

Examining Chaotic Convection with Superparameterization Ensembles

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CoRP Science Symposium

Fort Collins, CO



Forecasting with GCMs

Large Scale



Advection Forcing

Small Scale

Parameterized

Convection
Microphysics
Radiation
Turbulence

Heating & Drying

Deterministic



Expected Values
- something like -
Ensemble Means

Non-Deterministic



Individual Realizations
with
Sensitive Dependence on Initial Conditions

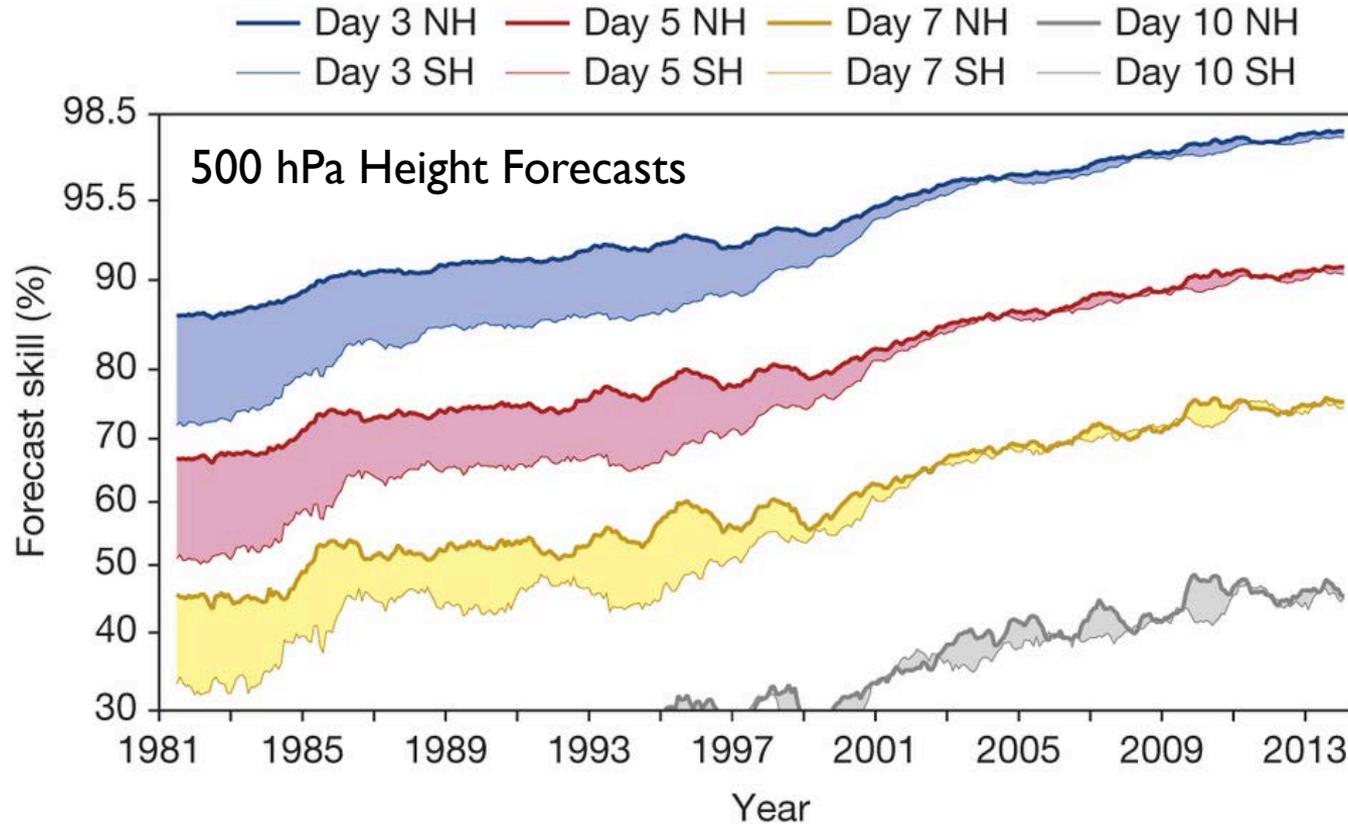
Limits of Convective Parameterizations

- Issues with SDIC, sample size, false scale separation, memory and time scales limit deterministic parameterizability.
- Real physical tendencies have a significant chaotic component.
 - They are only partially predictable.
 - Grid cell averaged precipitation is **intrinsically uncertain**.
- We confront this by providing the solution wandering space:
 - Ensembles
 - Stochastic parameterization
 - Superparameterization
 - Non-deterministic due to SDIC on the convective scale

