

Cloud Evaluation and Impacts in Large Scale Models

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Outline

Main Story: From identifying biases to process improvement using the 'weather' scale

- Motivation: the impact of clouds for weather & climate: focus mostly on climate examples
- Example of a cloud parameterization “suite”
- How we simulate and observe clouds
- Evaluation methods: some examples
 - The traditional way
 - Best practices
- Summary of Best Practices for Evaluation
- Challenges

Warning: Focus is on climate.

Motivation: Weather

Hypothesis: “most impacts of extreme weather occur through cloud processes.” Exceptions: windstorms, heat and cold.

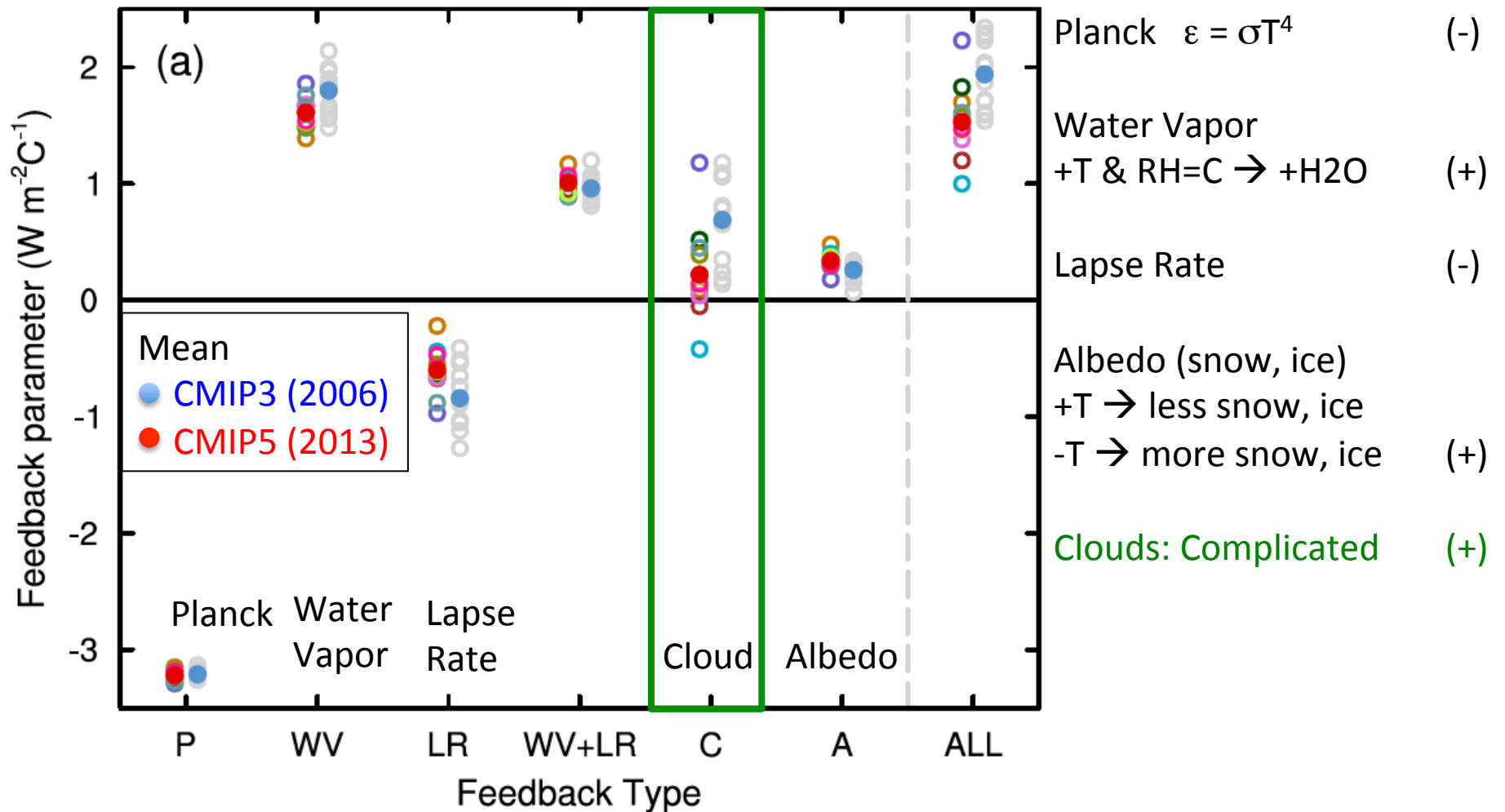
- Precipitation (or lack thereof)= flooding, drought
- Severe Thunderstorms, Hail, Tornadoes
- Tropical Cyclones
- Ice storms

Motivation: Climate

Hypothesis: “clouds are the largest uncertainty in our ability to predict future climate change.”

- Cloud Radiative Effects
- Cloud Feedbacks
- In addition to understanding the frequency of extreme weather

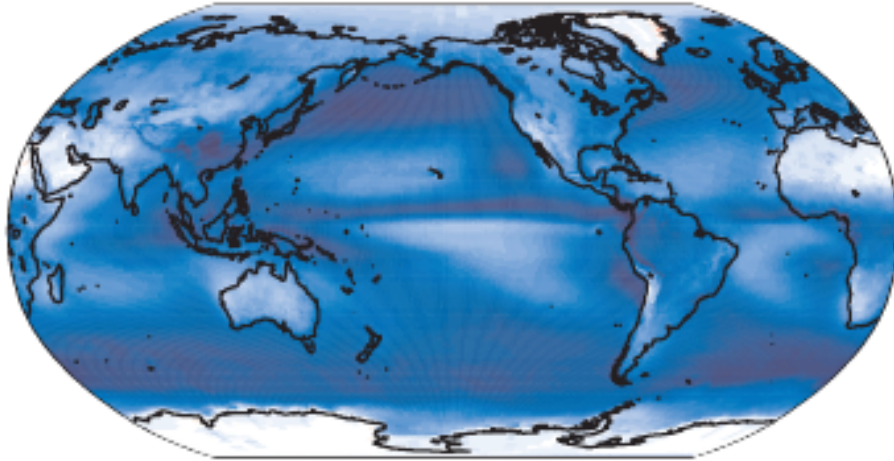
Climate Feedbacks: Cannot be Directly Observed



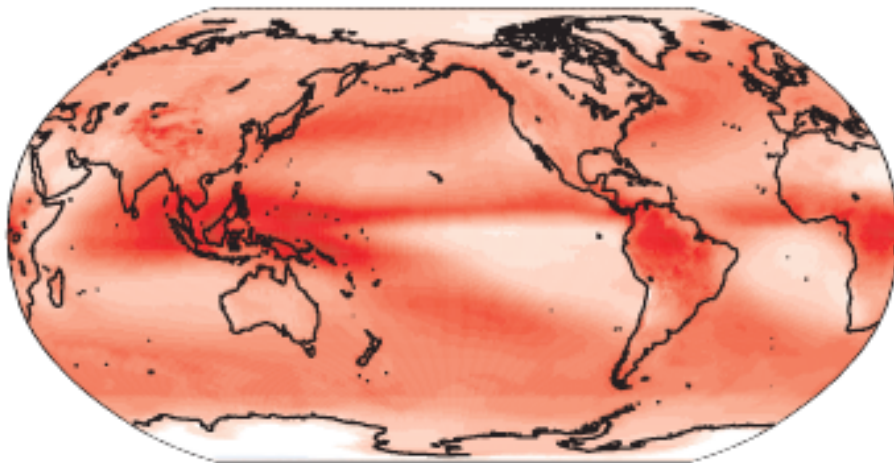
IPCC, 2013 (Ch 9, Hartmann et al 2013) Fig 9.43

Cloud Radiative Effects

(a) Shortwave (global mean = -47.3 W m^{-2})

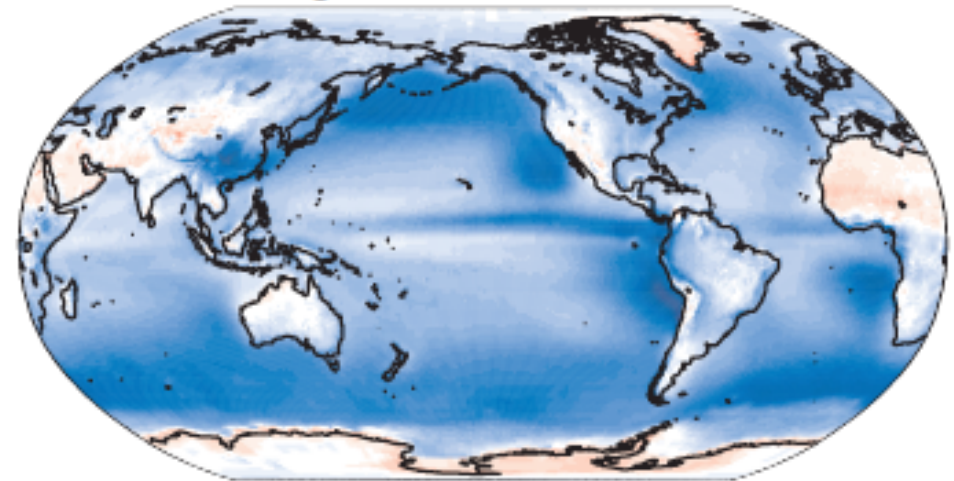


(b) Longwave (global mean = 26.2 W m^{-2})

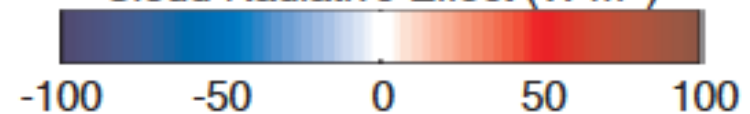


$$R_{\text{cloudy}} - R_{\text{clear}}$$

Net (global mean = -21.1 W m^{-2})

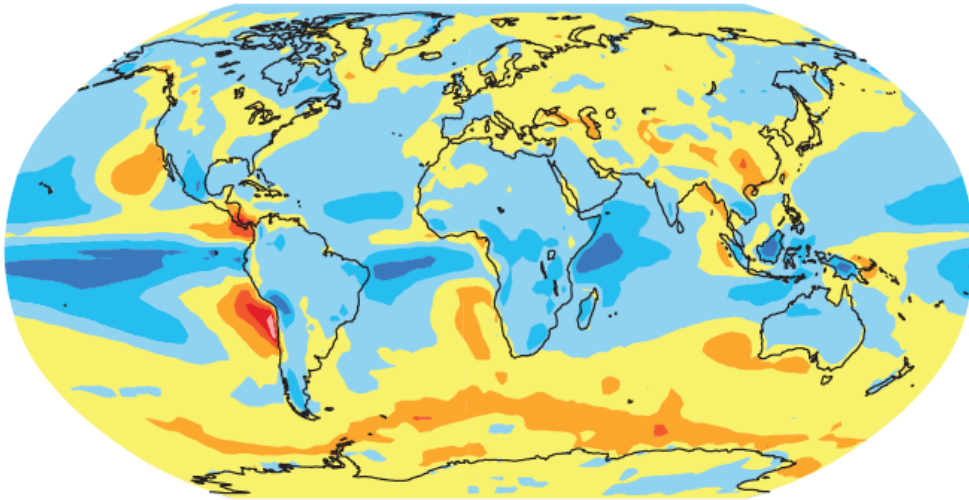


Cloud Radiative Effect (W m^{-2})

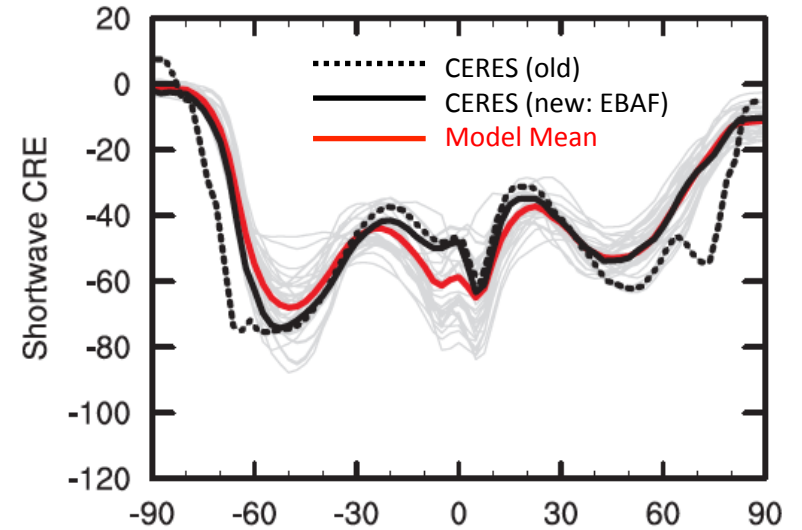


S. Ocean Cloud Biases: CMIP5

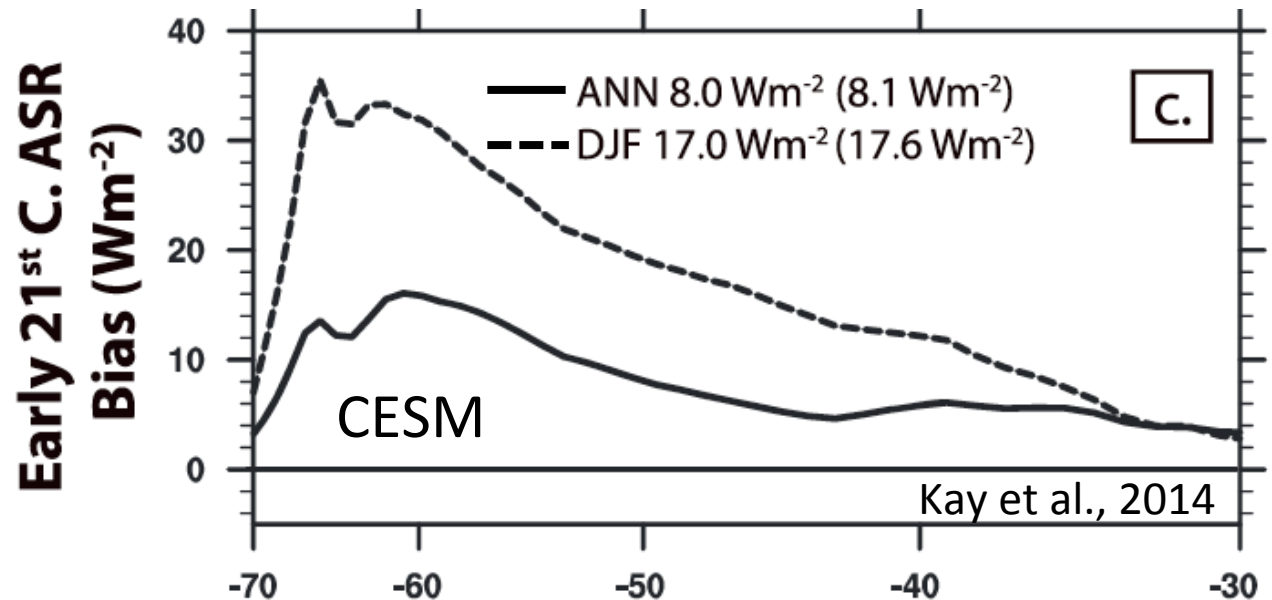
(a) Shortwave cloud radiative effect - MOD-OBS



(d) zonal average of shortwave CRE



IPCC AR5, 2013: Fig 9.5

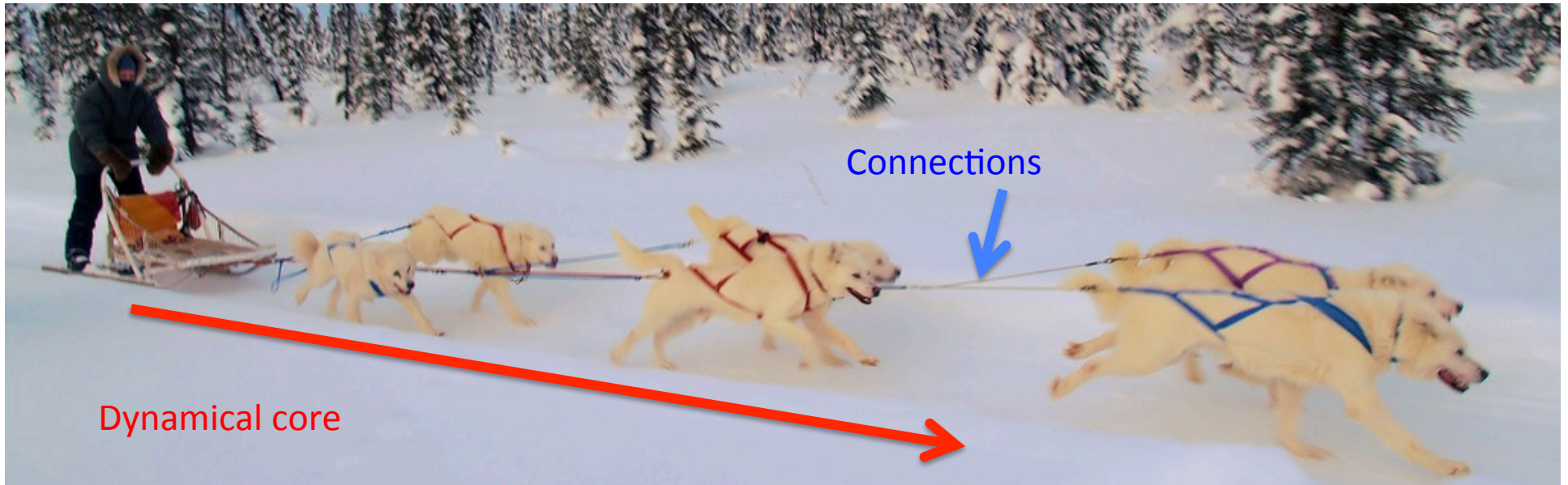


How do we represent clouds in models?

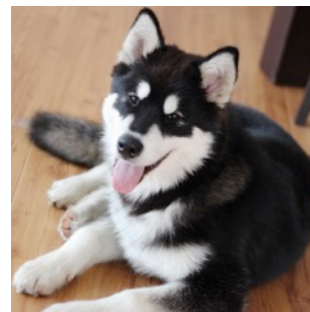
- Turbulence parameterizations [Yesterday]
 - Deep, Shallow convection
 - Boundary layer turbulence (moist)
 - Condensation, Large scale closure [Richard]
 - Microphysics (precipitation) [Jason]
- + Lots of Plumbing

Current state of cloud Schemes

Model developer just hanging on!



