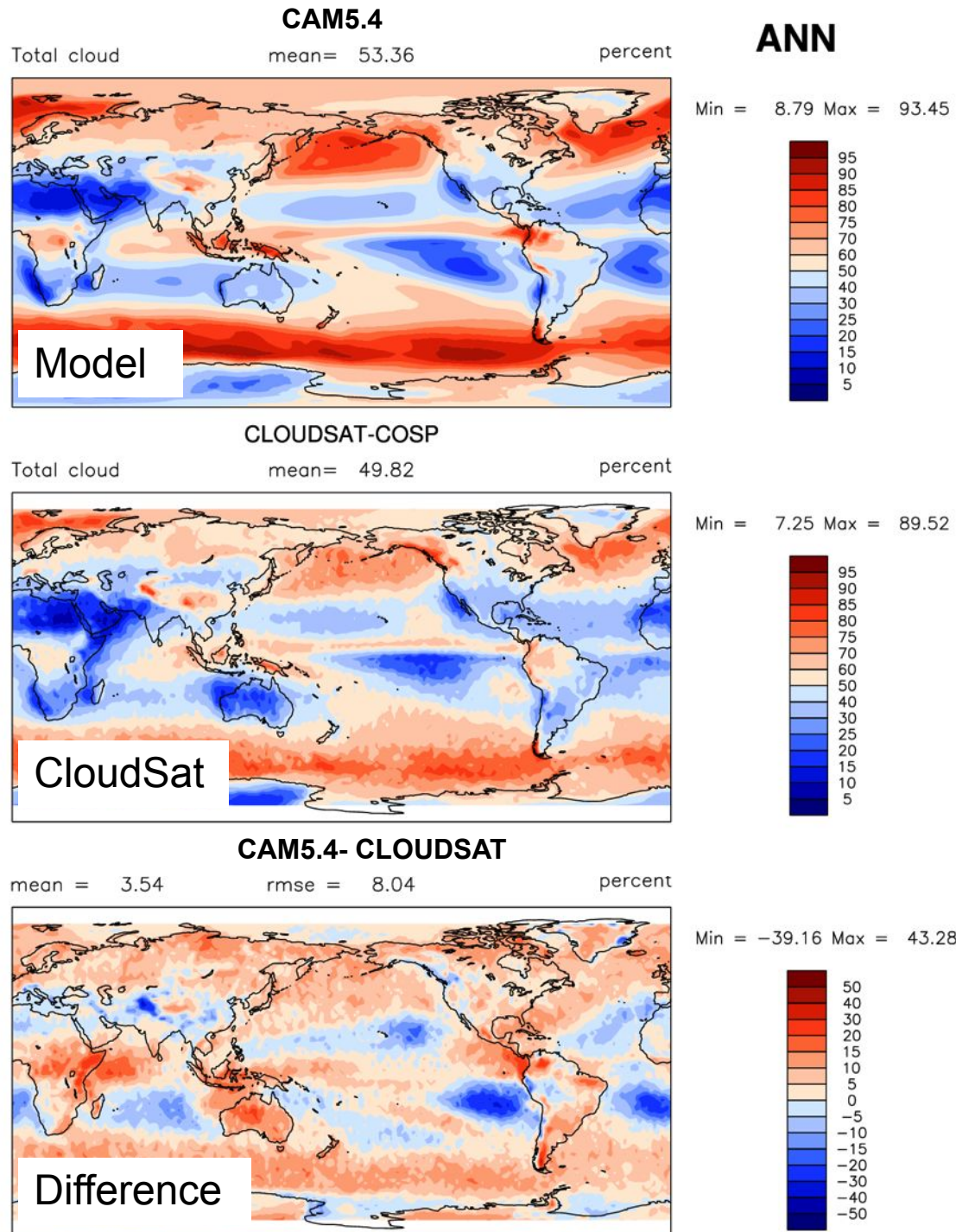


# Cloud Fraction

## SIMULATOR VIEW Model – CloudSat

CloudSat Product only. Fewer clouds (no Lidar). Model higher in tropics. Also removes some low clouds near surface (higher latitudes).

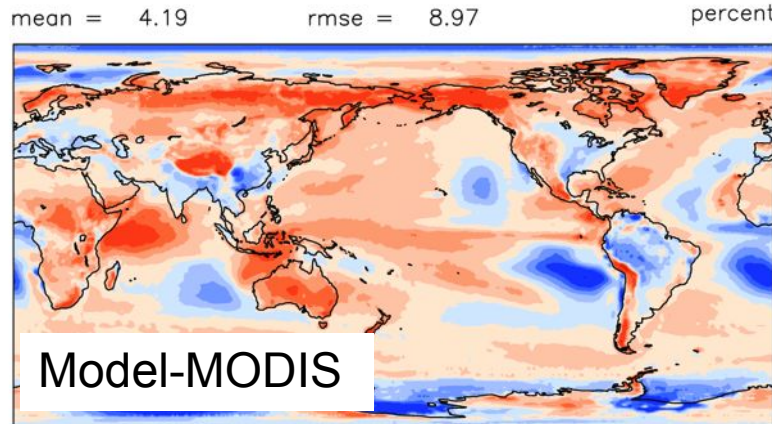
- Simulated cloud fraction is lower.
- Bias is a different sign.