Limits of Convective Parameterizations

- Many deterministic parameterizations yield convection that acts too quickly.
 - Poor timing and extremes
- A lack of high-frequency, small-scale variability in GCM precipitation and physical tendency fields may limit their ability to simulate low-frequency, large-scale aspects of climate variability.
 - Poor intraseasonal variability, MJO, QBO

Large-scale features are improving.

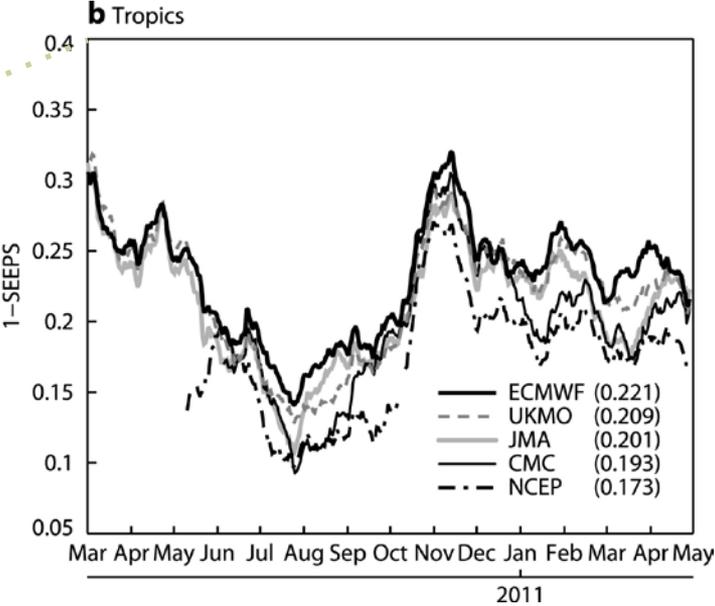
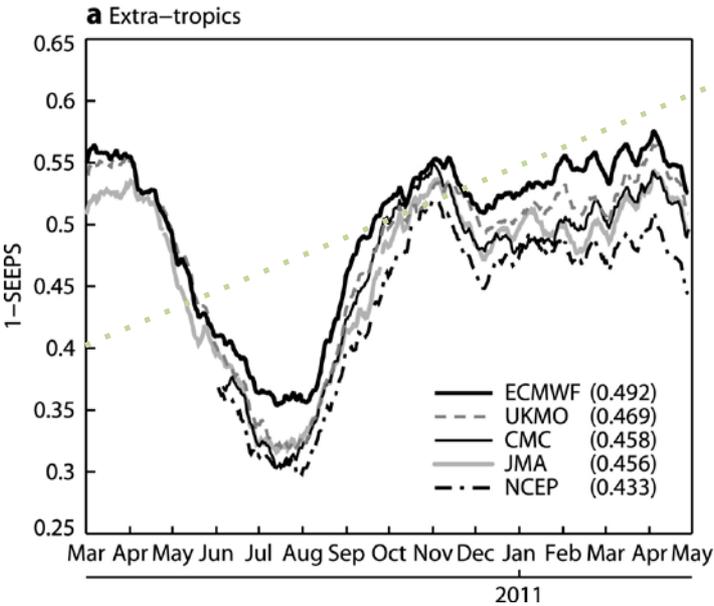


Values > 60 indicate useful forecasts.