Parameterizations
(Tendency Generators)



Deep Convection



Microphysics



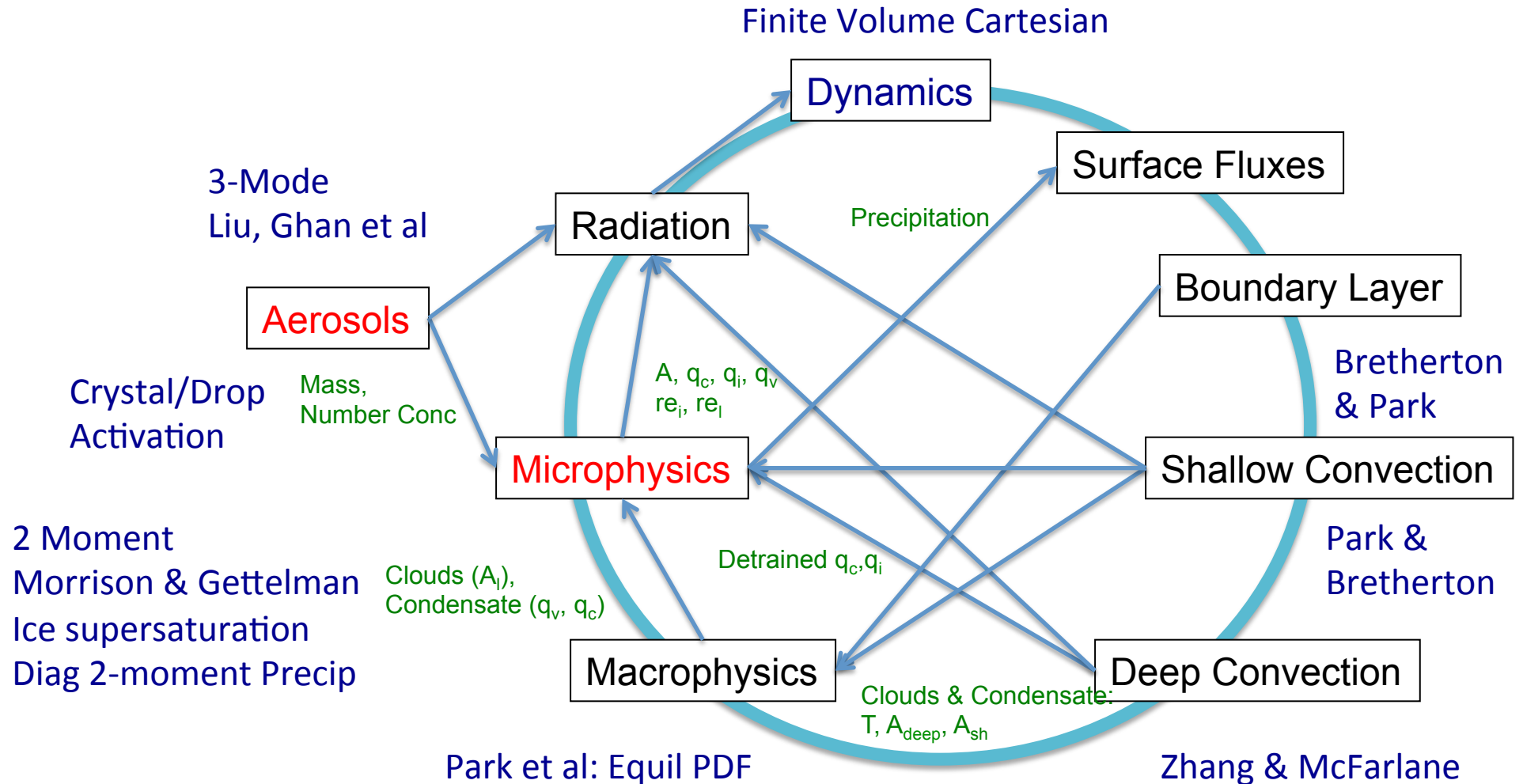
Condensation
/Fraction

Building a unified cloud scheme

- Can we do better with clouds?
- Connections (plumbing) are often a problem
- Goal: more unified and scale-insensitive codes
 - Not “scale aware”: want to do the same thing at all scales
- How? Unified Schemes.

Community Atmosphere Model (CAM5)

CAM5.1-5.3: IPCC AR5 version (Neale et al 2010)

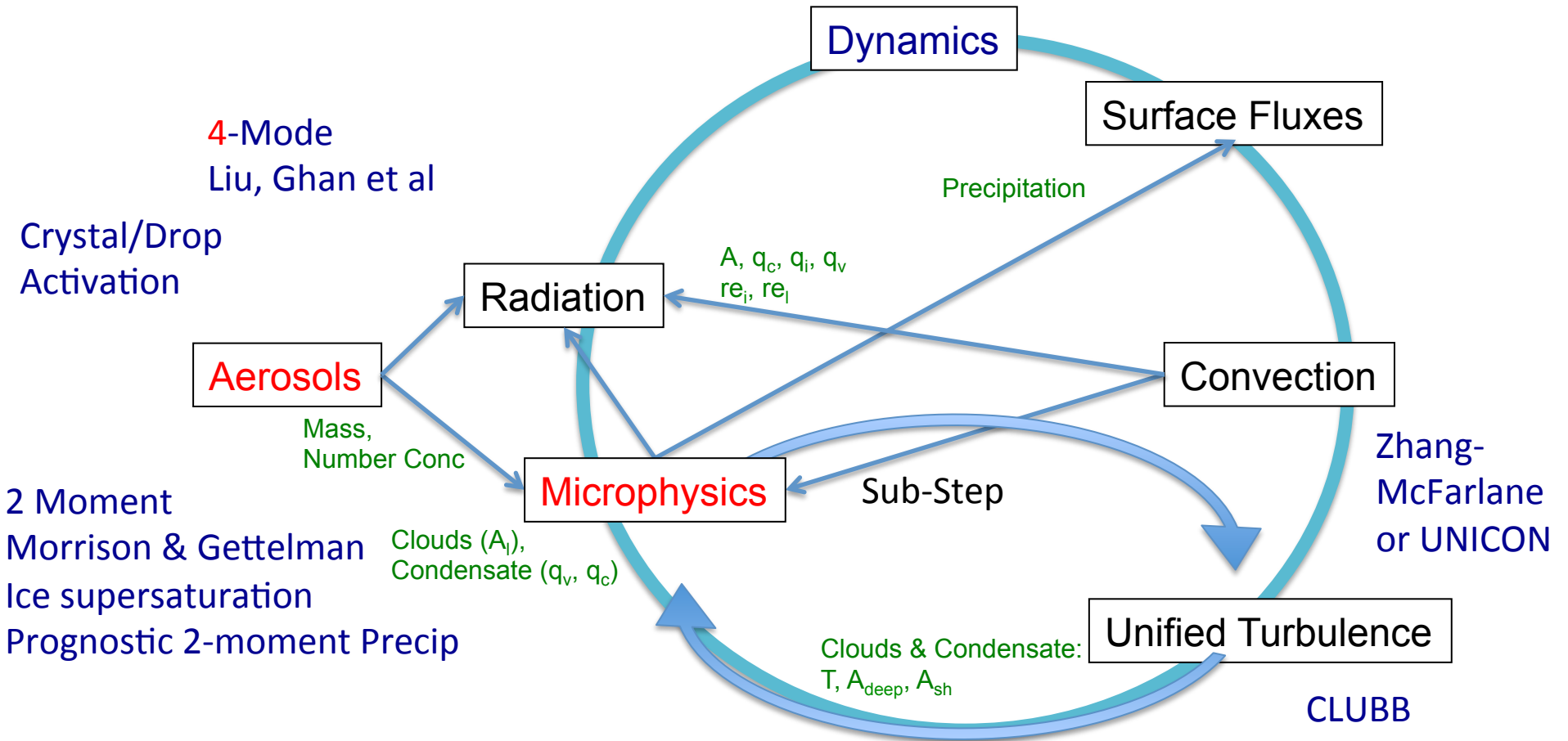


A = cloud fraction, q =H₂O, re =effective radius (size), T =temperature
(i)ce, (l)iquid, (v)apor

Community Atmosphere Model (CAM5.5)

Baseline for CMIP6 model (CAM6)

Finite Volume Cartesian

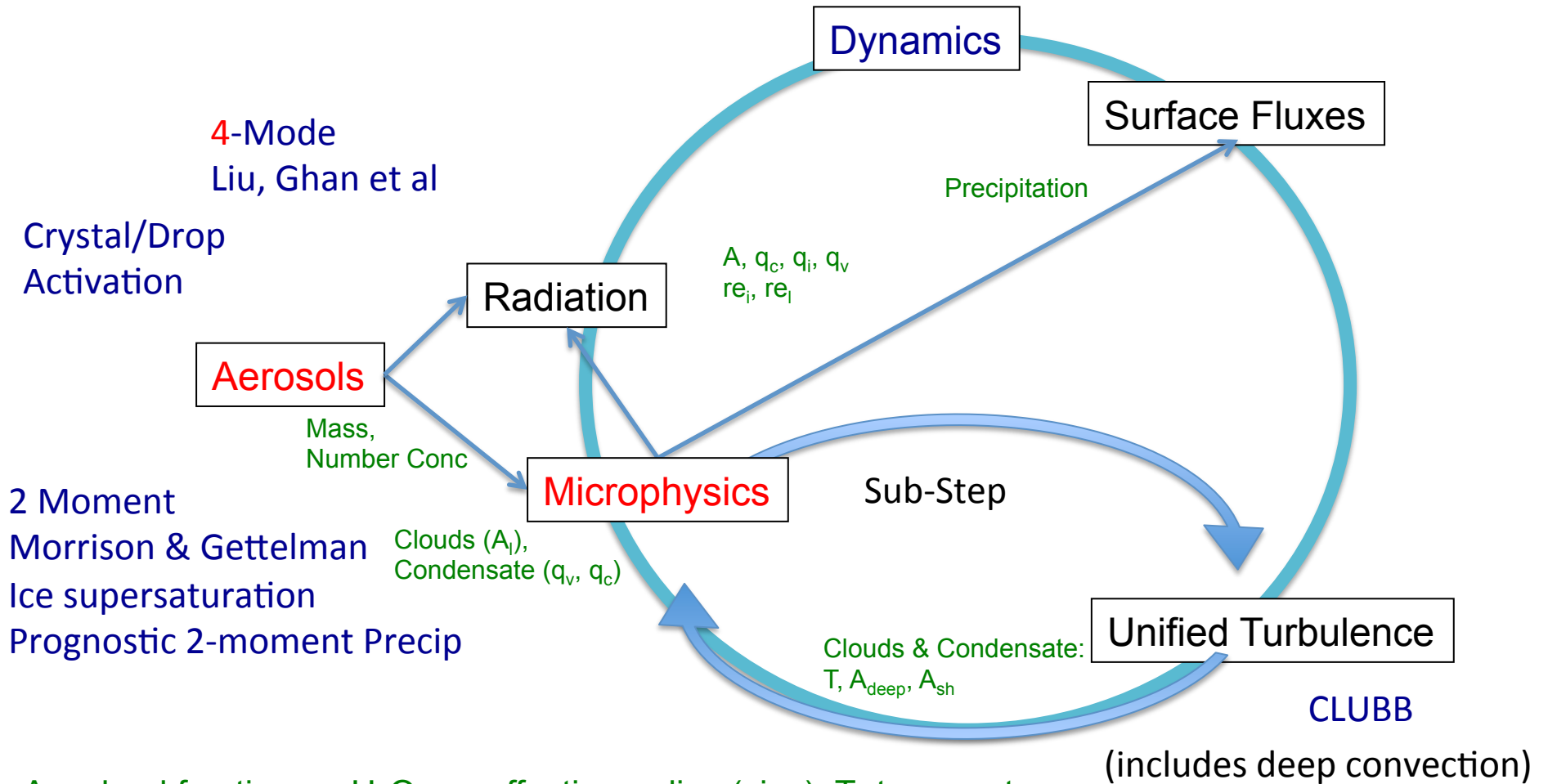


A = cloud fraction, q=H₂O, re=effective radius (size), T=temperature
(i)ce, (l)iquid, (v)apor

Community Atmosphere Model (0.25°)

Working on this as an option

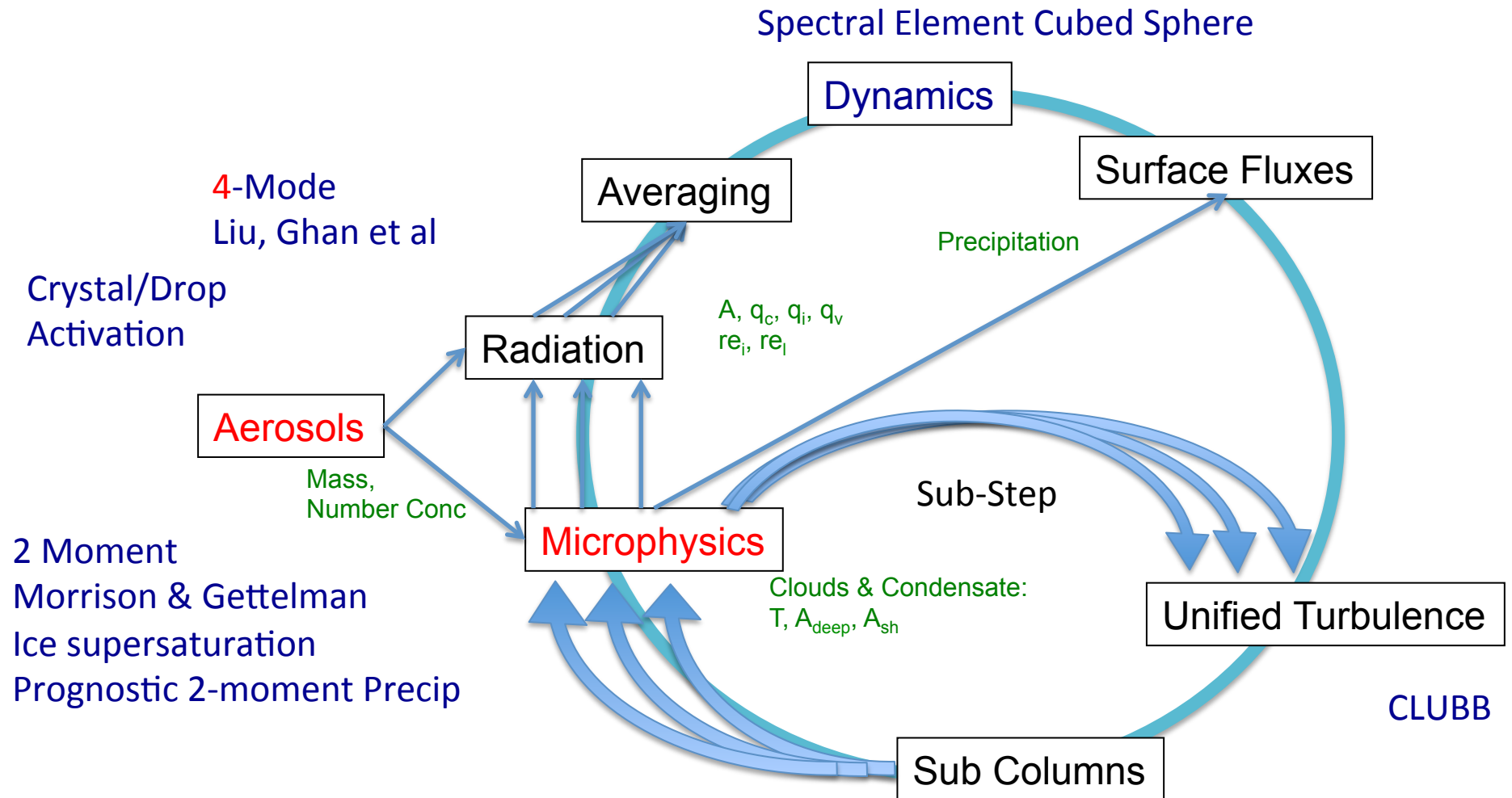
Spectral Element Cubed Sphere: Variable Resolution Mesh



A = cloud fraction, q =H₂O, re =effective radius (size), T =temperature
(i)ce, (l)iquid, (v)apor

Community Atmosphere Model (CAM5.X++)

Now in development: Sub-columns



A = cloud fraction, q =H₂O, re =effective radius (size), T =temperature
(i)ce, (l)iquid, (v)apor

Simulating and Observing Clouds



Ceci n'est pas une pipe.

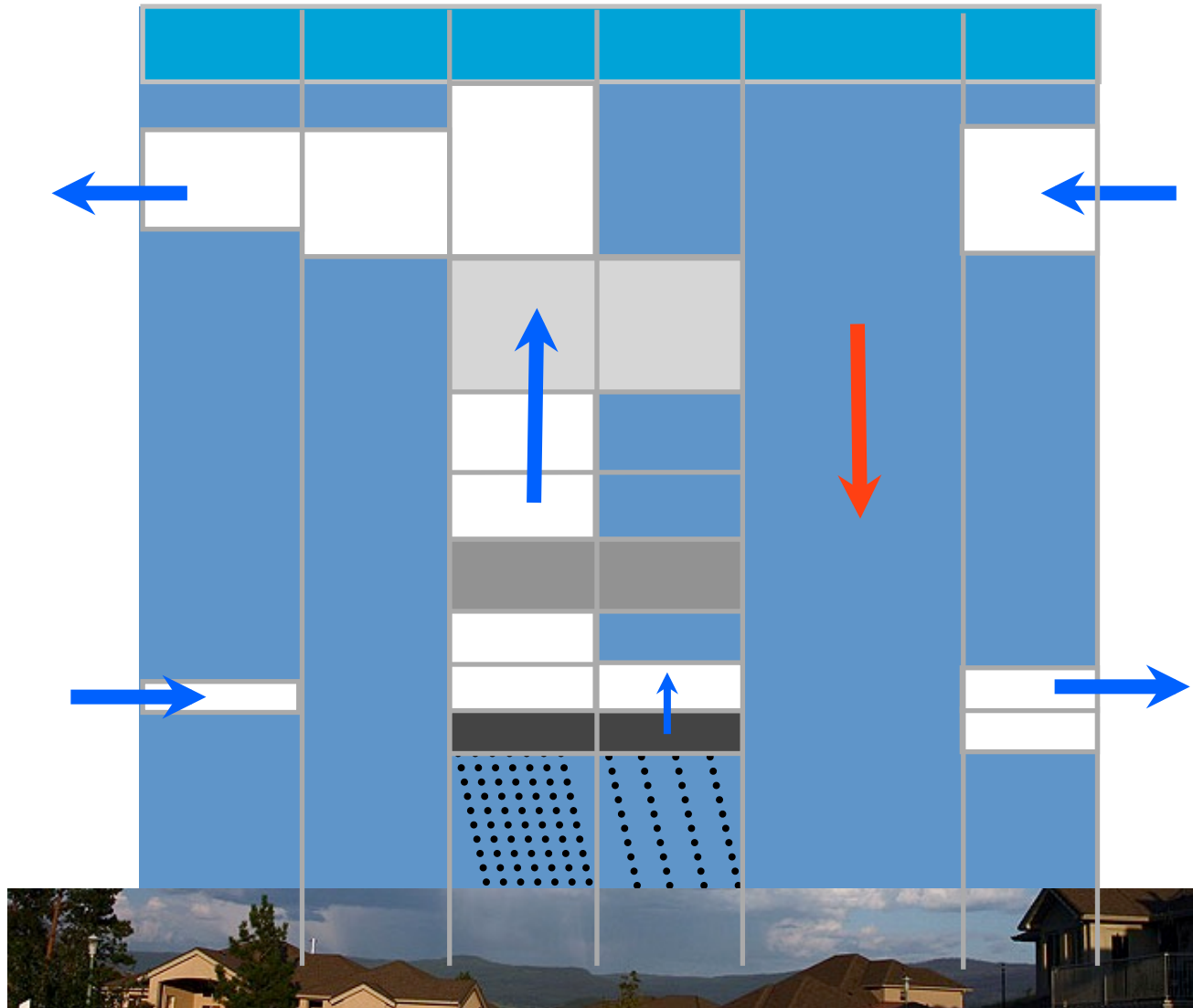
Ceci n'est pas un nuage



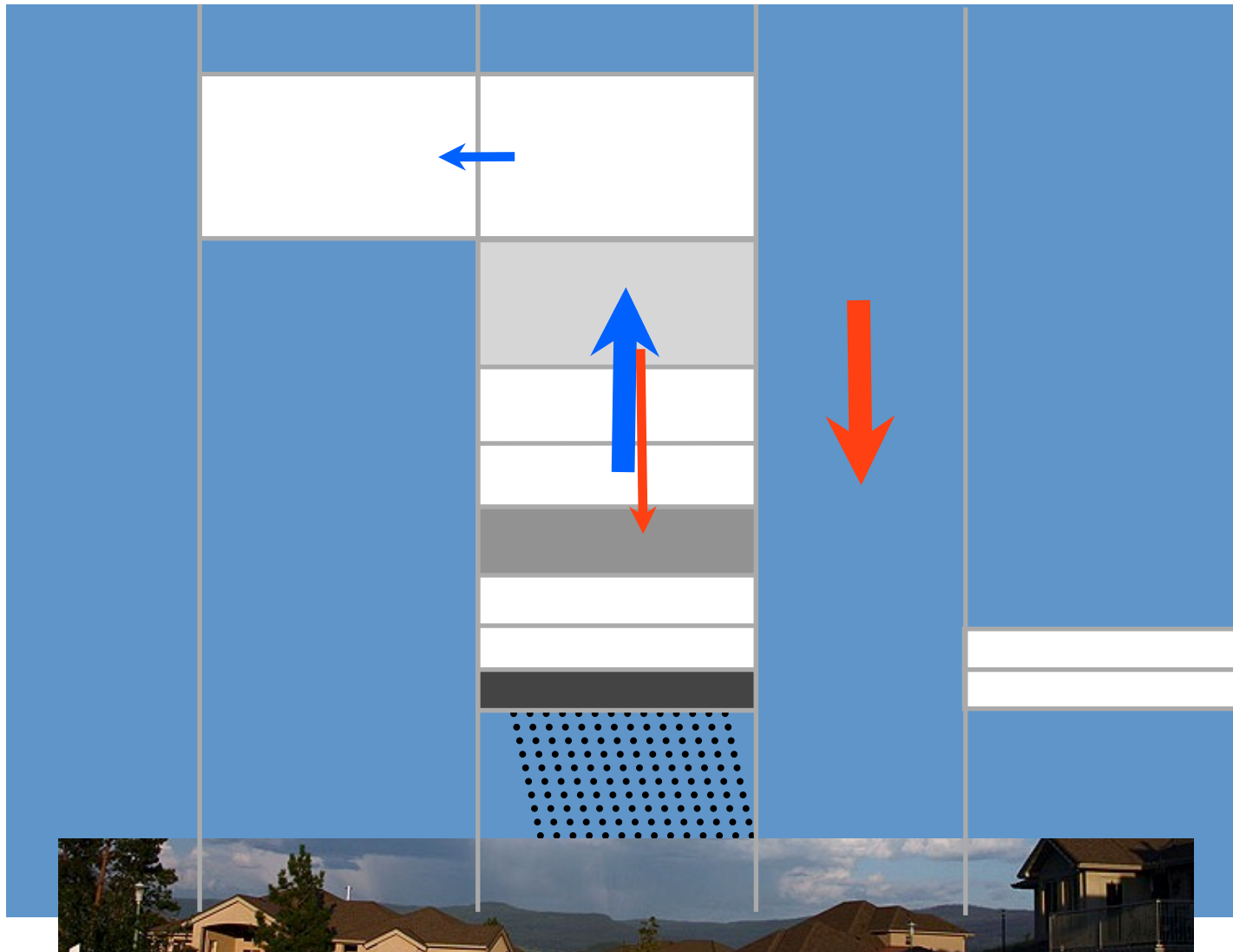
How does a model see clouds?