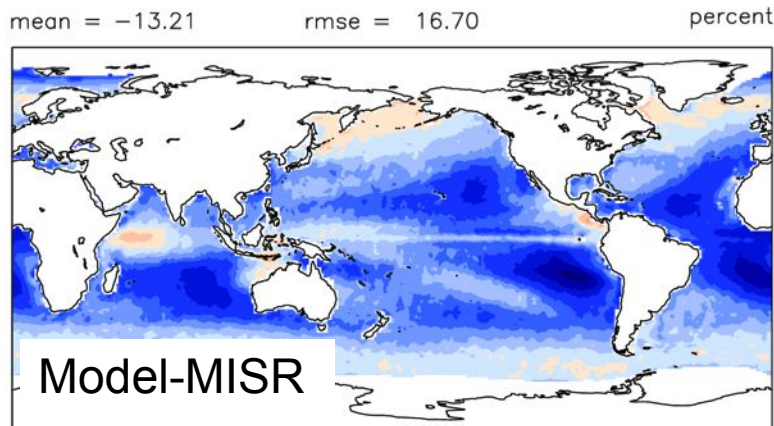
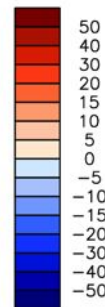
# Cloud Fraction Differences ( $\tau > 0.3$ )

Different bias against different  
instruments

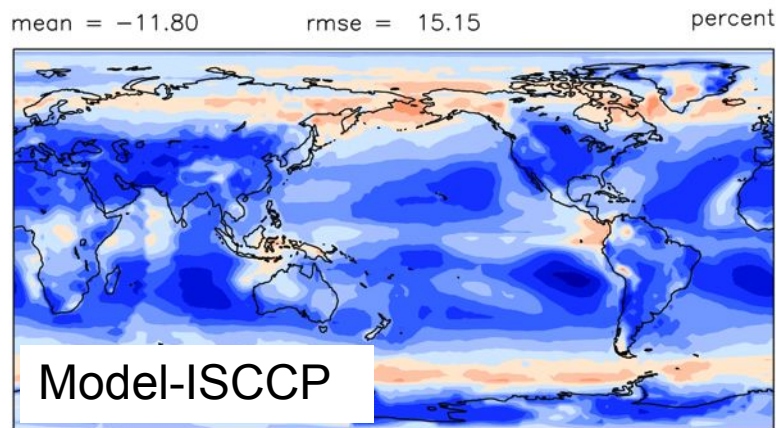
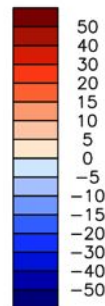
This is the same model run!



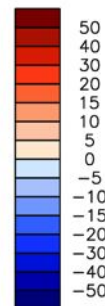
Min = -35.74    Max = 38.62



Min = -54.30    Max = 17.39



Min = -45.65    Max = 17.55



# LWP: Wrong Message

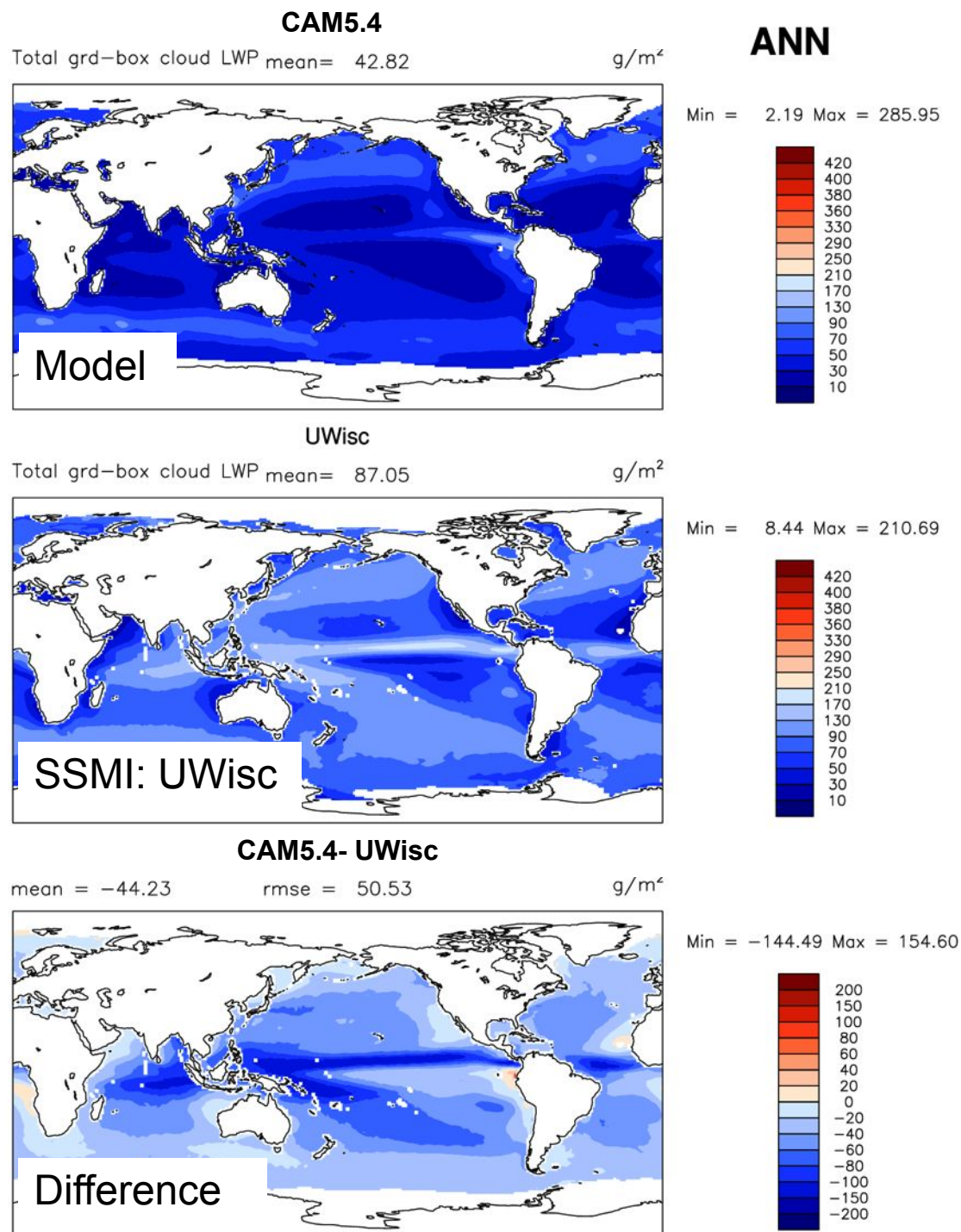
Traditional comparison of Model LWP field against Microwave Satellite Observations of LWP.