Bauer et al., Nature, 2015

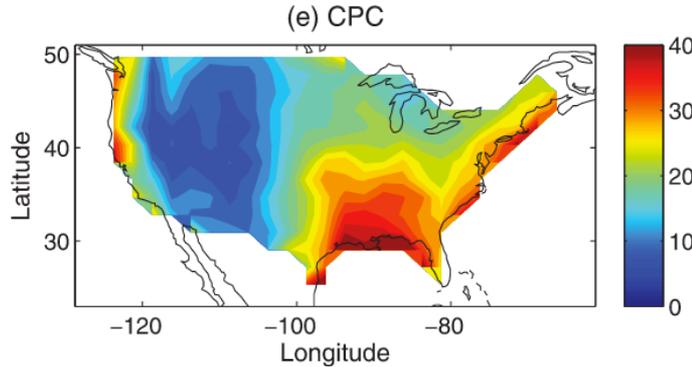
Precipitation forecasting is quite difficult...

Better Skill ↑

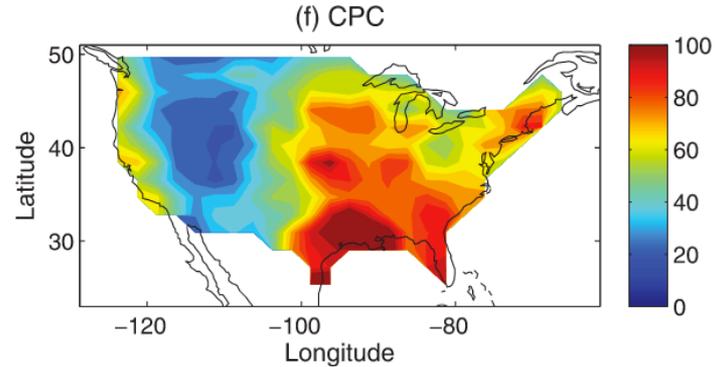


...particularly intense precipitation.

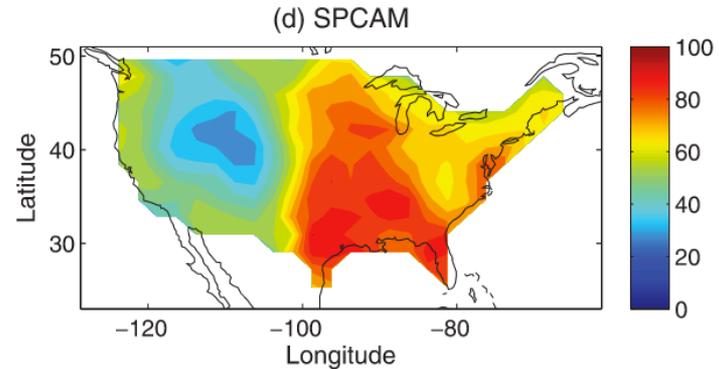
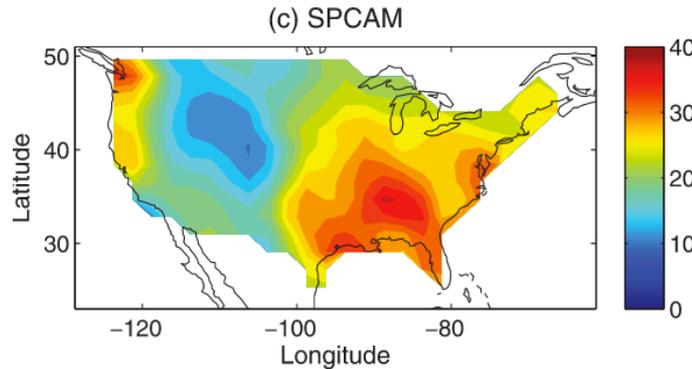
Obs 95th percentile, daily



99.5th percentile, 3-hourly



SP-CAM



We Wonder...

- Does the large-scale need to feel “jagged,” stochastic tendencies?
- Is it possible for a smoothed, more deterministic tendency to yield the same gains?
- Can we explore the range of possible realizations at the same time?