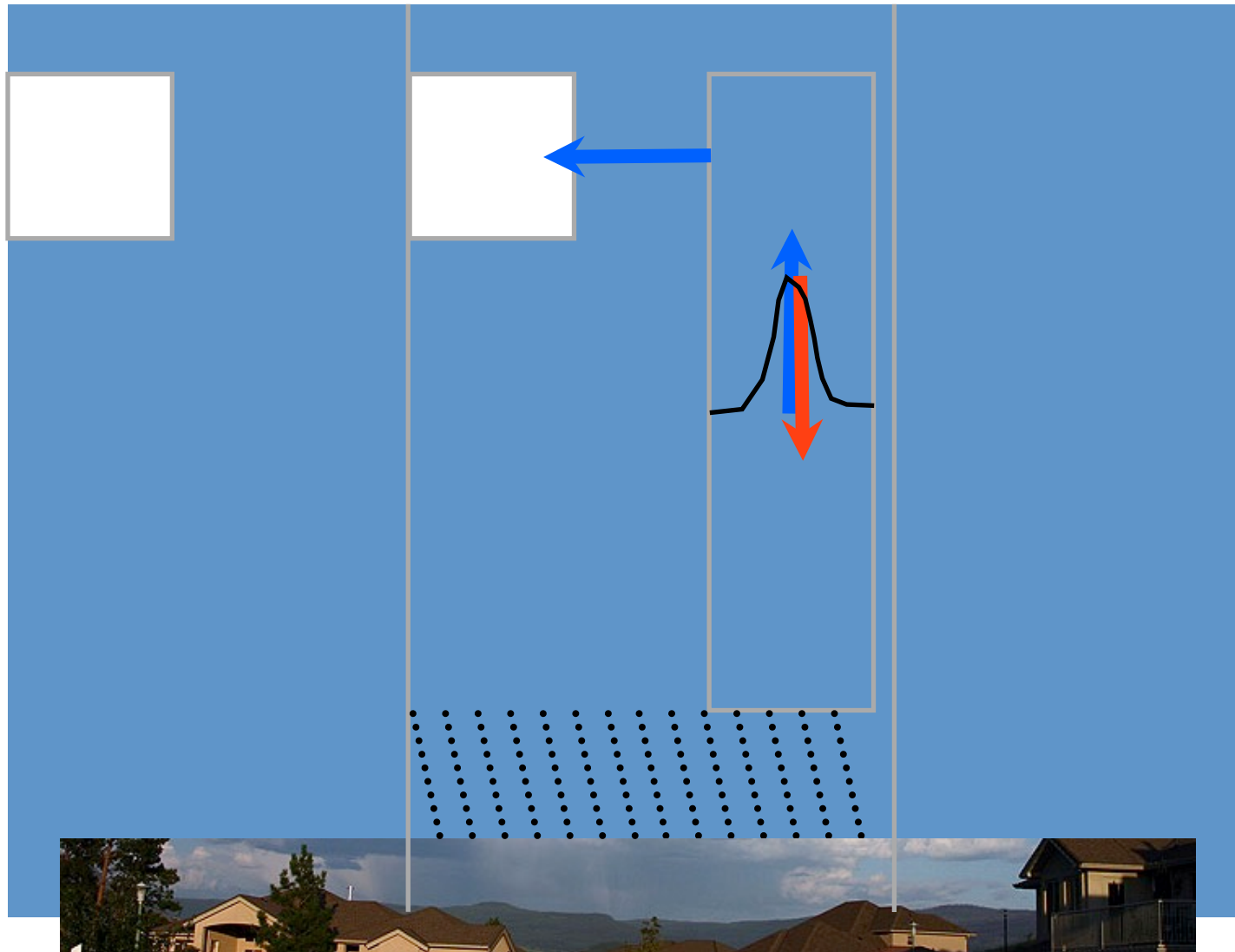
CRM/LES Convective Cloud



Mesoscale Convective Cloud



GCM Convective Cloud

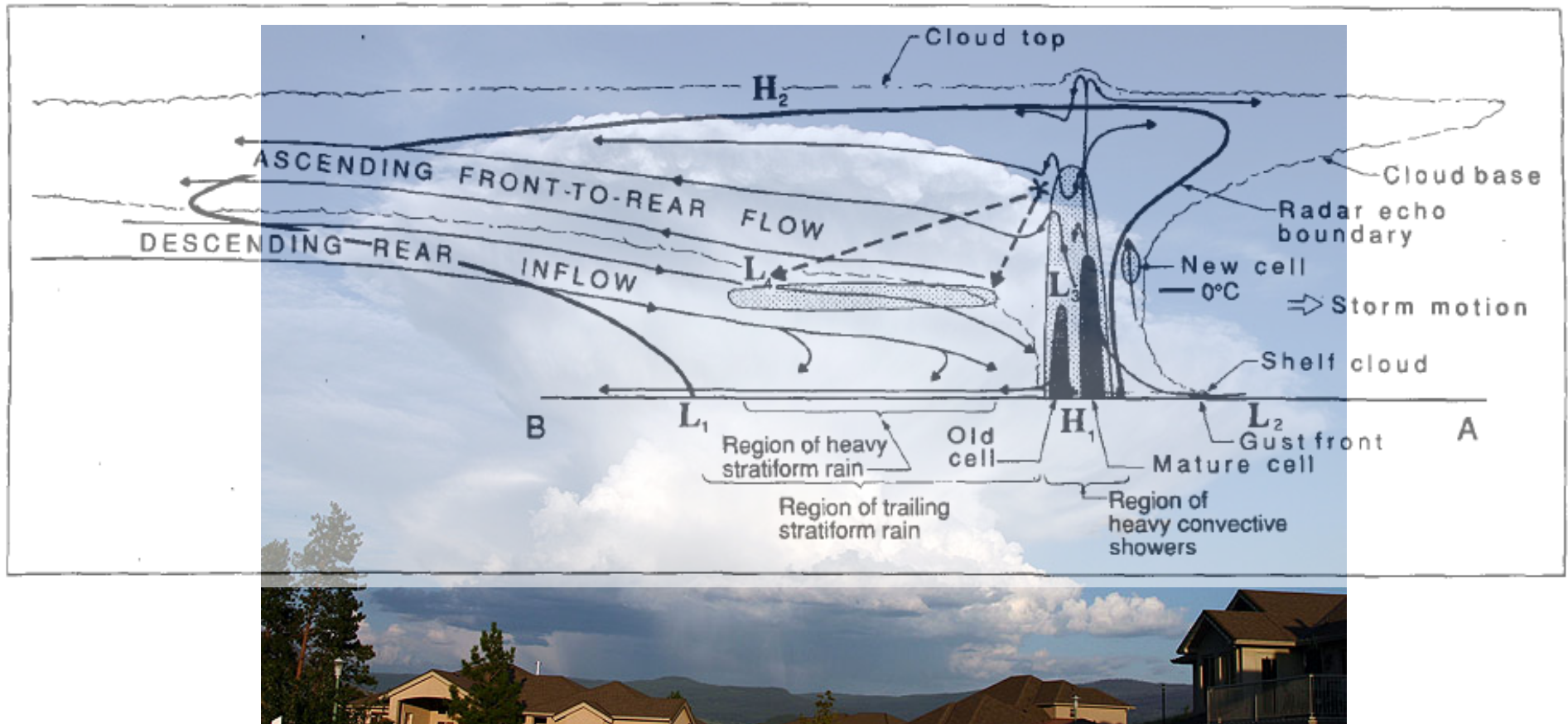


This is a (picture of a) Cloud

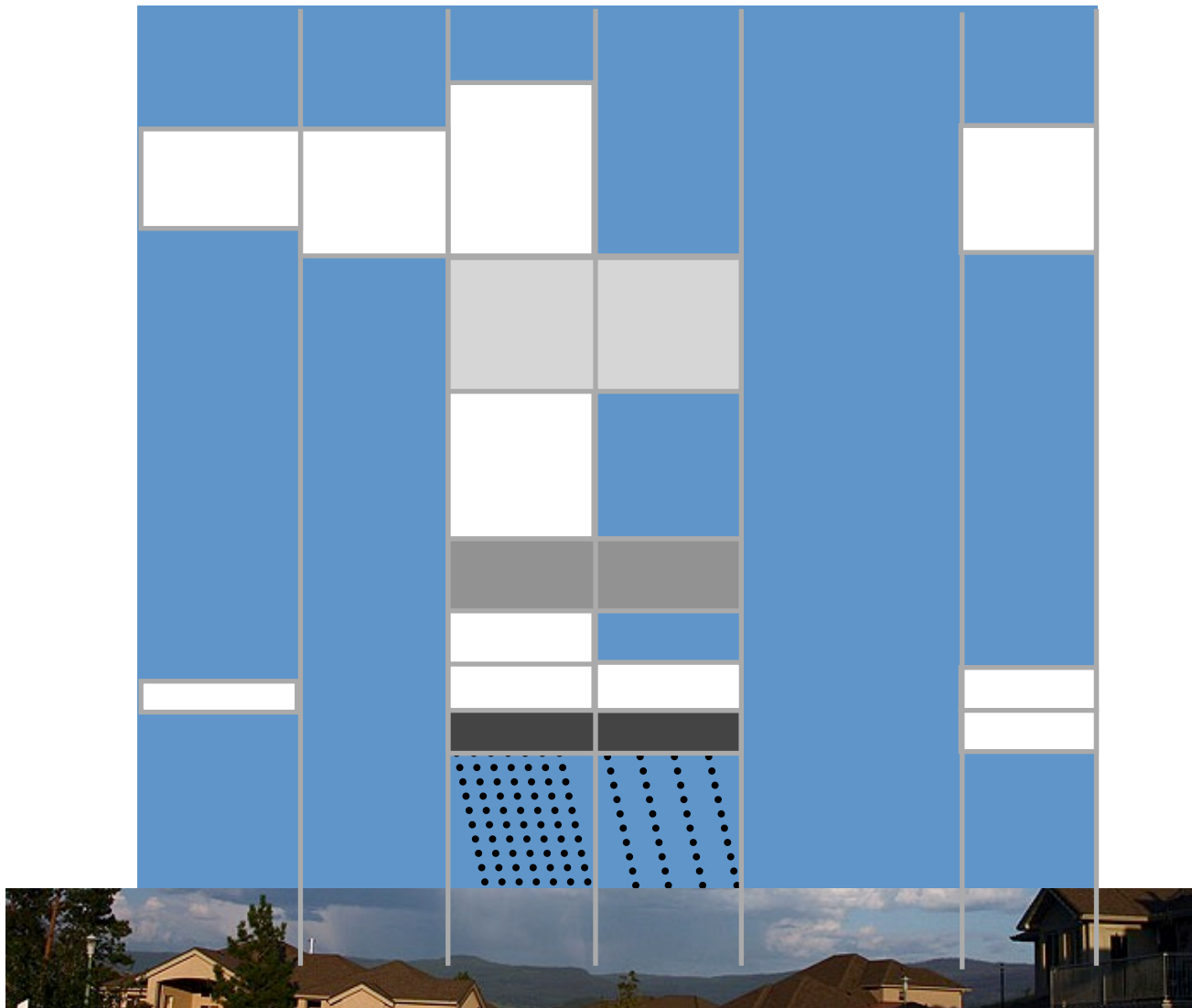
How do we observe it?