Model is low.

But cloud forcing looks okay.  
Cloud fraction looks okay.

What is going on?

Same problem with comparison with MODIS LWP retrievals...

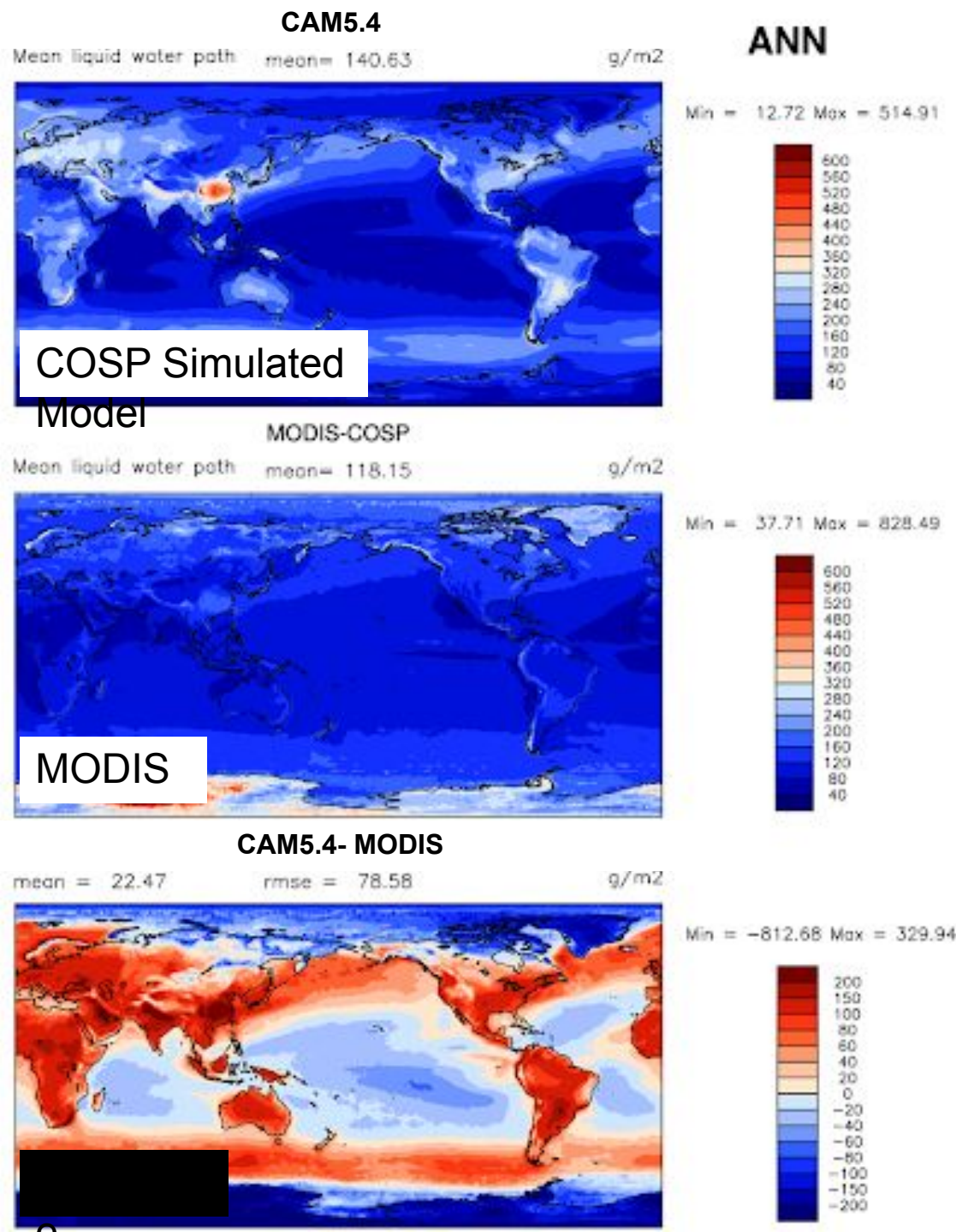


# LWP: Correct Message

Use of the MODIS simulator for LWP: implies an Adiabatic assumption for low clouds.

The model is not Adiabatic, but assuming it is Adiabatic increases LWP, especially over land and storm tracks

Now the model is slightly HIGHER than observations (+20%) rather than -50% LOW.