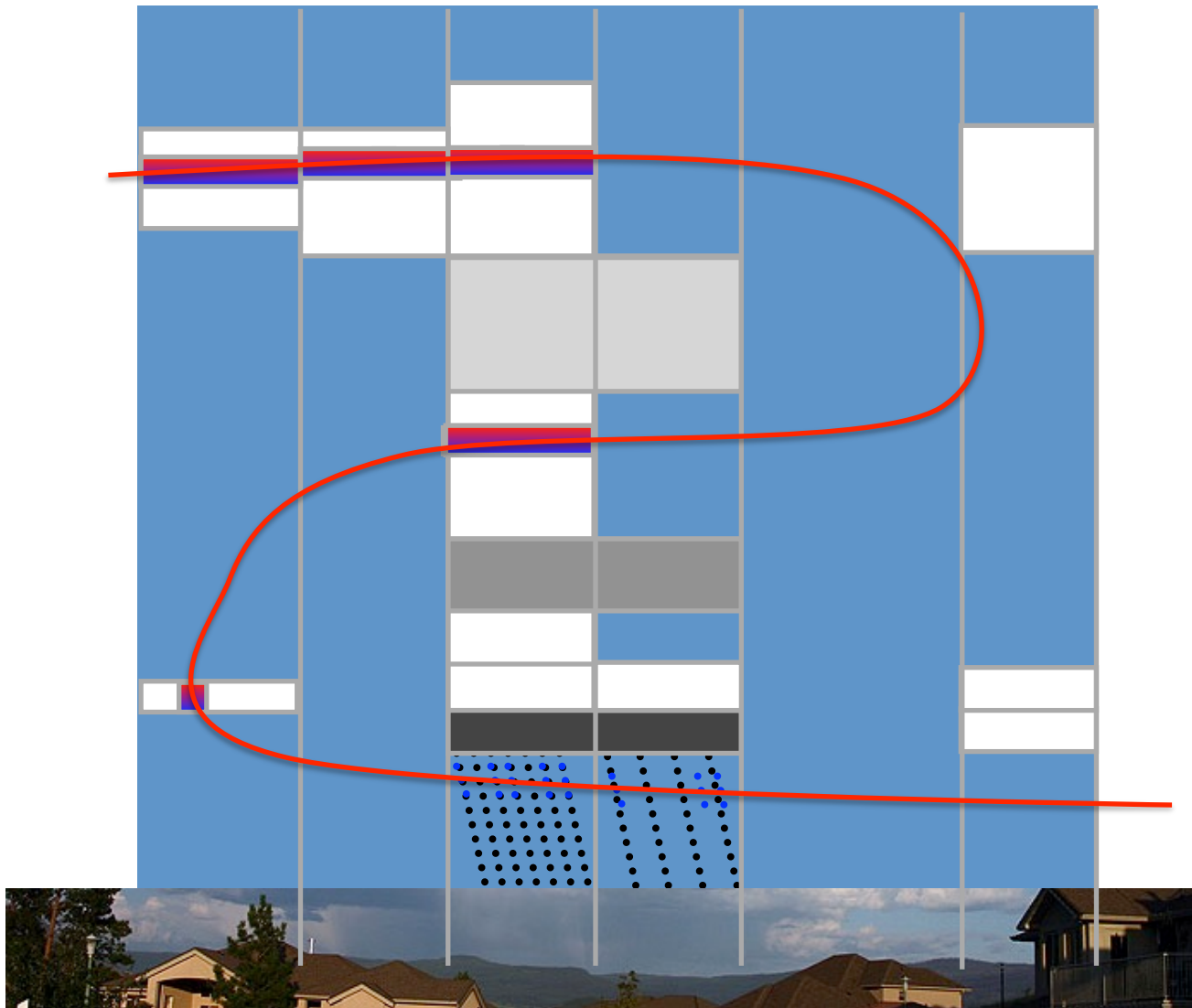
Conceptual Picture



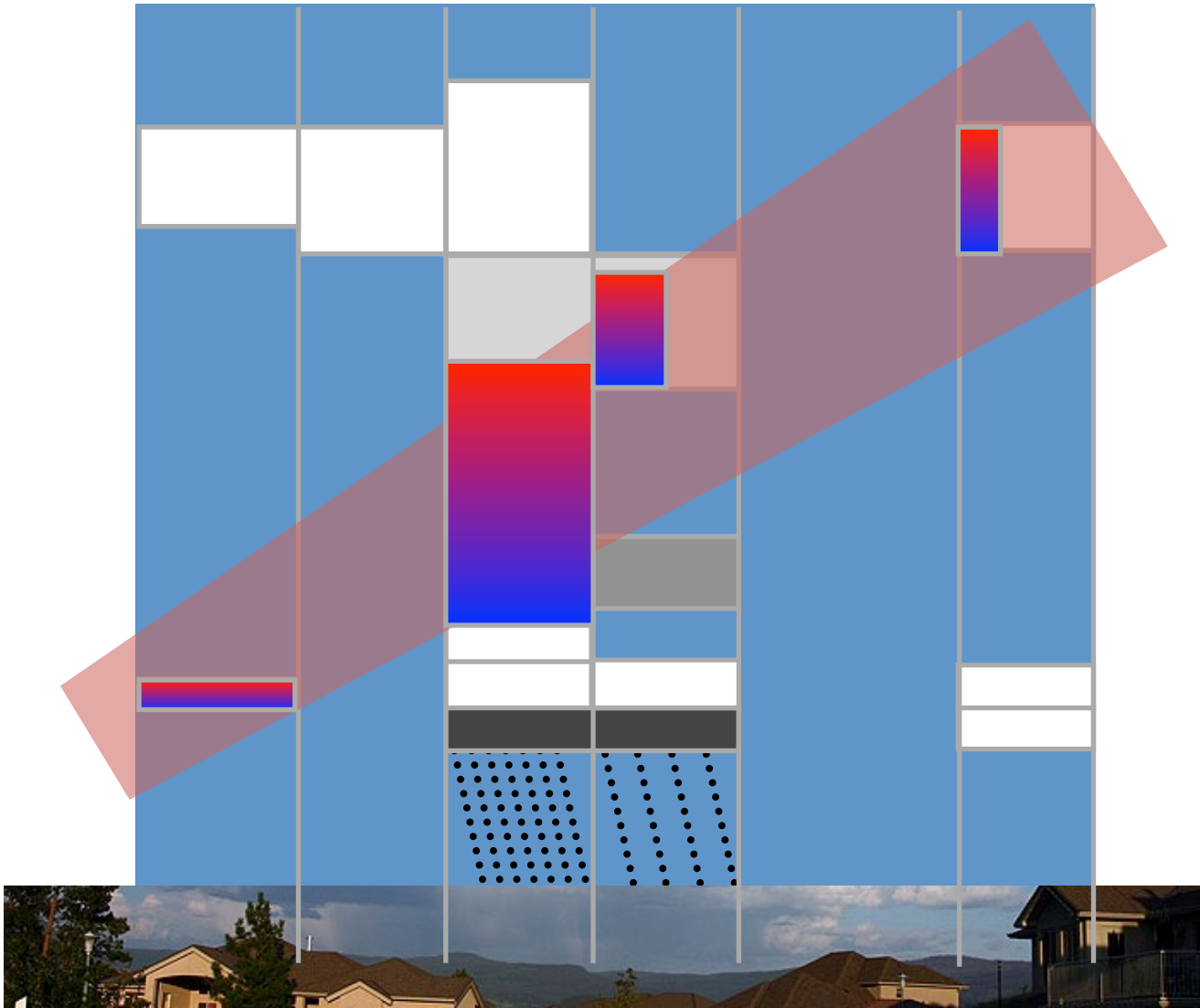
Convective Cloud



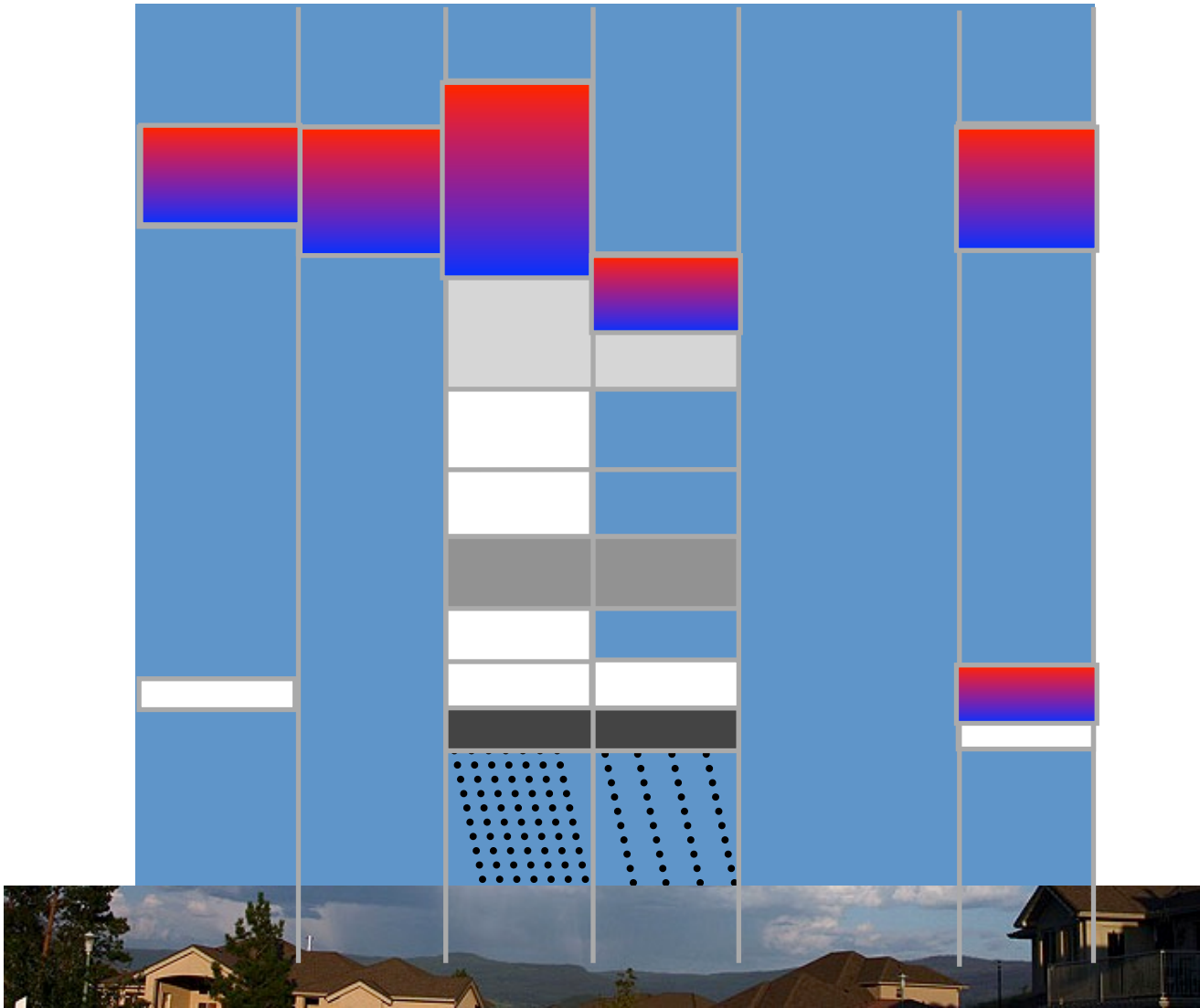
Aircraft Convective Cloud



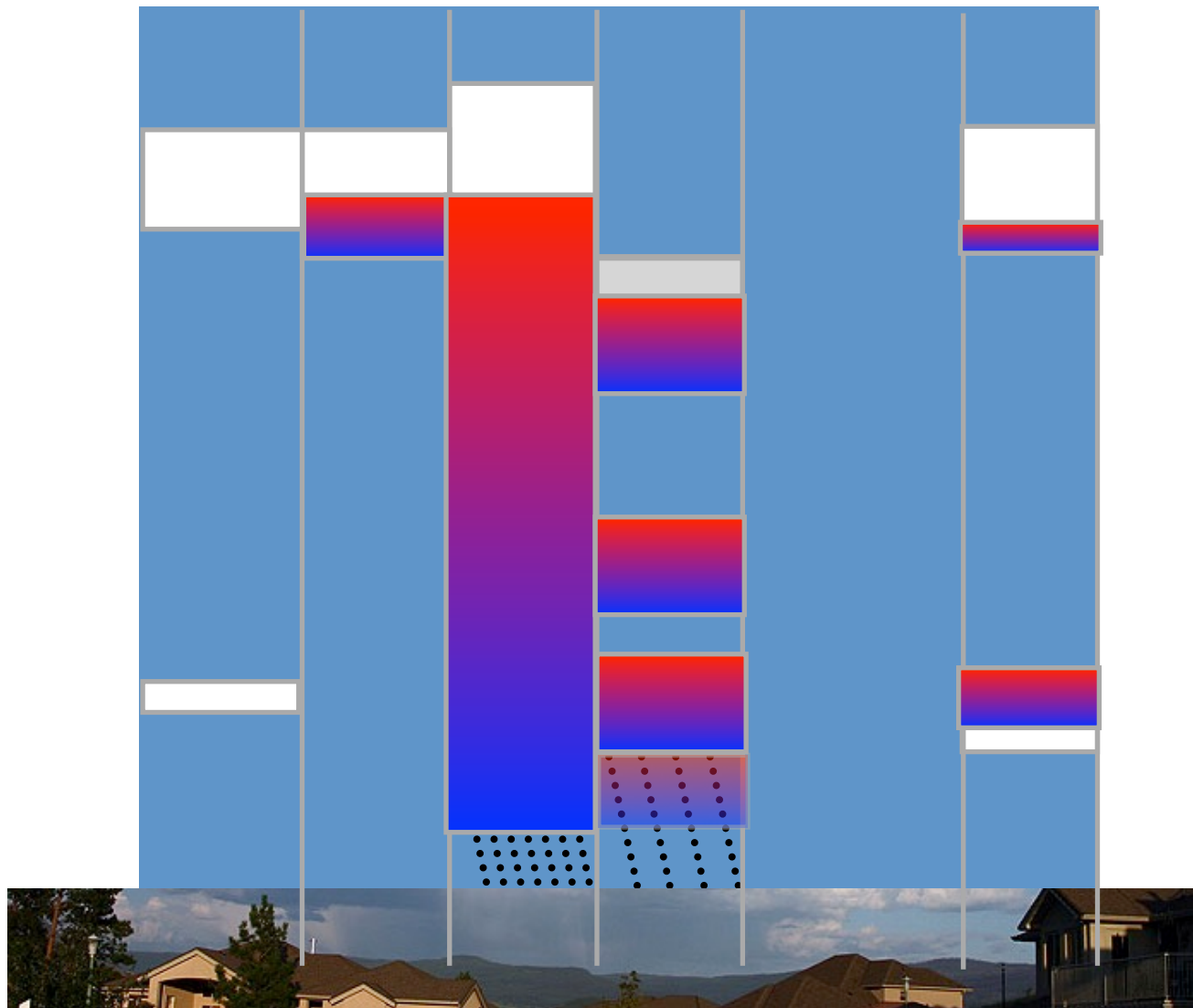
Radar Convective Cloud



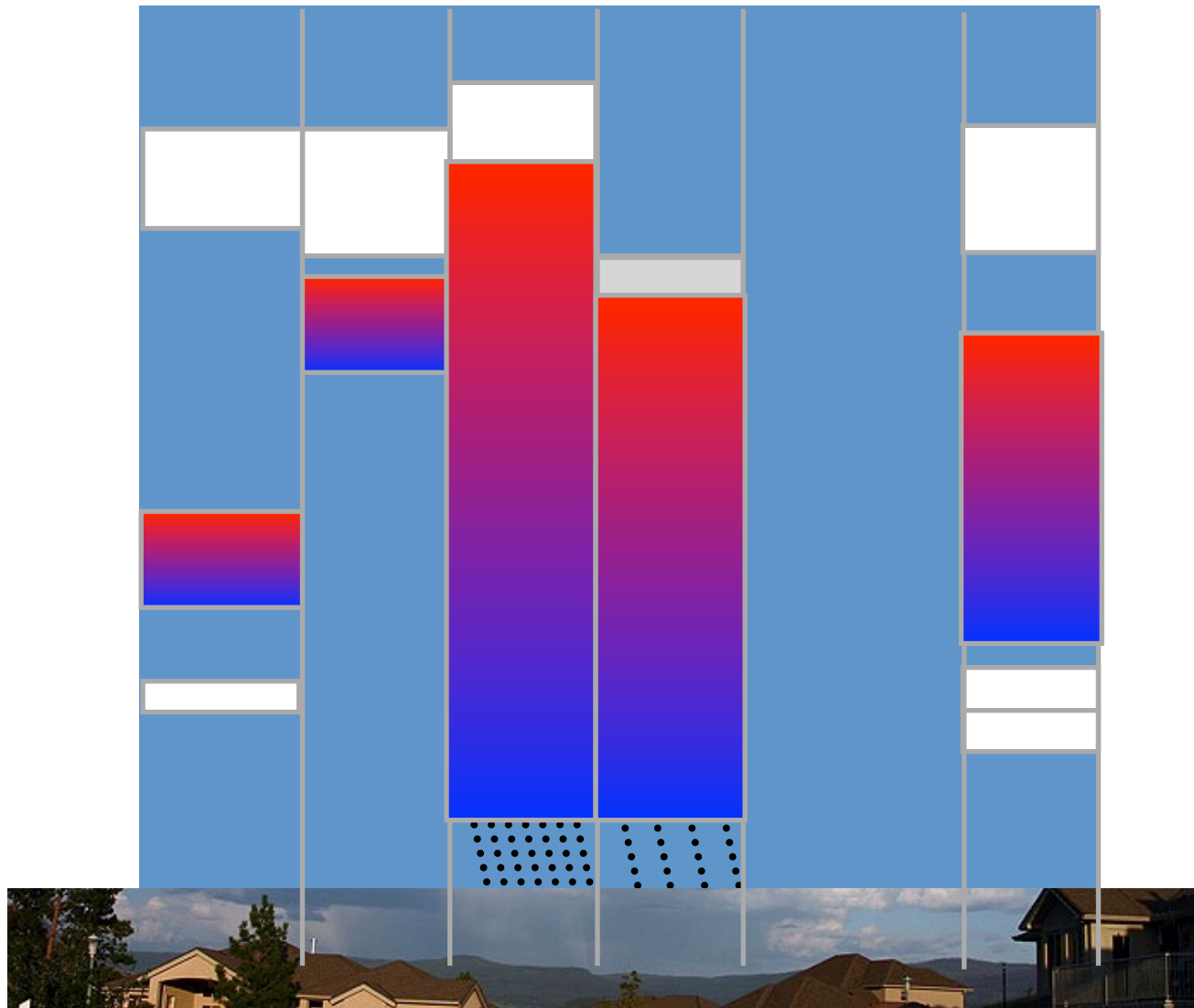
Lidar (CALIPSO) Convective Cloud



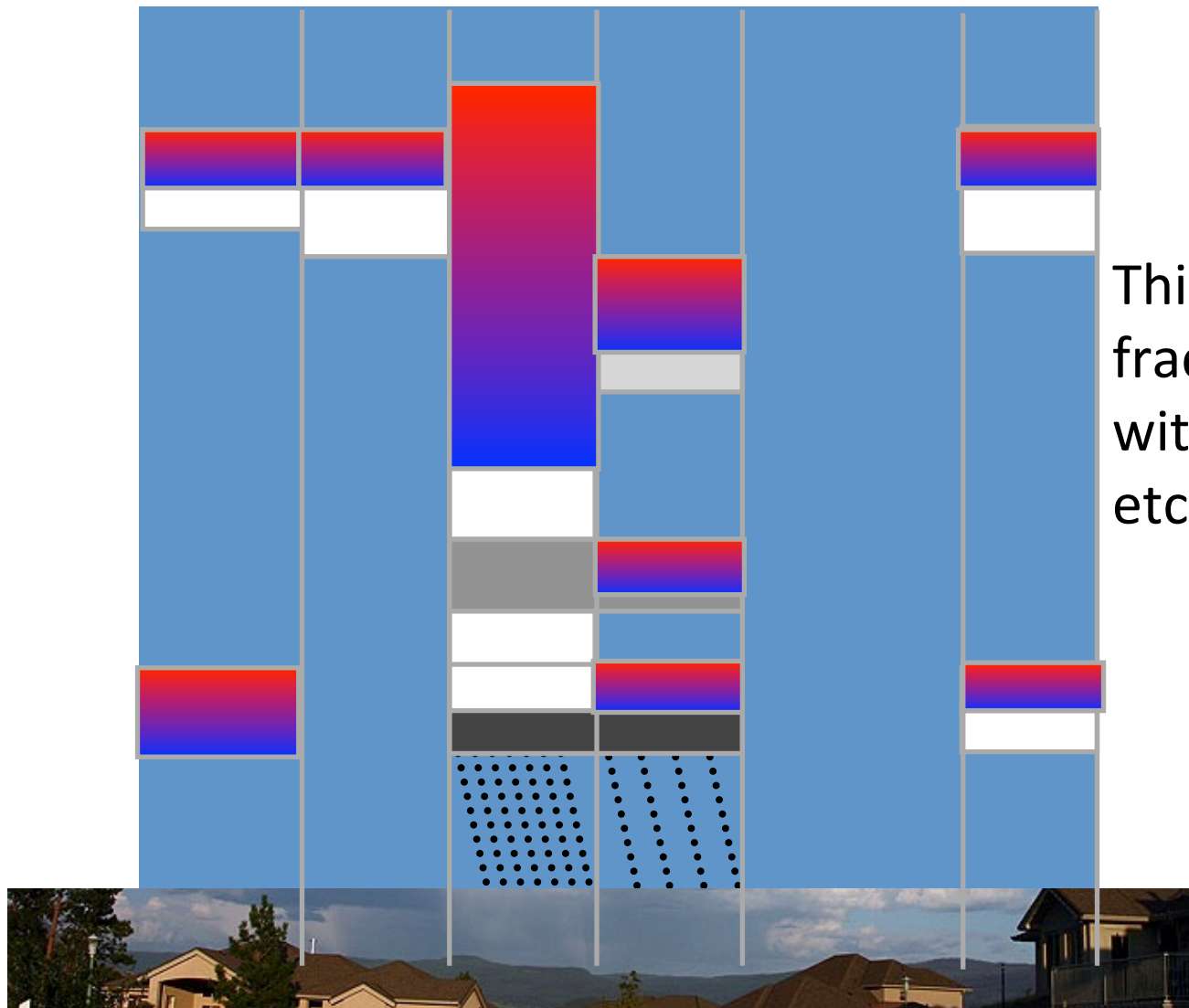
Radar (CloudSat) Convective Cloud



IR (MODIS) Convective Cloud



Vis/IR (MISR) Convective Cloud



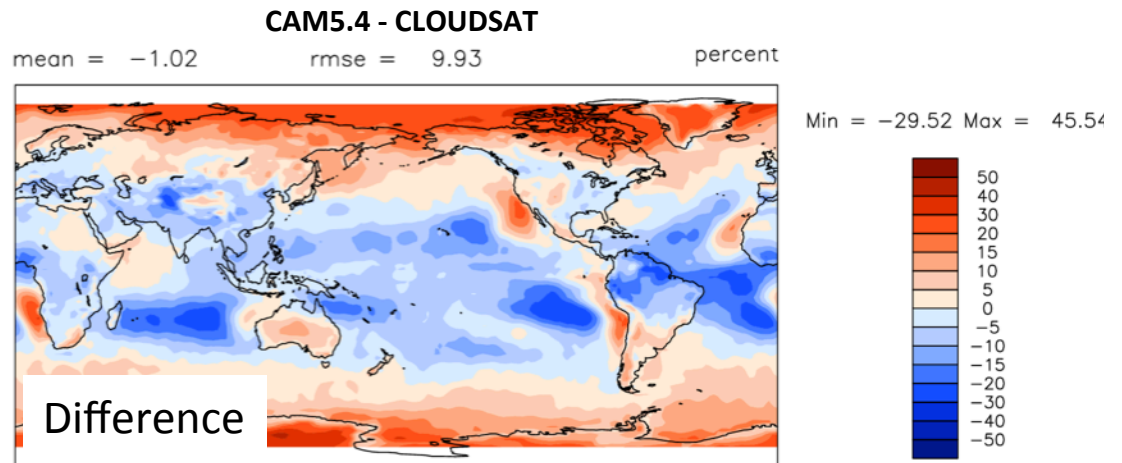
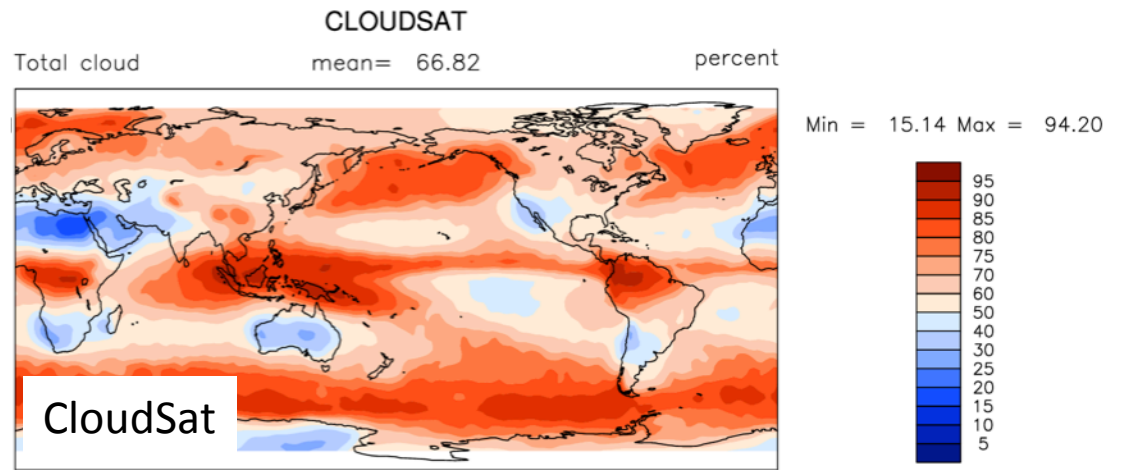
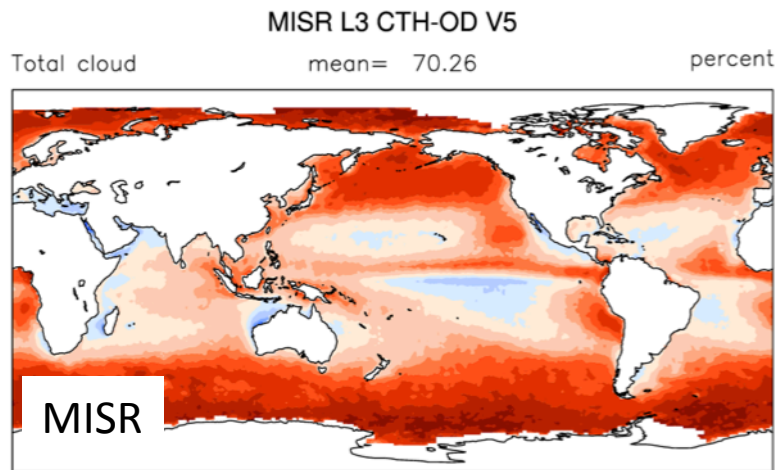
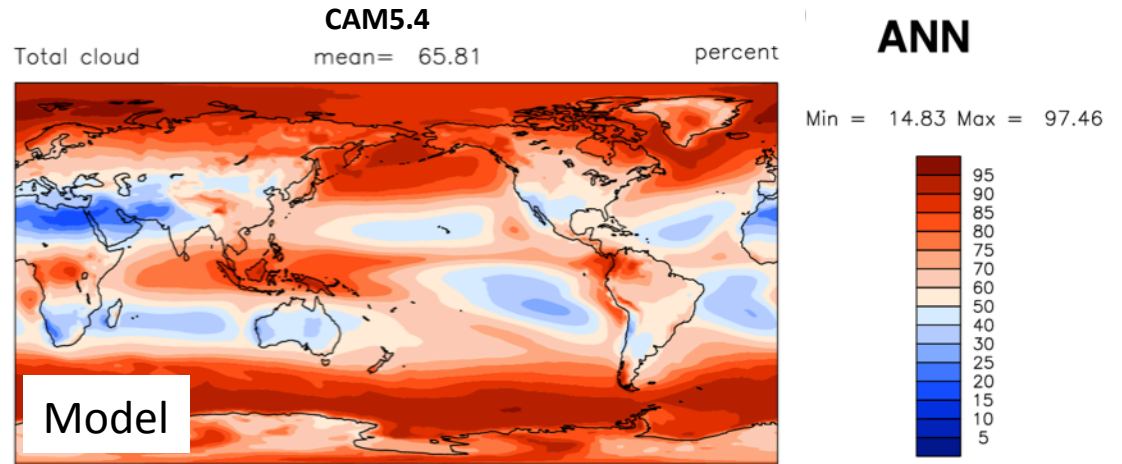
This is just cloud fraction: try it with LWP, re, etc !