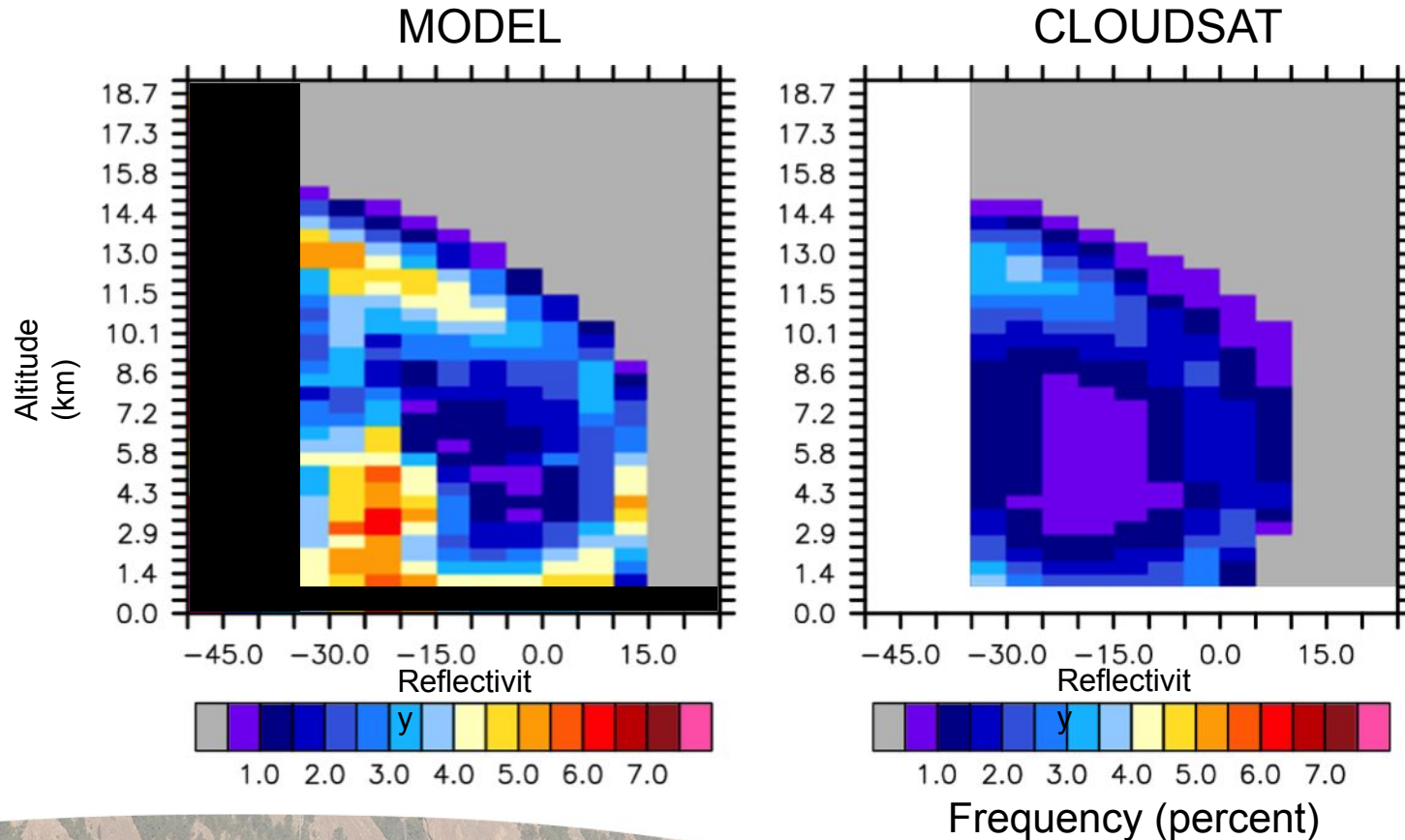


# Simulate Reflectivity

Simulate observed quantities: in this case, Reflectivity ( $Z \propto D^{-6}$ )  
Cumulative Frequency by Altitude Diagram (CFAD)