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...

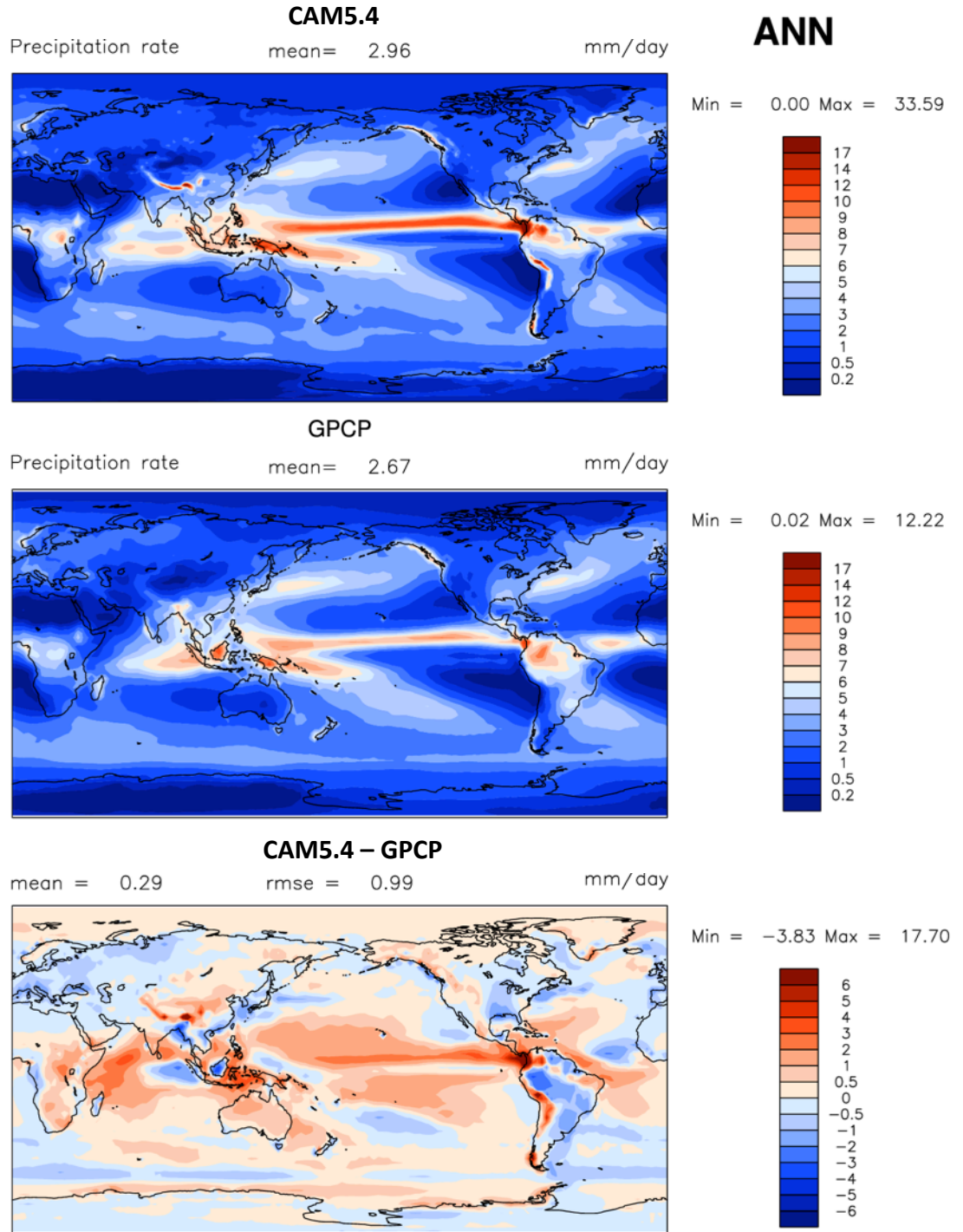
Traditional View: Model – CloudSat
(Radar+ Lidar)



Biases change with data set.
May even change sign!

Mean Metrics

Example: Precipitation



Weather: Forecast Verification

Brier (1950) skill score, or cost function

total precipitation

Brier skill score (3M running mean)

Europe N Africa (lat 25.0 to 70.0, lon -10.0 to 28.0)

T+96

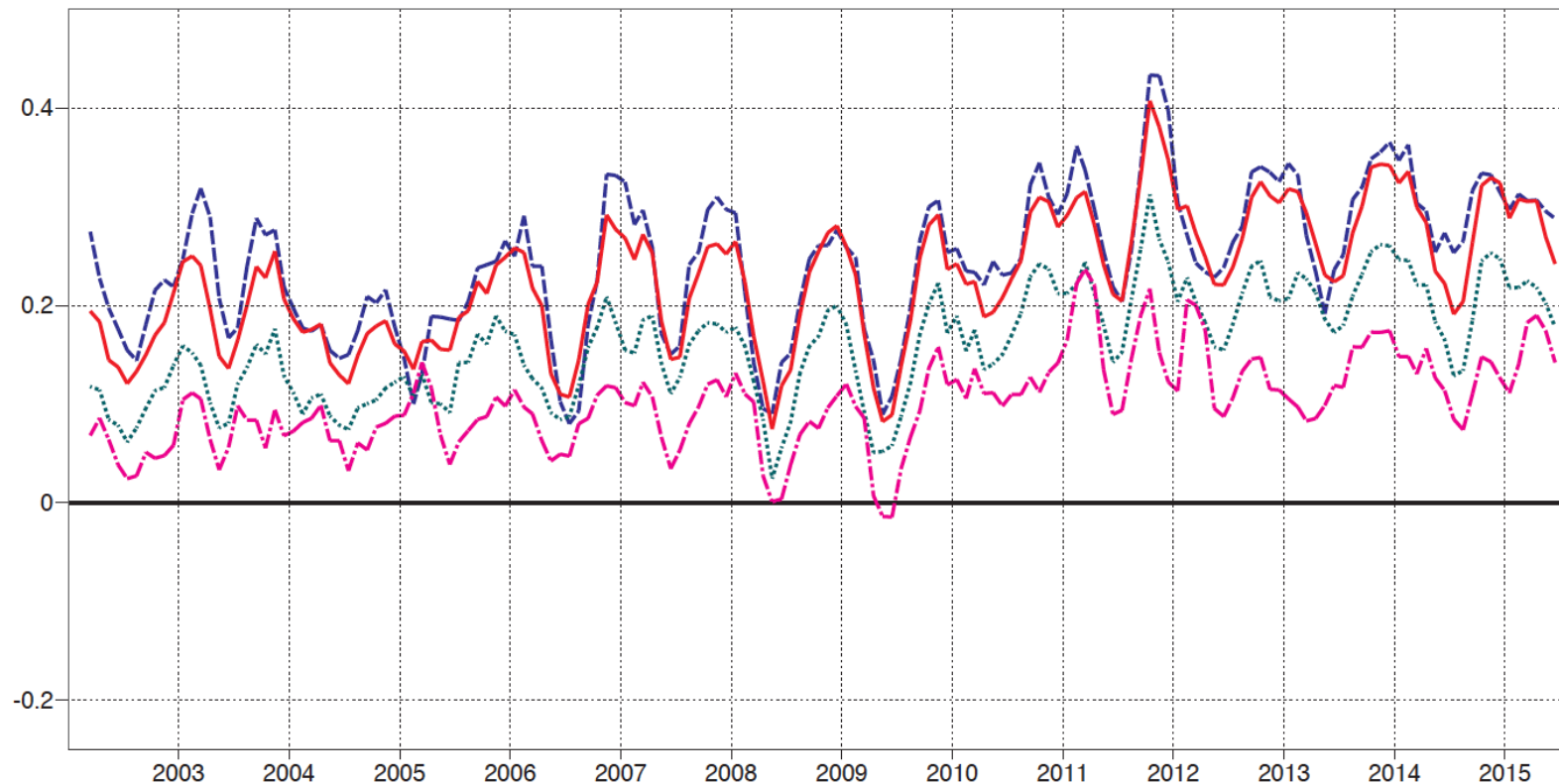
oper_ob_imported ti enfo prod I 12UTC

value >20.0

value >10.0

value >5.0

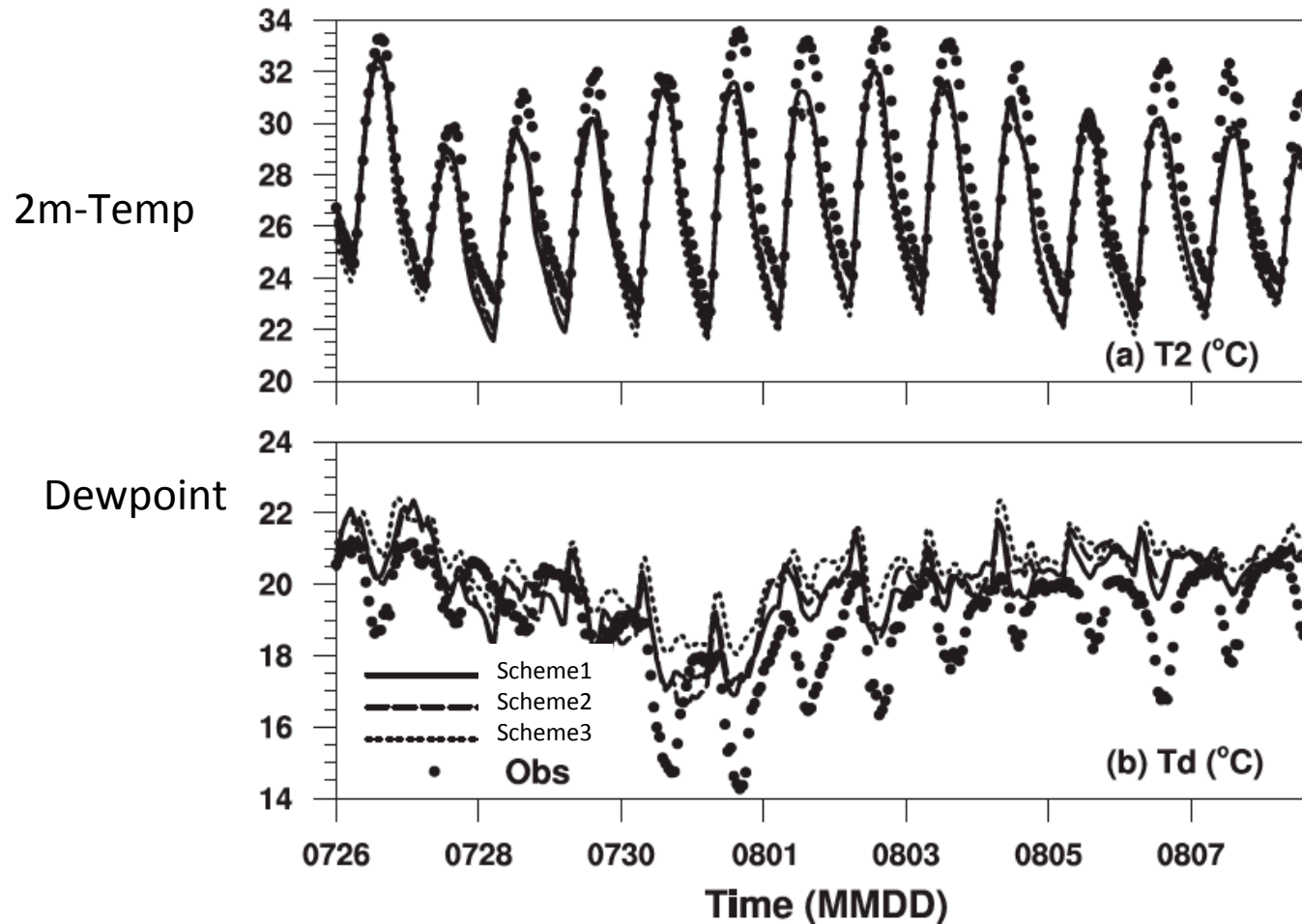
value >1.0



Weather Forecast Evaluation

Observed and simulated Temperature and Dewpoint
In a Mesoscale model over Texas

Votes for your favorite boundary layer scheme anyone?



Reference Redacted

Reference Redacted (see me later)

How do we do this better?

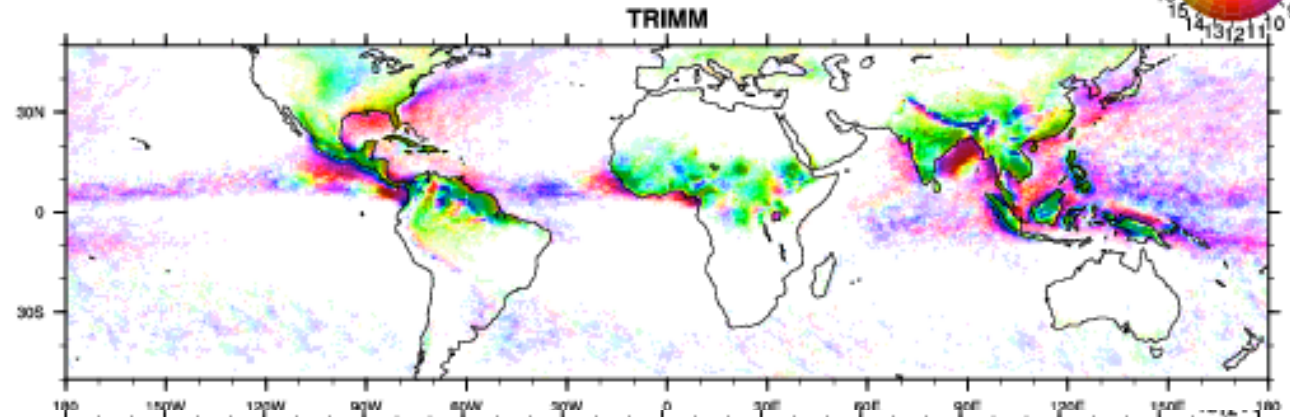
Issue: these evaluations are loosely related to specific processes.

- 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

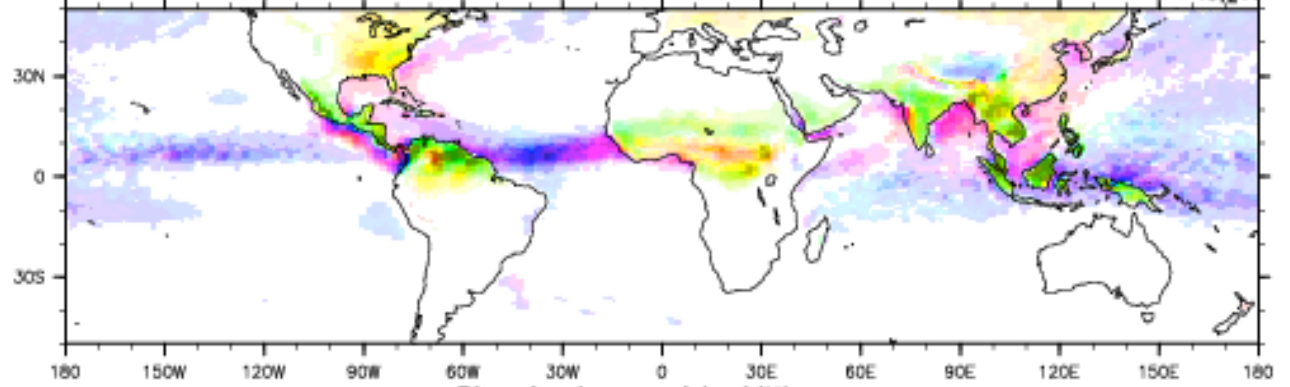
Variability: Diurnal Cycle of Precip



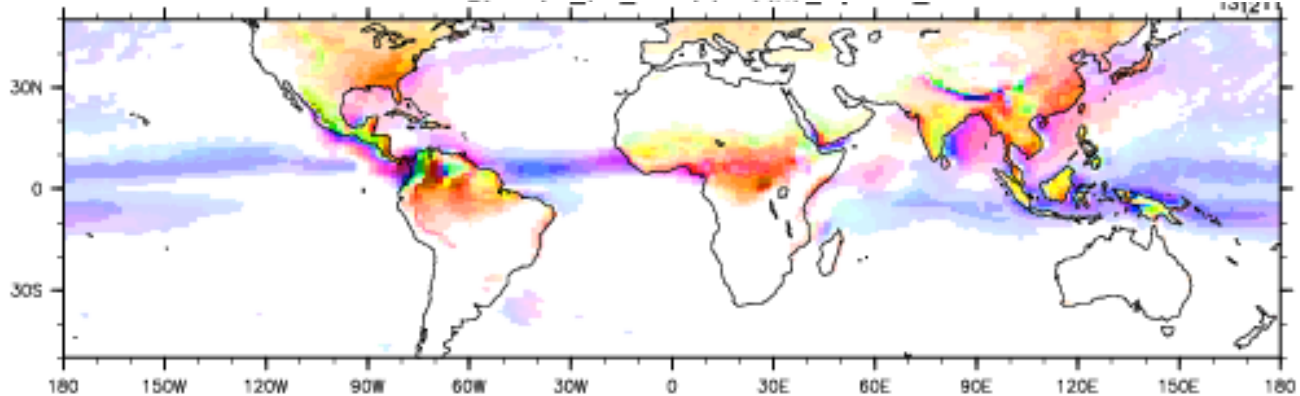
JJA Precipitation Rate
TRMM: OBS



CAM5.5 (new)



CAM5.3 (Old)

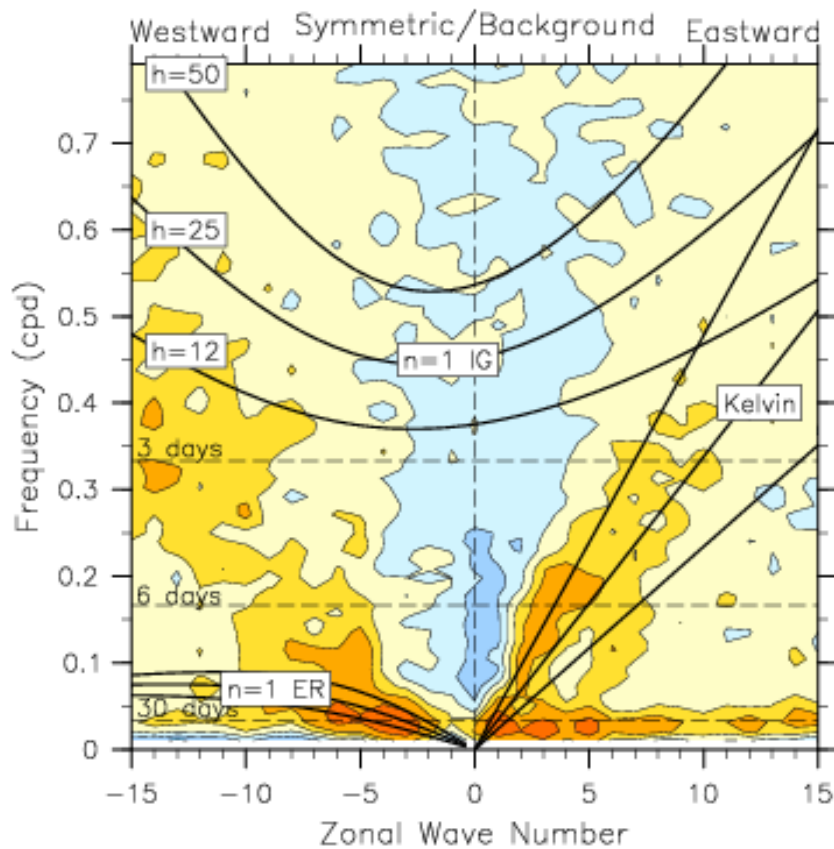


Thanks to R. Neale, C-C. Chen

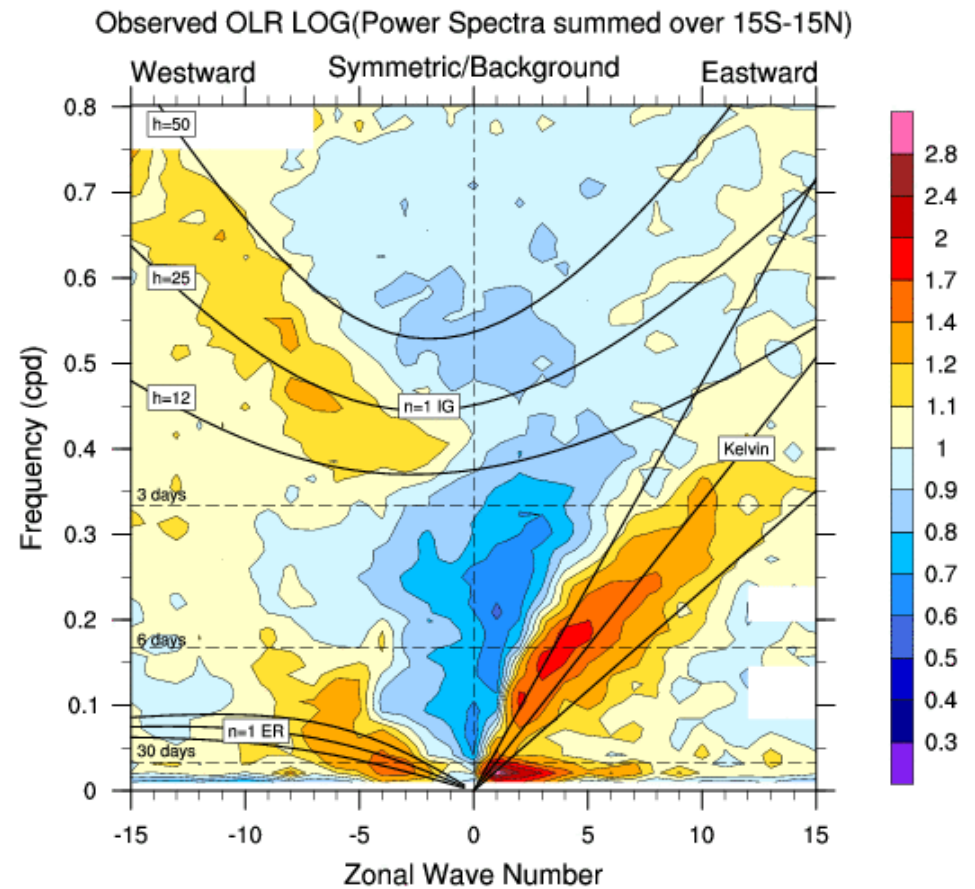
Variability: Tropical Waves

Symmetric OLR Spectrum (Wheeler & Kiladis, 1999)

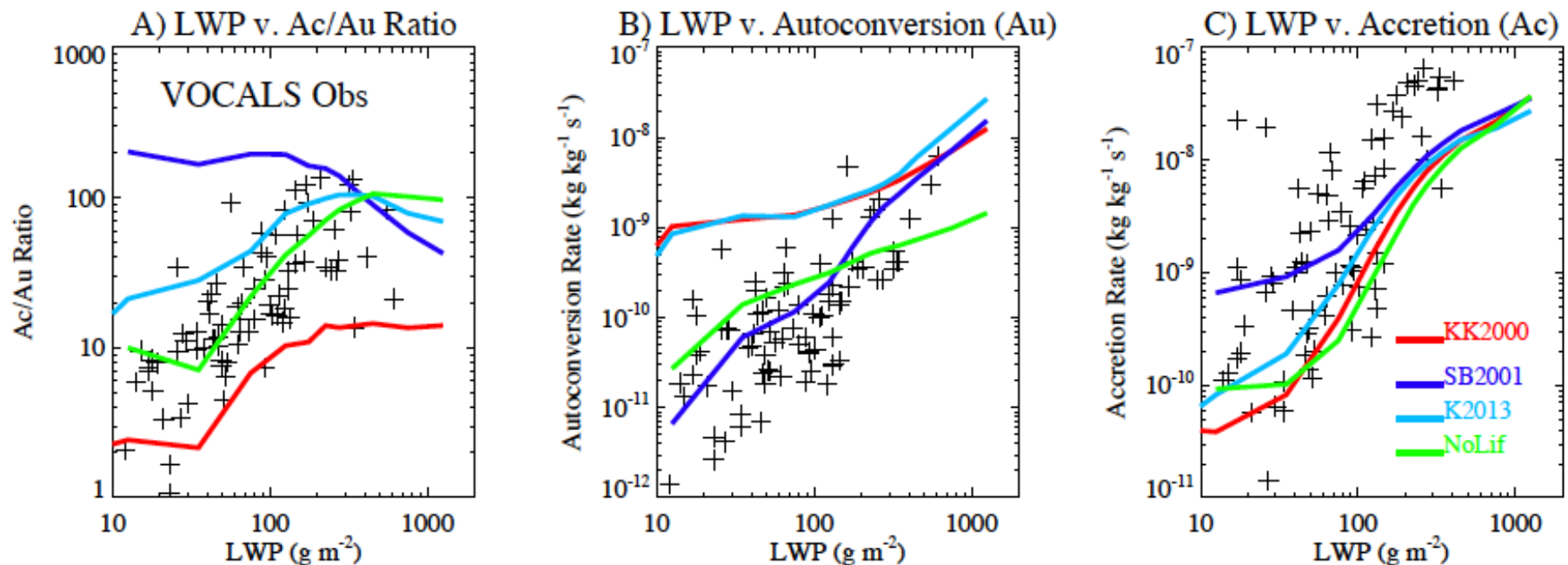
CAM5.5



NOAA-OLR



Process Rate Based Evaluation: Autoconversion



Gettelman, 2015, submitted to ACP

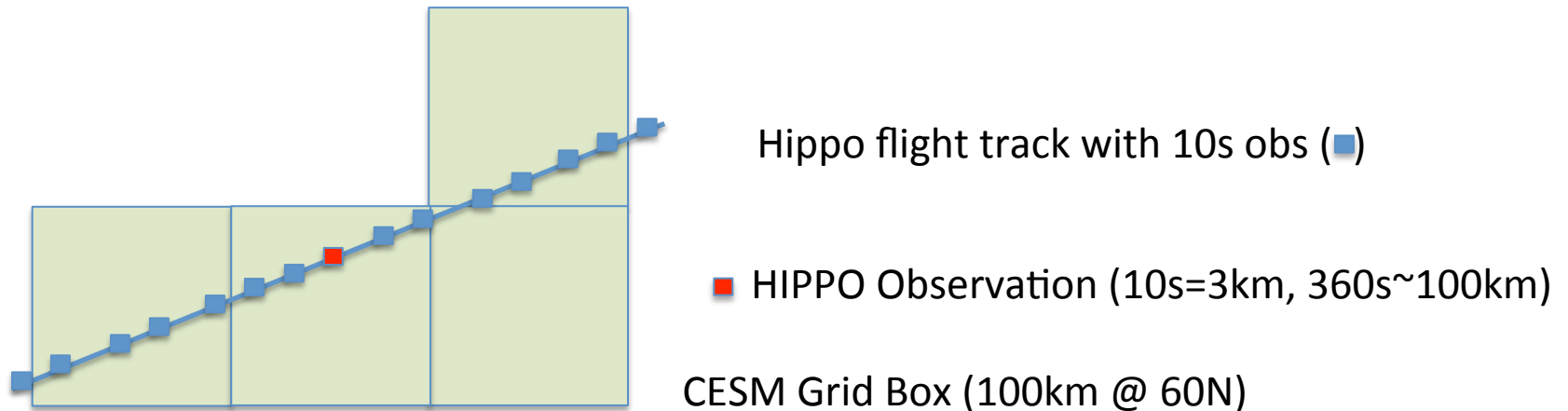
Observations = Calculations with detailed model and observed size distributions from S. E. Pacific (Terai and Wood, 2013)

Current Autoconversion, Alternative Schemes, Fixed Drop Number

Global Model Forecasts

Use a climate model like a weather model. Simulate individual cases.

- Specified Dynamics simulation: 2008-2011
- CESM1.2 (CAM5.3): GEOS-5 Meteorology, 200km resolution (equator)
 - Winds and Temps forced
 - Water species (q, clouds, aerosols) model calculated
 - Climate is reasonably in balance (-1.6 Wm^{-2} TOA)
- Output columns along (and around) HIPPO flight tracks
- Sample CESM box containing point & adjacent grid boxes
- Do every 10s. Model timestep is 1800s (oversample model)

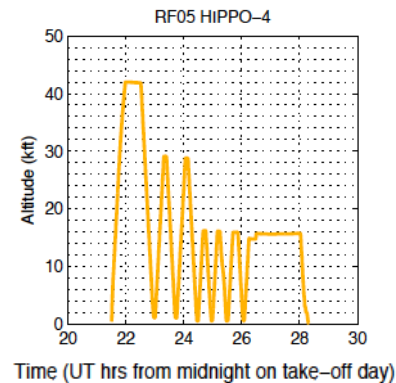
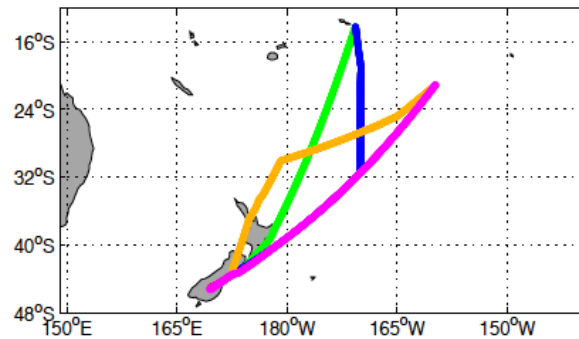


NSF G-V HIPPO Experiment

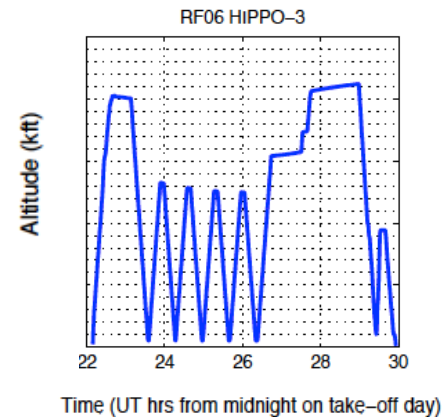
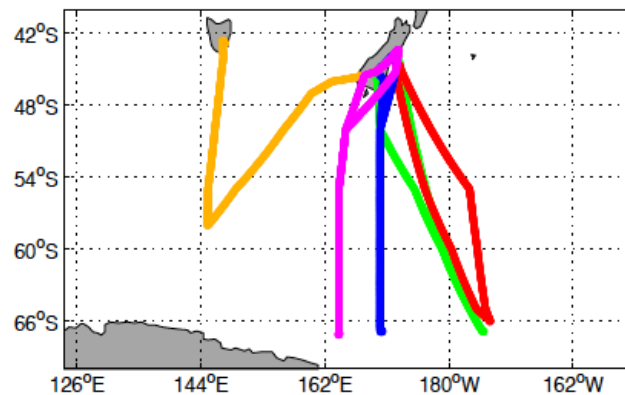
‘HIAPER Pole to Pole Observations’: multiple deployments (different seasons)

- Measured mass of liquid & ice and particle number concentrations
- Selected 2 flights with microphysics data in S. Ocean or S. Pacific

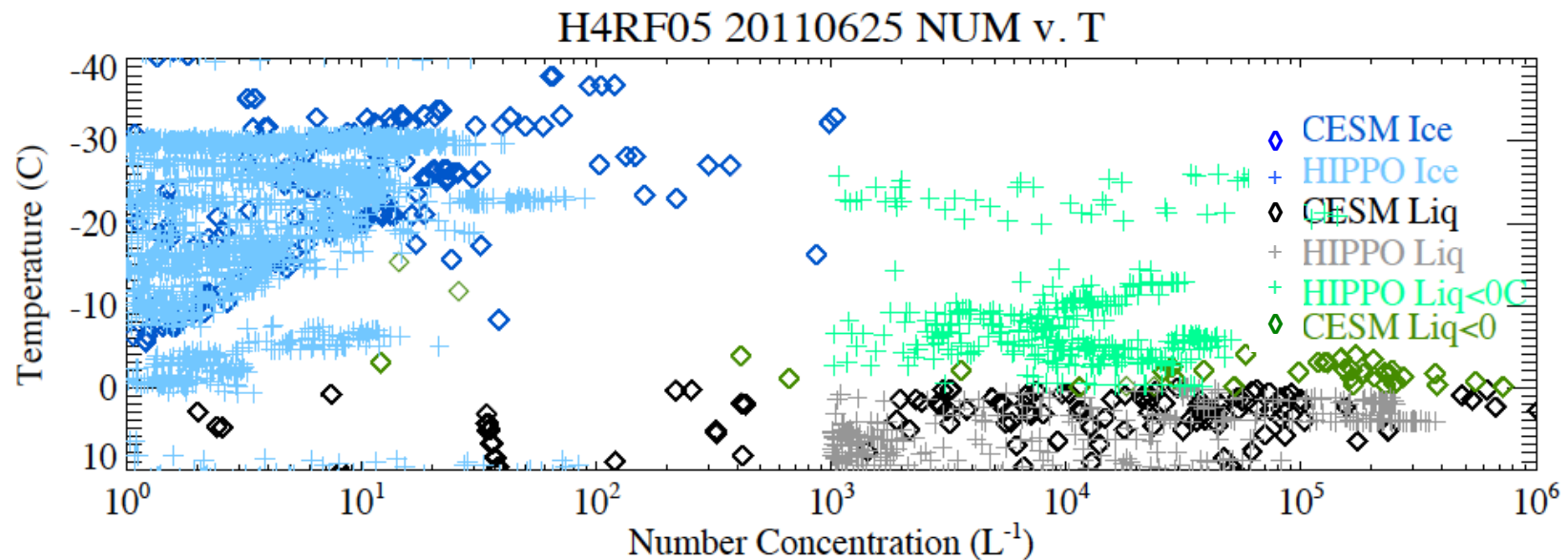
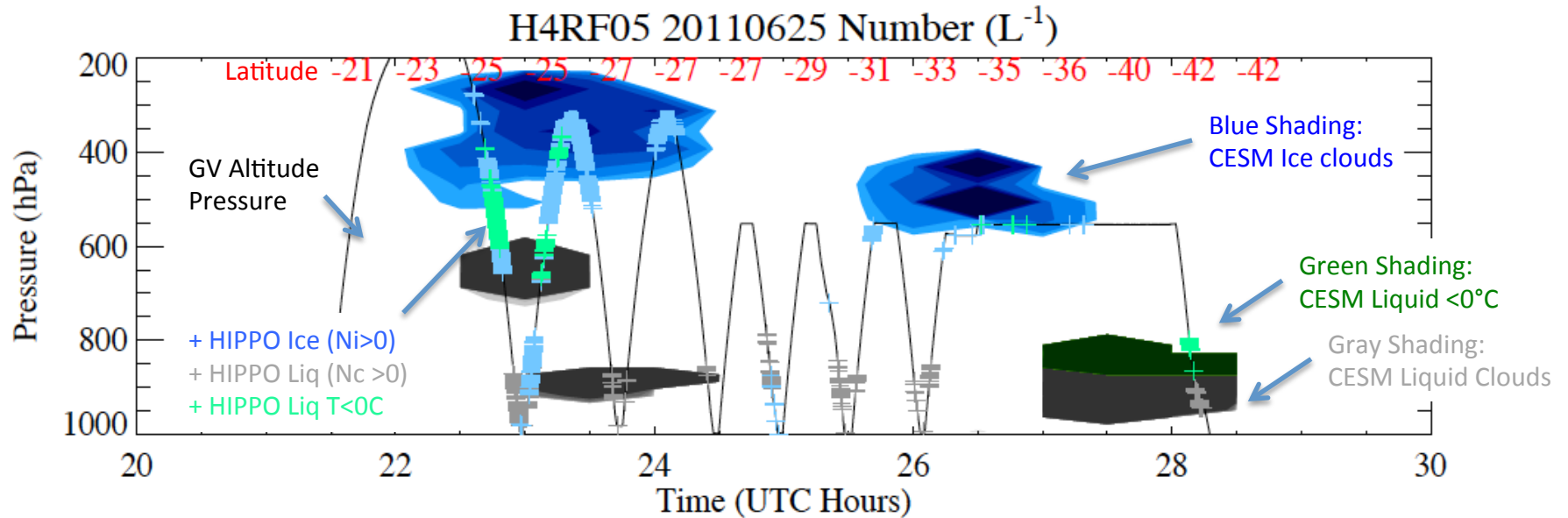
Subtropical Winter



S. Ocean in Fall

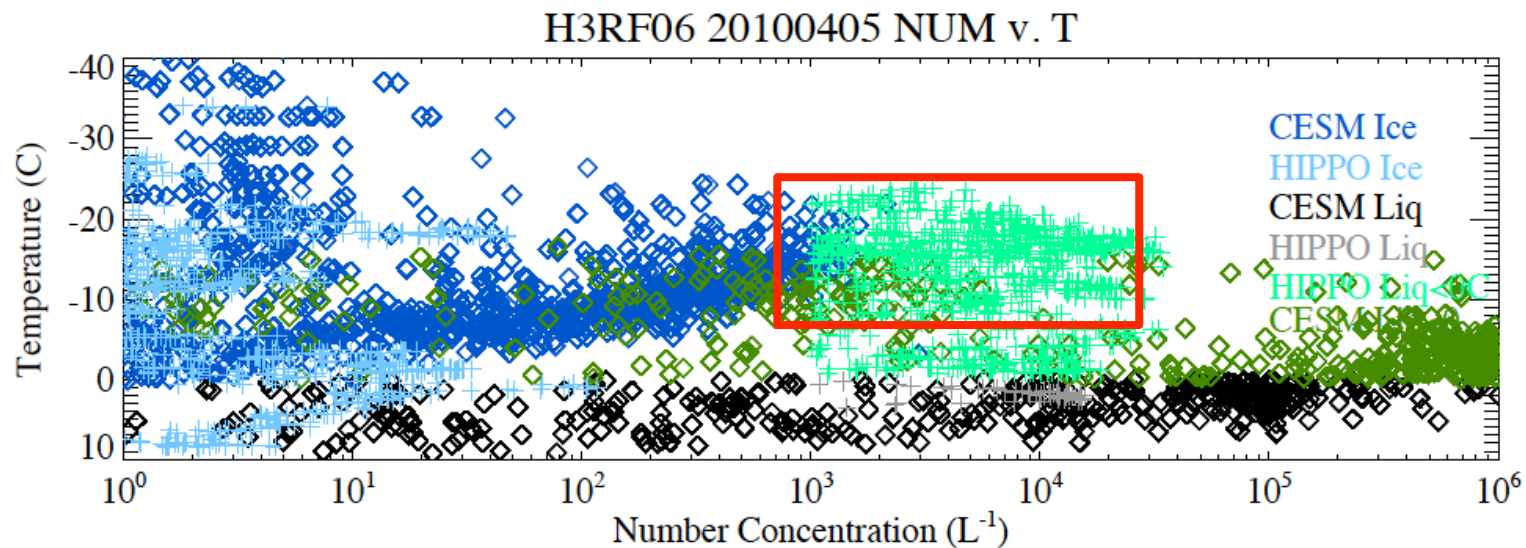
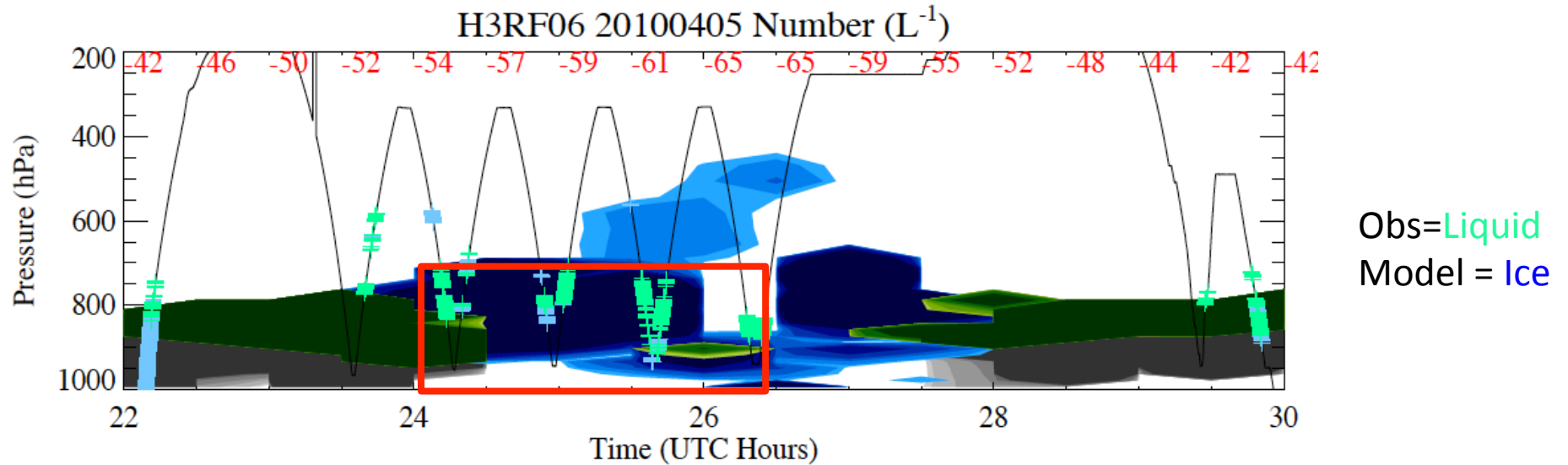