Shows modes of variability and regimes in models and observations

Here: thin, low clouds too extensive and high, too much moderate drizzle



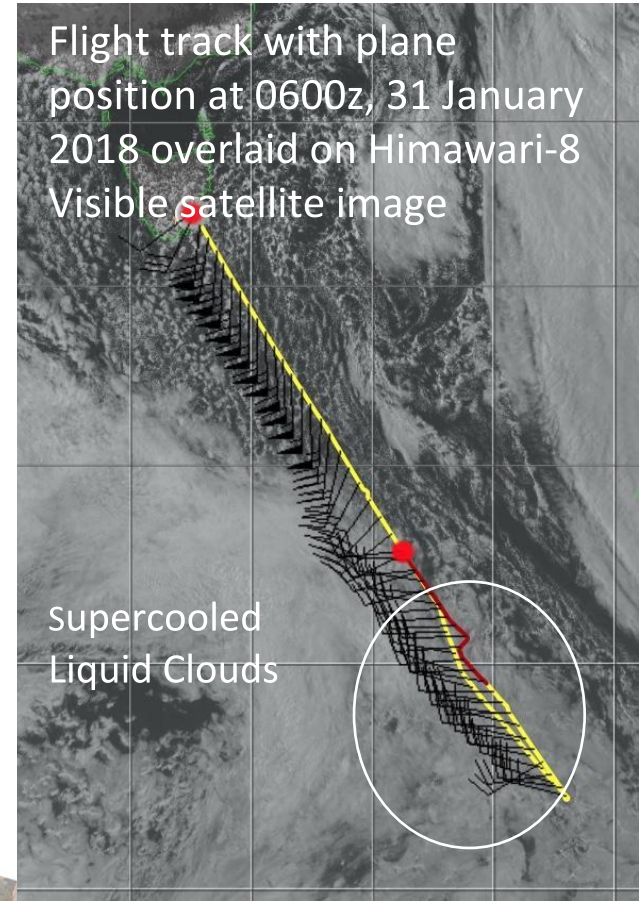
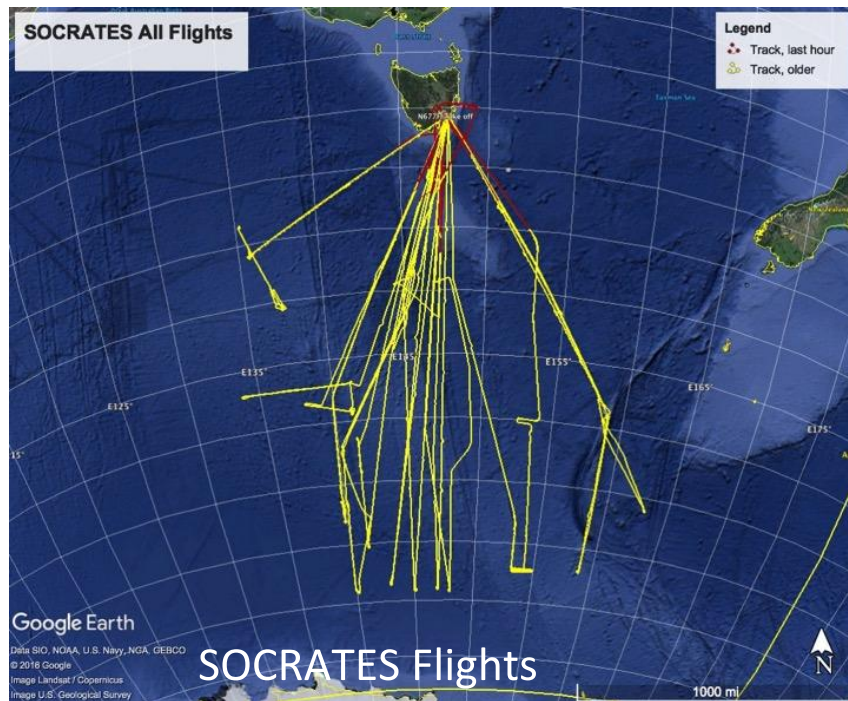
# Complexities

- Sub-grid scales are hard to observe
  - Hard to make model and observations consistent
- Global constraints on clouds are good (& bad)
  - Benefit of the climate scale
  - Integral of regimes: zero in on regimes, analogues
  - Easy to get the right clouds for the wrong reasons
- Simulators provide an integrated view
  - Sometimes hard to disentangle processes

# In Situ Observations



The 2018 SOCRATES project used Southern Ocean aircraft and ship observations to sample clouds. Simulate details of cloud drop size distributions