Section along H4RF05 (Jun) Flight Track

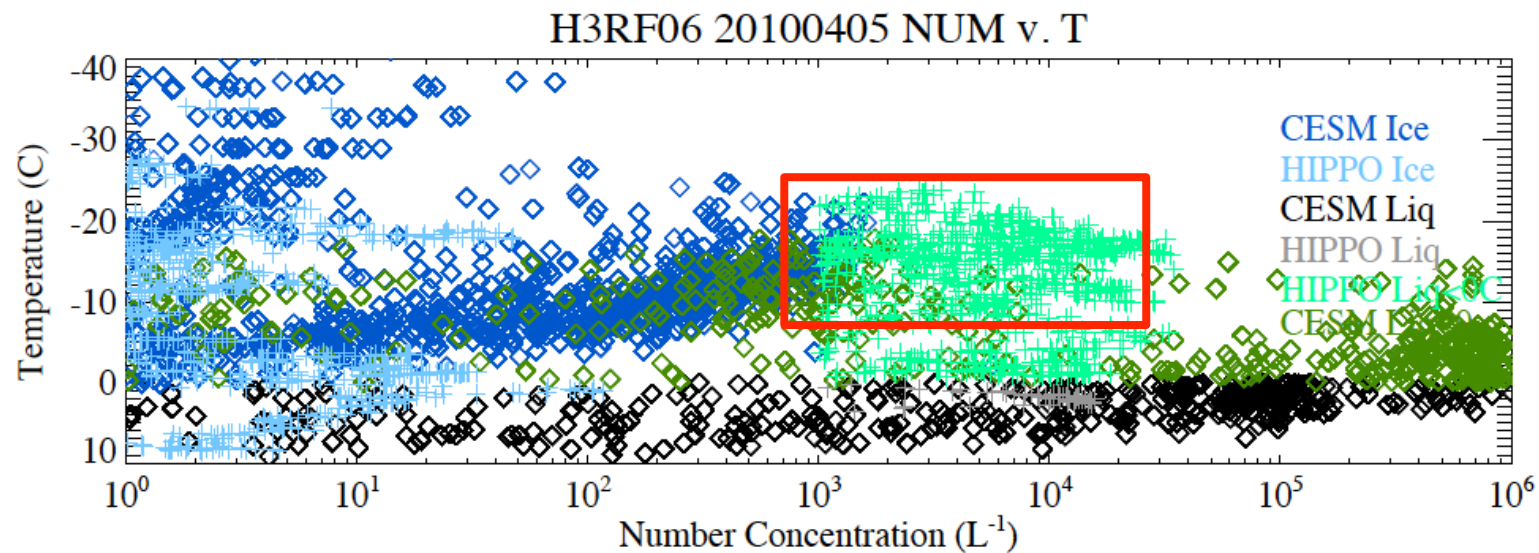
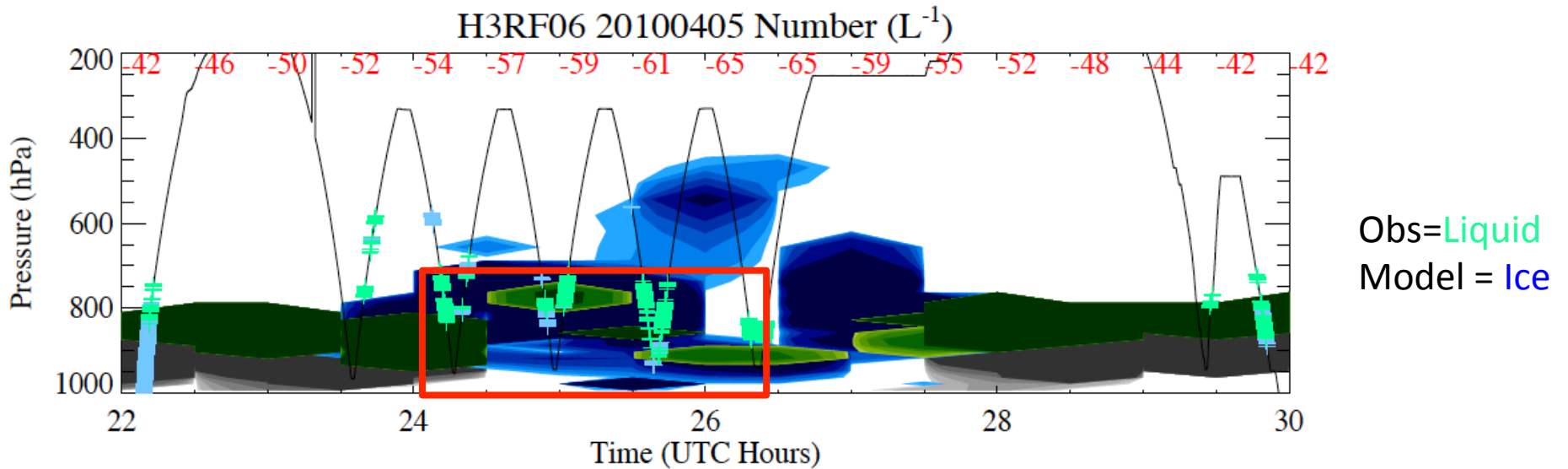


CAM5.3

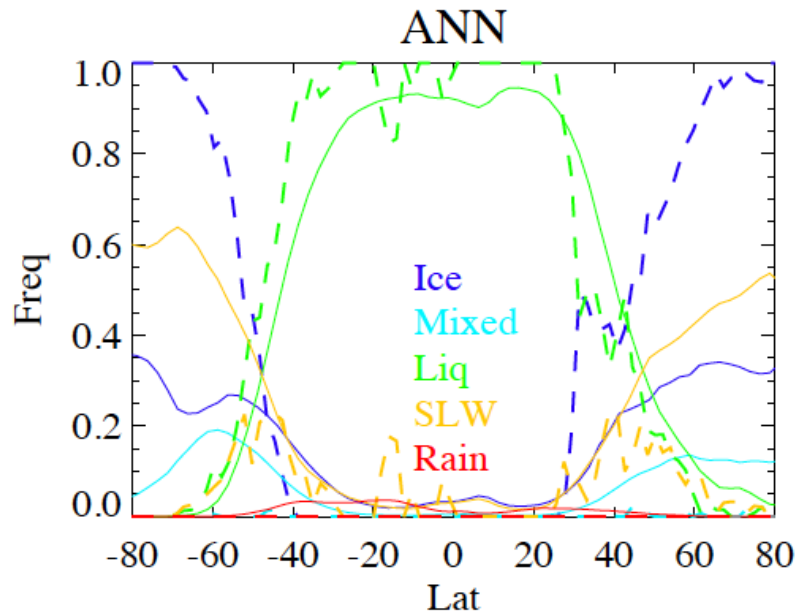
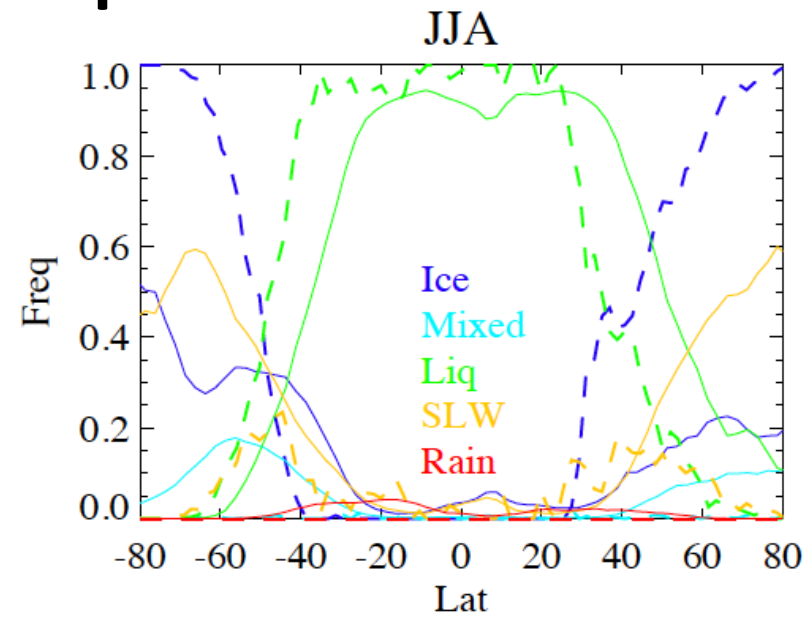
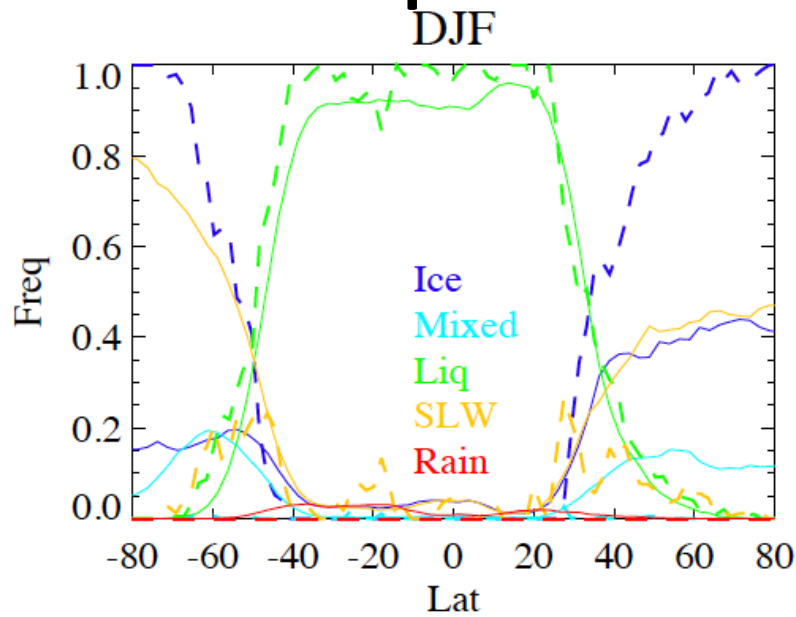
Across S. Ocean (H3RF06) April



CAM5.4 (+ New Mixed Phase Ice Nucleation [Hoose et al. 2010]) Across S. Ocean (H3RF06) April



Supercooled liquid in CESM



Frequency of occurrence of different hydrometeors at cloud top.
Solid = Satellite observations. (DARDAR)
Dashed = CAM5.4

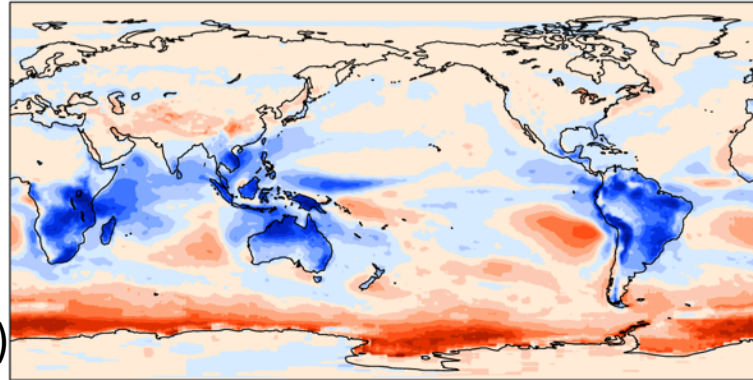
Getting some **super-cooled liquid water (SLW)**, not quite enough
Liquid looks good (too much **Ice**)

CESM2 development

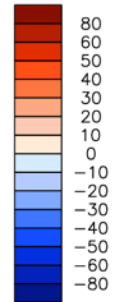
DJF SW Cloud Radiative Effect
Bias v. Satellite (CERES)
Bias = too much Absorbed Solar (ASR)
Free running (Fixed SST) simulations

Current (CESM1.2-CAM5.3)

mean = -0.91 rmse = 18.62 W/m²

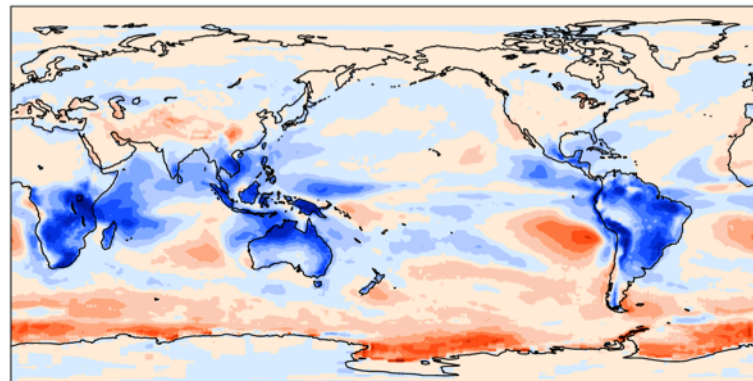


Min = -115.77 Max = 114.74

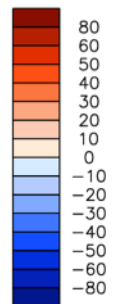


CAM5.4 (new ice nucleation)

mean = -3.02 rmse = 17.23 W/m²

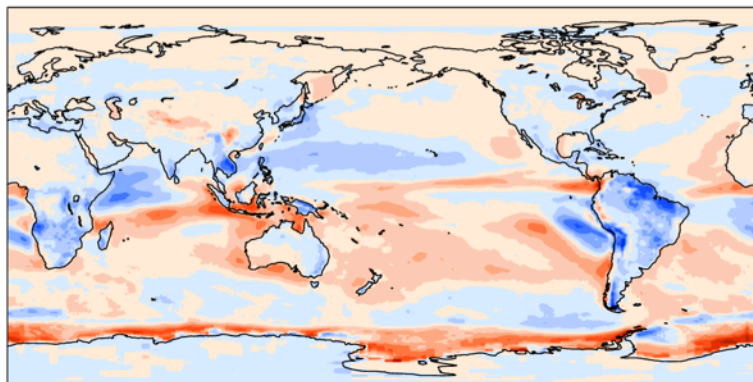


Min = -116.20 Max = 106.75

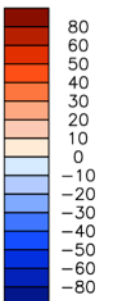


CESM2 α

mean = 2.35 rmse = 12.58 W/m²



Min = -72.89 Max = 98.74



Instrument Simulators

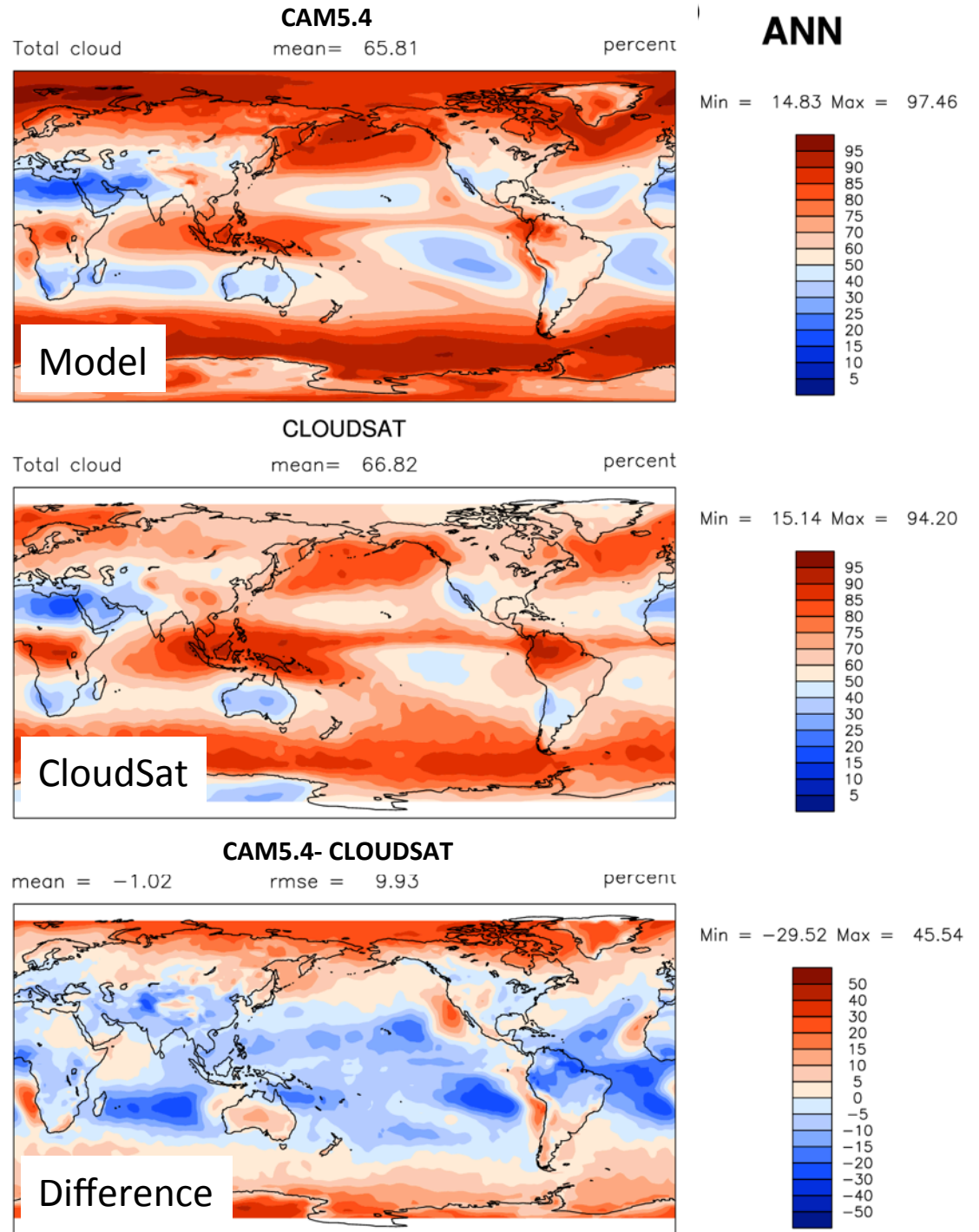
- Designed to “simulate” retrievals of a variety of satellites: includes MODIS, MISR, CloudSat, CALIPSO, ISCCP
- Why? Better comparisons between models and observations
- Mostly Cloud fraction, but also cloud microphysics from MODIS
- CFMIP Observation Simulator Package (COSP)
 - CFMIP = Cloud Feedback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

Cloud fraction...

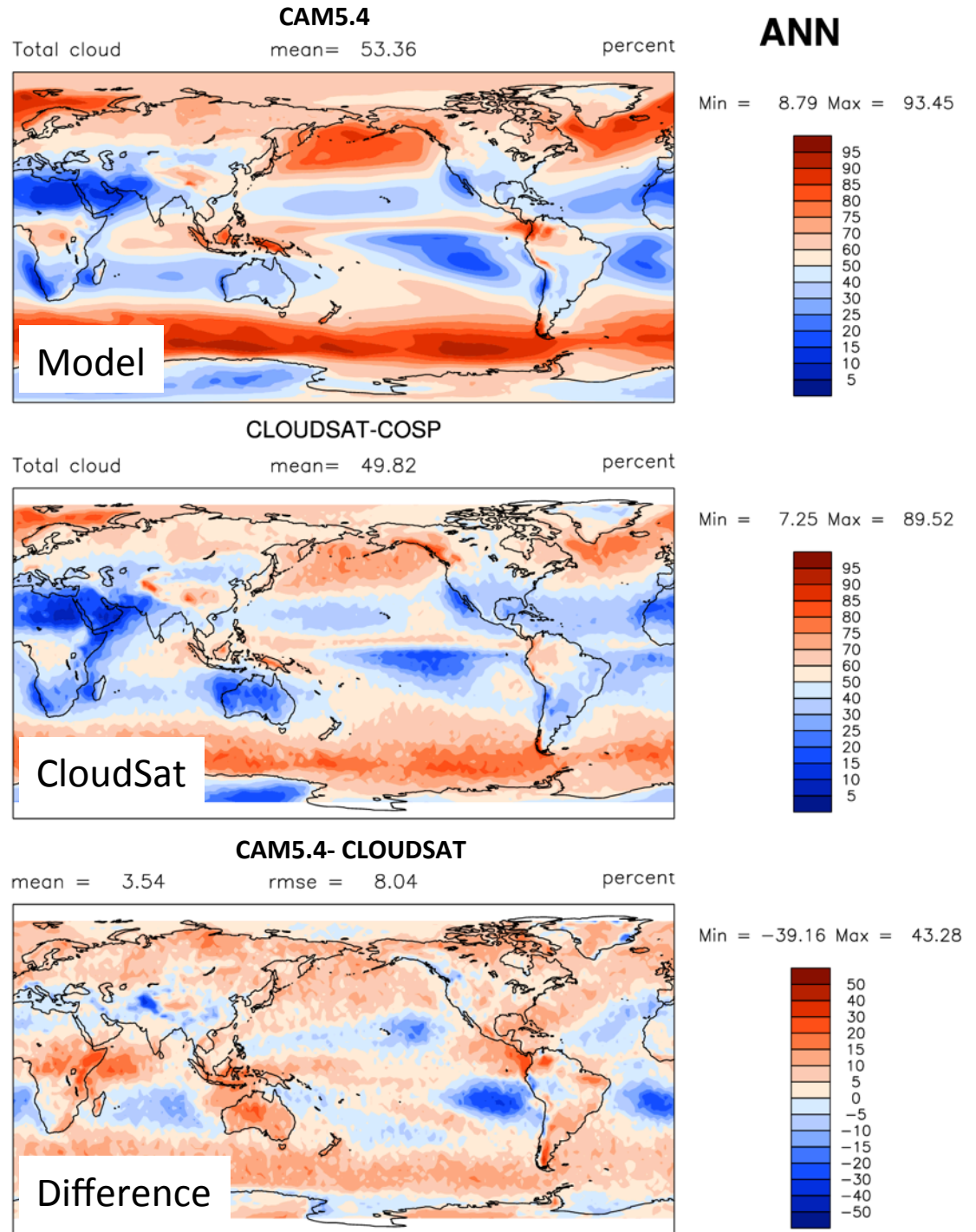
Traditional View: Model – CloudSat
(Radar+ Lidar)



Cloud fraction...

SIMULATOR VIEW Model – CloudSat

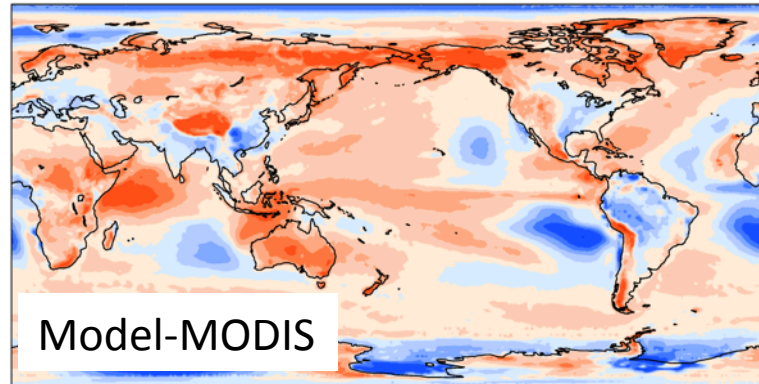
CloudSat Product only. Fewer clouds (no Lidar). Model higher in tropics.



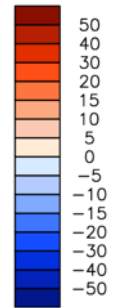
Cloud Fraction Differences ($\tau > 0.3$)

Different bias against different instruments

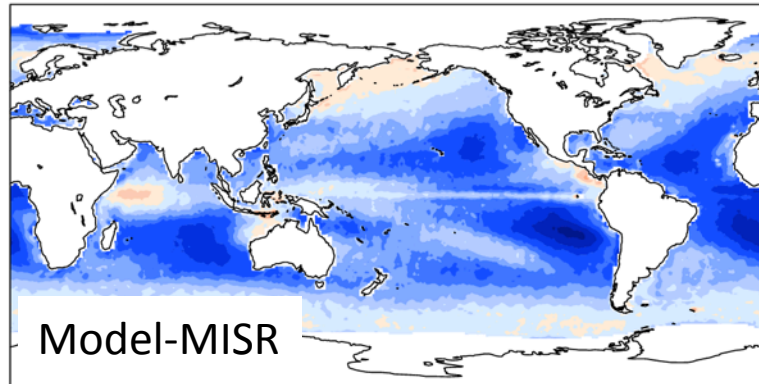
mean = 4.19 rmse = 8.97 percent



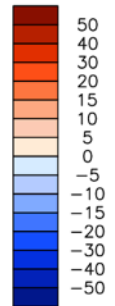
Min = -35.74 Max = 38.62



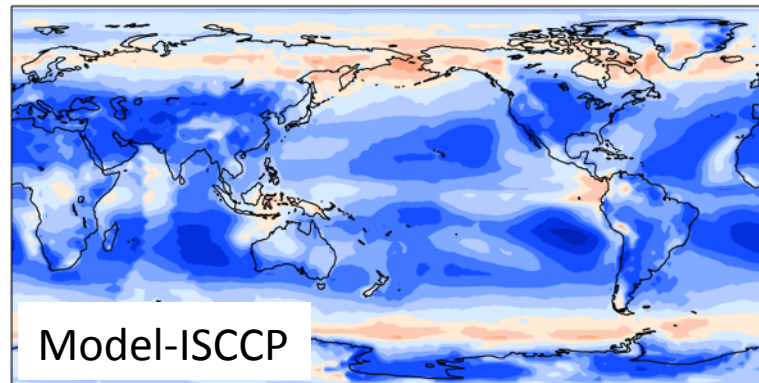
mean = -13.21 rmse = 16.70 percent



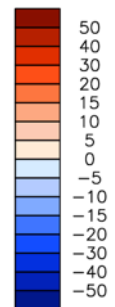
Min = -54.30 Max = 17.39



mean = -11.80 rmse = 15.15 percent



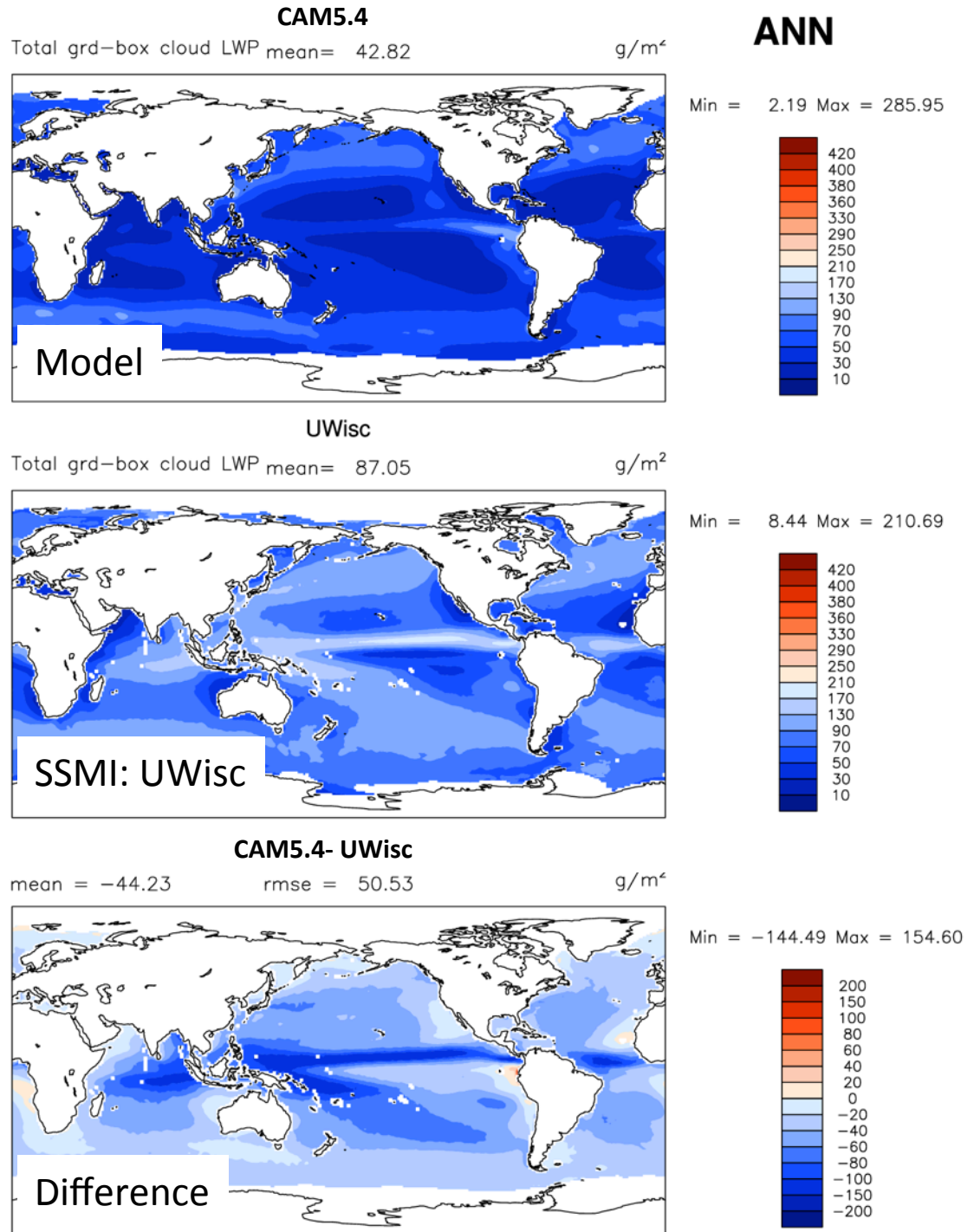
Min = -45.65 Max = 17.55



LWP: Wrong Message

Traditional comparison of Model LWP field against Microwave Satellite Observations of LWP. Model is low. But cloud forcing looks okay, and the cloud fraction looks okay. What is going on?

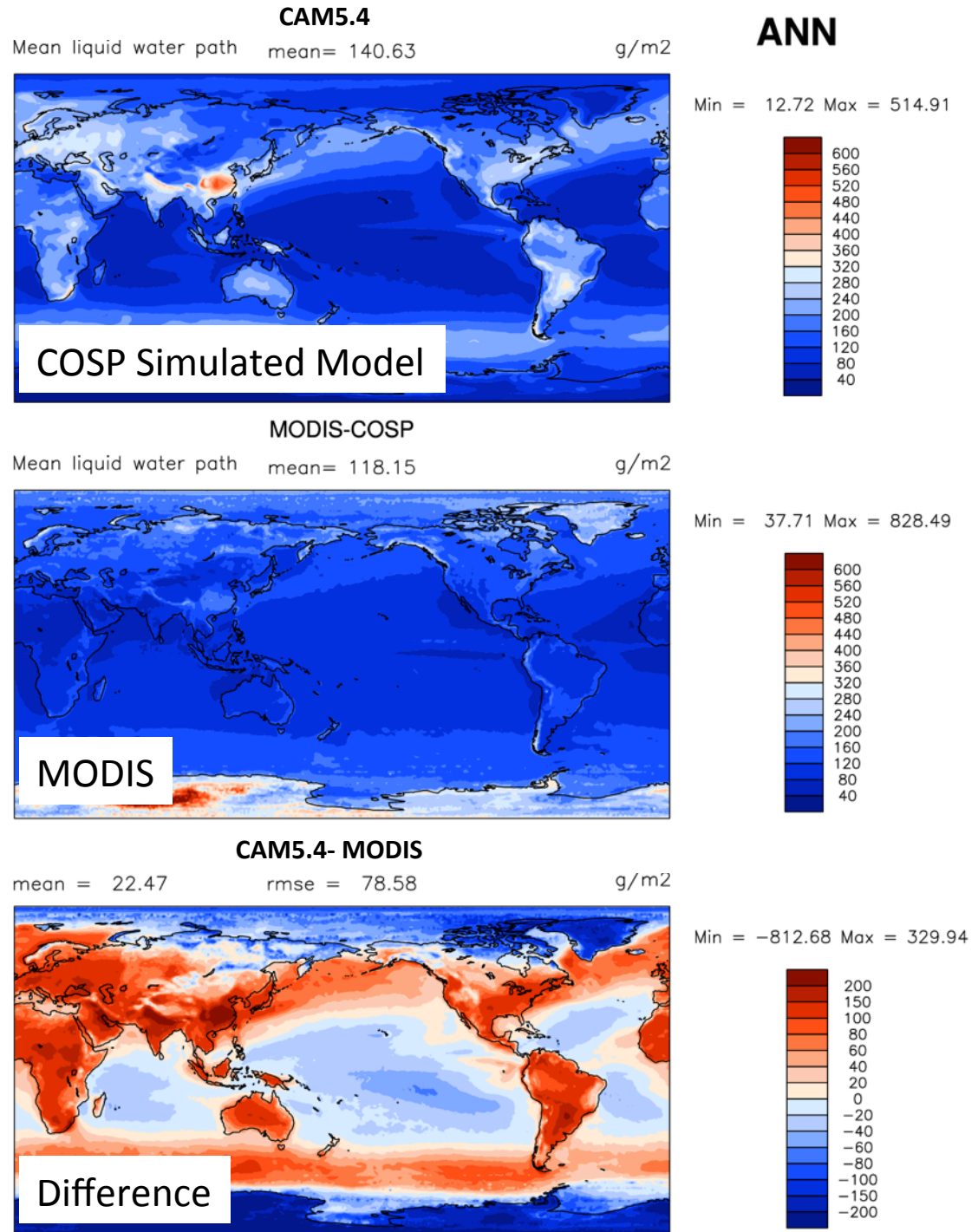
Same problem with comparison with MODIS LWP retrievals...



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

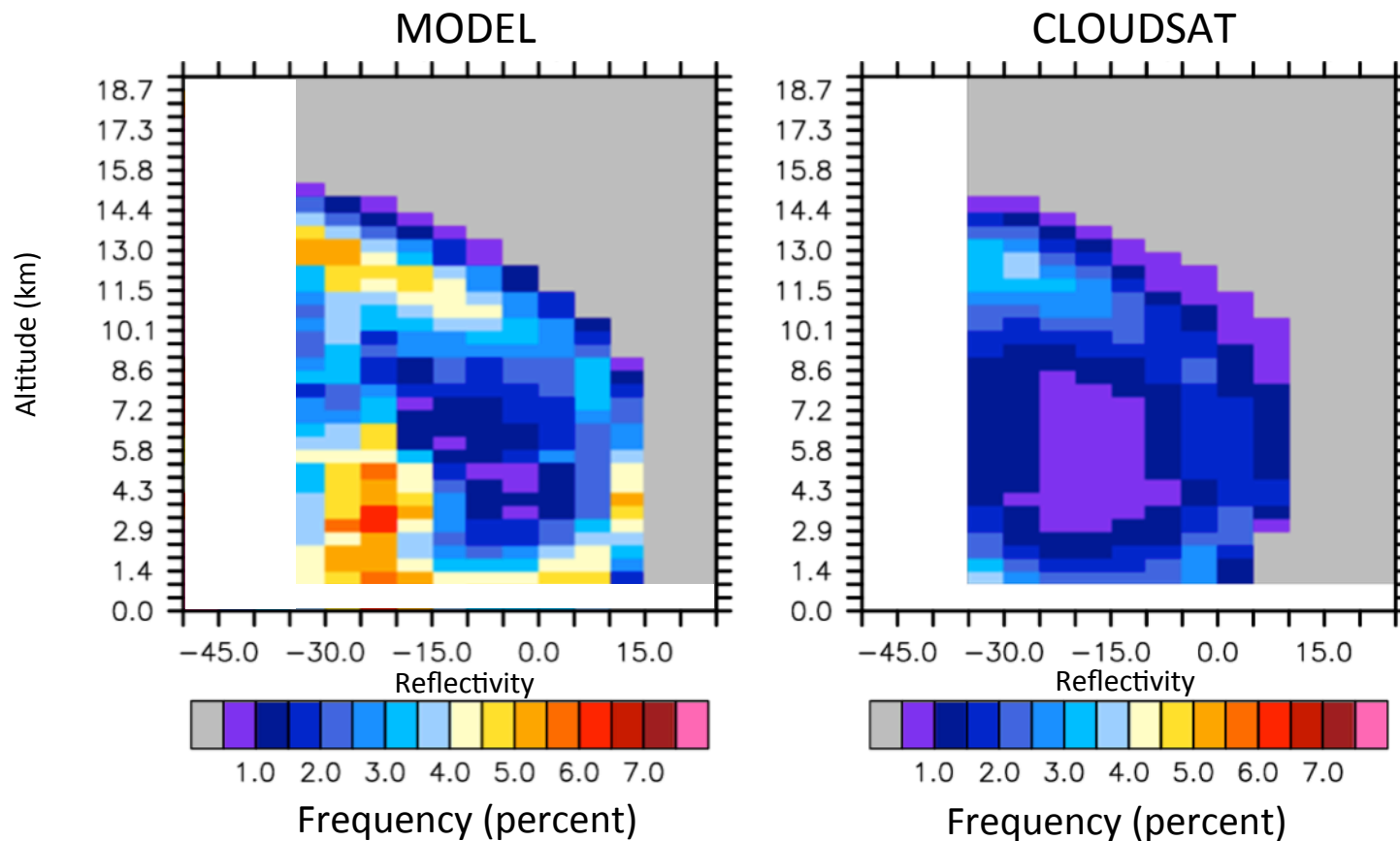
Now the model is slightly HIGHER than observations (+20%) rather than -50% LOW.



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

Summary

Best practices for cloud observations:

1. Use forecast mode for verification
2. Simulate the observation with the model (simulators).
Find the critical processes: focus on these processes (field studies). Here: mixed phase, auto-conversion

Challenges

- Scales: Cloud scales hard to observe, harder to simulate
- No easy way to get simulations to work across scales
 - We are trying: Some dogs are better than others for this
 - Fewer dogs may be better (less plumbing)
- Improving processes sometimes breaks key metrics: the challenge of rippling compensating errors
 - Some dogs work/play better with others

Thanks!

This is how René Magritte sees a cloud...

The Empire of Lights, R. Magritte, Peggy
Guggenheim Museum, Venice