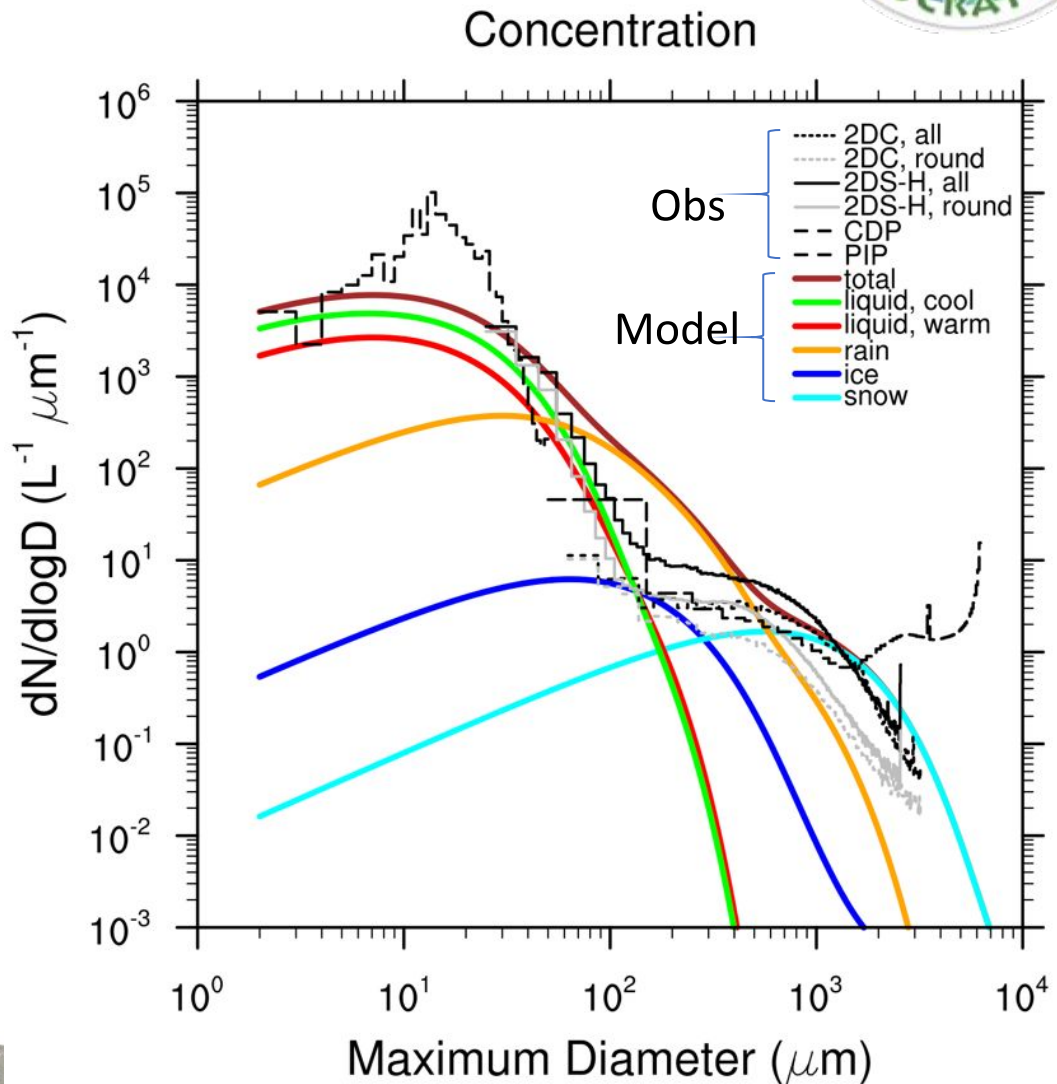


# Model v. Obs Size Distributions

Model (CAM6) nudged to observations and sampled along SOCRATES flight tracks over the S. Ocean.

Model does a good job at reproducing obs, with some caveats.

Hints at fundamental issues with modal formulations (Exponential and Gamma functions).



# CAM Diagnostic 'Packages'

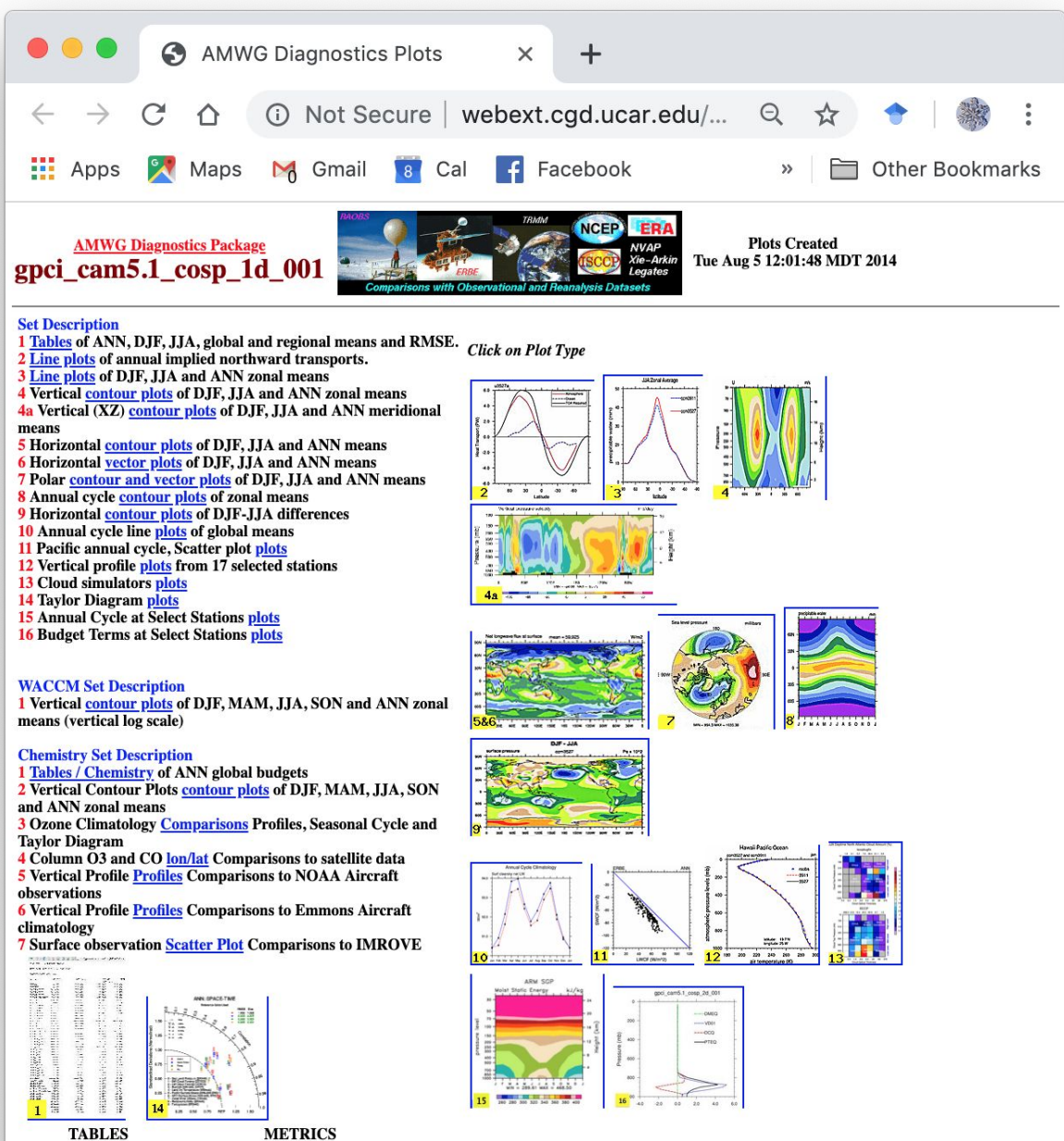
- **Mean Climate:** Atmospheric Model Working Group (AMWG) Diagnostic Package
- **Atmospheric Processes:** NOAA Model Diagnostics Task Force (MDTF) Package
- **Climate Variability:** Climate Variability Diagnostic Package (CVDP)

# Mean Climate

[http://www.cesm.ucar.edu/working\\_groups/Atmosphere/amwg-diagnostics-package/](http://www.cesm.ucar.edu/working_groups/Atmosphere/amwg-diagnostics-package/)

- AMWG Diagnostic Package
- Shell script that Calls NCL
- Set analysis
  - Model v. Obs
  - Model v. Model
  - Works on monthly mean model output
- Extensions for:
  - Chemistry
  - WACCM (Stratosphere and Mesosphere)

Model v. Obs  
[Link to example](#)



# Atmospheric Processes

[http://www.cesm.ucar.edu/working\\_groups/Atmosphere/mdtf-diagnostics-package/](http://www.cesm.ucar.edu/working_groups/Atmosphere/mdtf-diagnostics-package/)

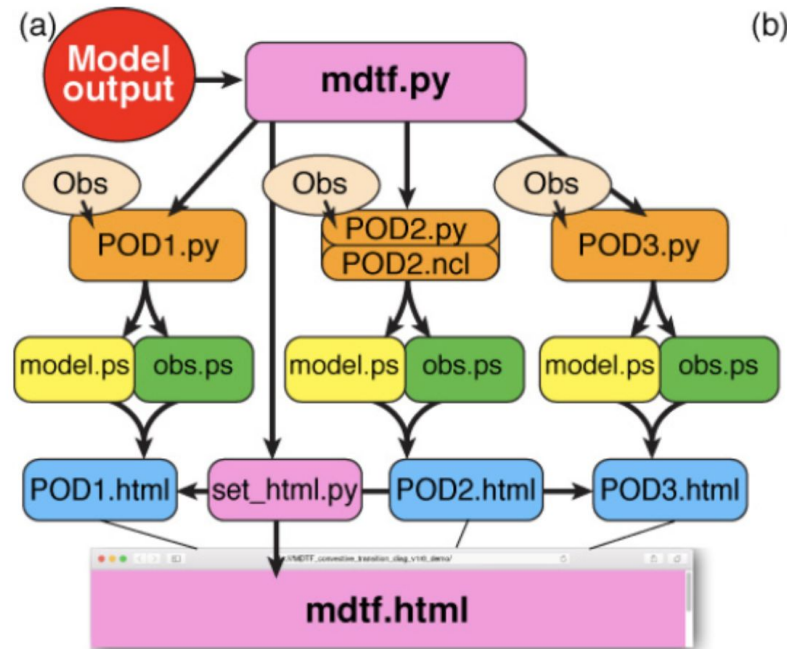
## Model Diagnostics Task Force (MDTF) diagnostic package

Designed to validate physical processes in climate models.

- Focus on model development through Process-Oriented Diagnostics (PODs)
- **Sub-monthly model output** (daily -> 3hr, different for different PODs)
  - *Latest incarnation of AMP's variability package:  
contains MJO, Wheeler-Kiladis and Diurnal cycle figures*
- For Model Developers
- Runs on CESM, GFDL and CMIP model output
  - *Once a diagnostic is implemented, can run easily on other model output*
- Modular, portable, extensible, open-source, community\*
  - *Run on your data: 6 implemented modules + 3 in testing + > 7 in development*

# MDTF Design

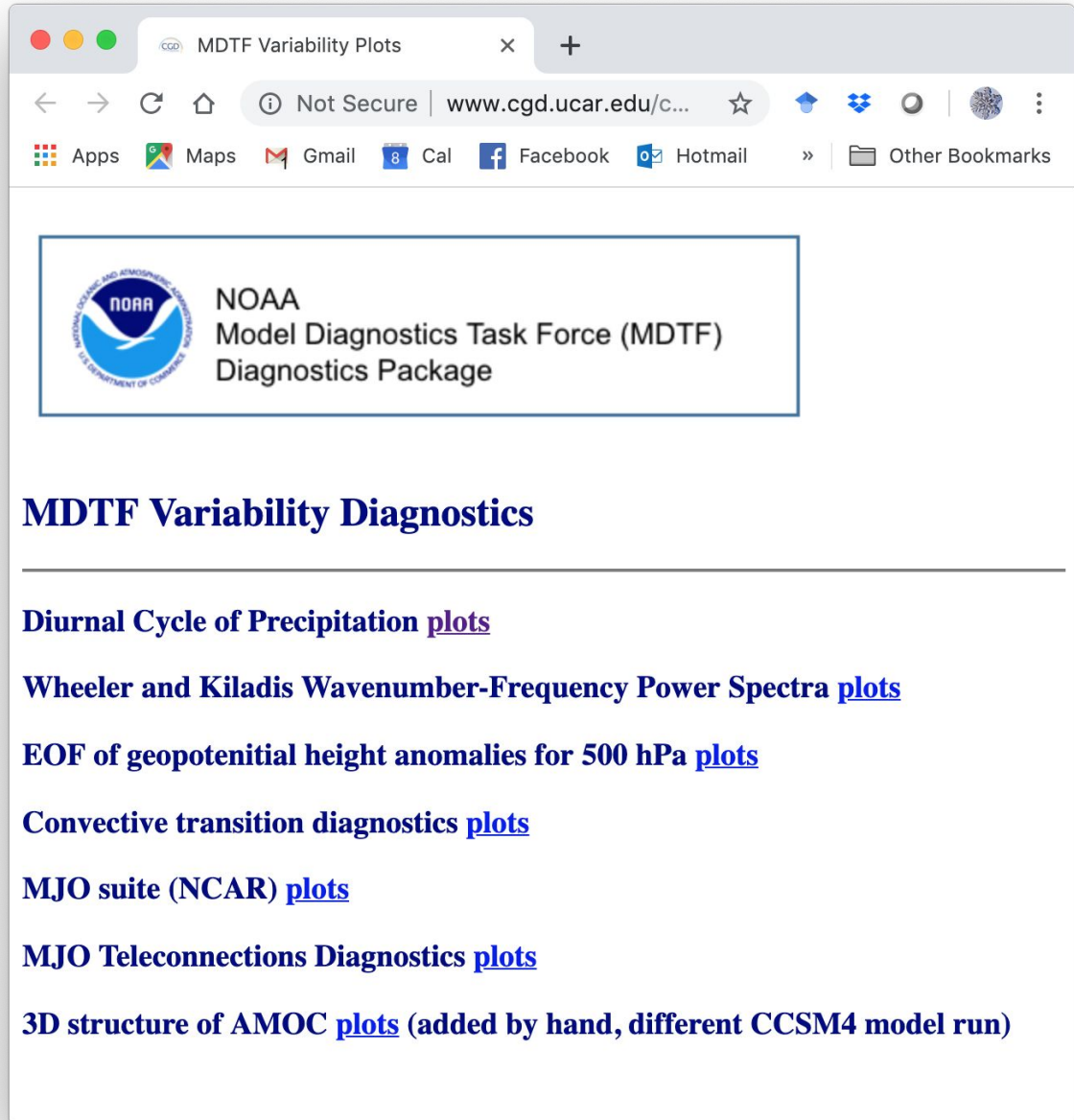
- Open-source and distributed under the LGPLv.3 license
- Python driver *mdtf.py*
  - Establishes file paths, variable names, output location
  - Calls individual PODs
    - Written in any open-source language, currently all are NCL or python
    - Currently provides its own observatio
    - Each creates a web page to display its output
    - Independent from each other within the framework
  - Writes a top-level webpage with links to the individual pages



# MDTF example

Analysis CAM6 v. Obs

[Link to example](#)



The screenshot shows a web browser window with the title "MDTF Variability Plots". The address bar shows the URL "www.cgd.ucar.edu/c...". The browser's toolbar includes icons for Apps, Maps, Gmail, Cal, Facebook, Hotmail, and Other Bookmarks. The main content area features the NOAA logo and the text "NOAA Model Diagnostics Task Force (MDTF) Diagnostics Package". Below this, the heading "MDTF Variability Diagnostics" is followed by a list of diagnostic categories, each with a link to "plots":

- Diurnal Cycle of Precipitation [plots](#)
- Wheeler and Kiladis Wavenumber-Frequency Power Spectra [plots](#)
- EOF of geopotential height anomalies for 500 hPa [plots](#)
- Convective transition diagnostics [plots](#)
- MJO suite (NCAR) [plots](#)
- MJO Teleconnections Diagnostics [plots](#)
- 3D structure of AMOC [plots](#) (added by hand, different CCSM4 model run)

# Climate Variability

[http://www.cesm.ucar.edu/working\\_groups/CVC/cvdp/](http://www.cesm.ucar.edu/working_groups/CVC/cvdp/)

- CVDP uses monthly mean data to analyze climate variability
- Means, Standard Deviations
- Modes of variability (PDO, ENSO)
- Multiple model simulations against observations (separate panels)

# CVDP Example

## CESM Simulations [Link to Example](#)

Climate Variability Diagnostics x +

Not Secure | webext.cgd... | Search | Star | Profile

Apps | Maps | Gmail | Cal | Other Bookmarks

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**NCAR UCAR** | CGD's Climate Analysis Section  
**Climate Variability Diagnostics Package**

[Methodology and Definitions](#)  
Metrics Tables: [Pattern](#)  
[Correlations](#) | [RMS](#)  
Input Namelists: [Observations](#)  
| [Models](#)  
[MOC](#) | [PR](#) | [PSL](#) | [SIC](#)  
Derived [NH](#) | [TS](#) | [SIC SH](#) | [SND](#) | [TAS](#) | [SIC SH](#)  
CESM Control Comparison

Climatological Period Used:  
Full  
Created: Mon May 21  
20:45:38 MDT 2018  
CVDP Version 5.0.0

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**Means**

SST	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
TAS	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
PSL	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
PR	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
[PR]	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
SND	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
SIC NH	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
SIC SH	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>

**Standard Deviations**

SST	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
TAS	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
PSL	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
PR	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
SND	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
SIC NH	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>
SIC SH	<a href="#">DJF</a>	<a href="#">MAM</a>	<a href="#">JJA</a>	<a href="#">SON</a>	<a href="#">Annual</a>

**Coupled Modes of Variability**

ENSO	Spatial Composites	SST/TAS/PSL	PR
		<a href="#">JJA<sup>0</sup></a> <a href="#">SON<sup>0</sup></a>	<a href="#">JJA<sup>0</sup></a> <a href="#">SON<sup>0</sup></a>
		<a href="#">DJF<sup>+1</sup></a> <a href="#">MAM<sup>+1</sup></a>	<a href="#">DJF<sup>+1</sup></a>
		<a href="#">El Niño</a> <a href="#">Hovmöller</a>	<a href="#">La Niña</a> <a href="#">Hovmöller</a>
		<a href="#">Timeseries</a>	<a href="#">Power Spectra</a>

# Summary: Diagnostics

- Measurements have models in them
- Try to compare Apples to Apples
- Often better to bring the model to the observations
- Satellite simulators are a way forward
- Sometimes you need in-situ data
- CAM diagnostics:
  - Mean Climate
  - Atmospheric Processes
  - Climate Variability