Superparameterization

CAM



Parameterized

Convection
Microphysics
Radiation
Turbulence

SP-CAM



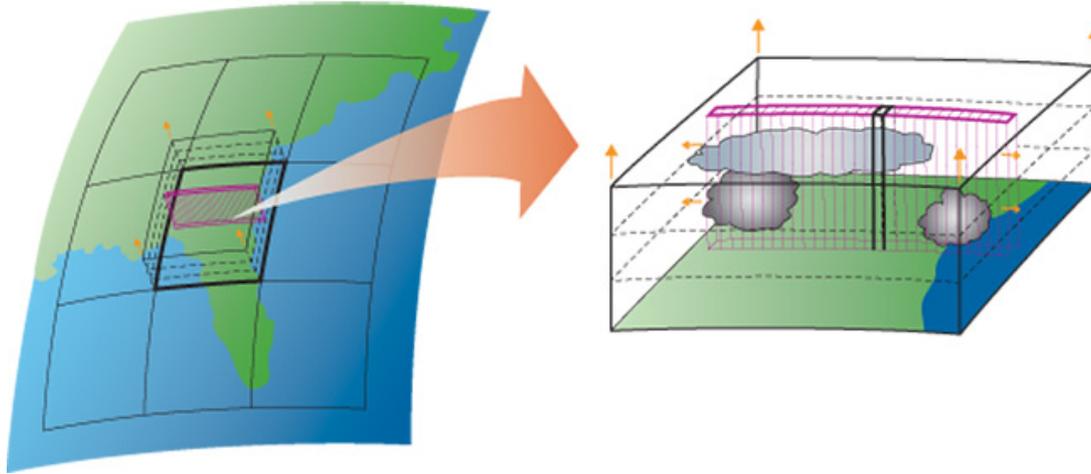
Parameterized

Microphysics
Radiation
Turbulence

2.5° × 1.9°

32 × 4-km curtain

Superparameterization

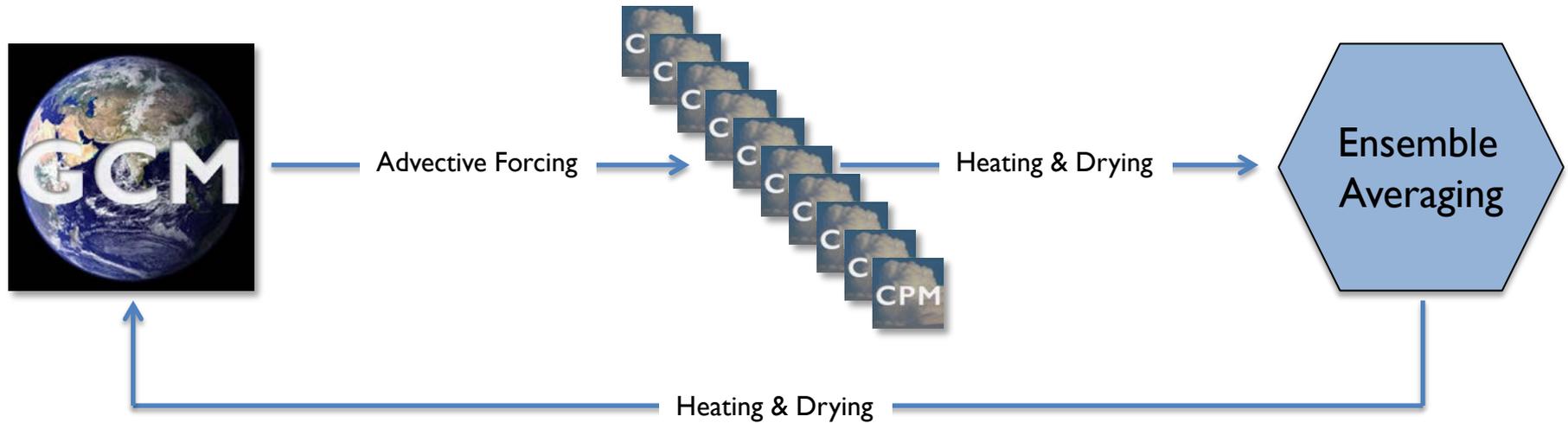


Solves equations of motion - No closure assumption - No triggers
Mesoscale organization - Convective memory - SDIC

Computationally Expensive

More Deterministic, Multiple Realizations

MP-CAM

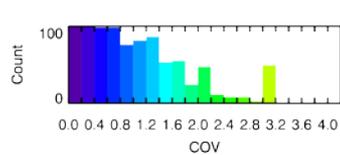
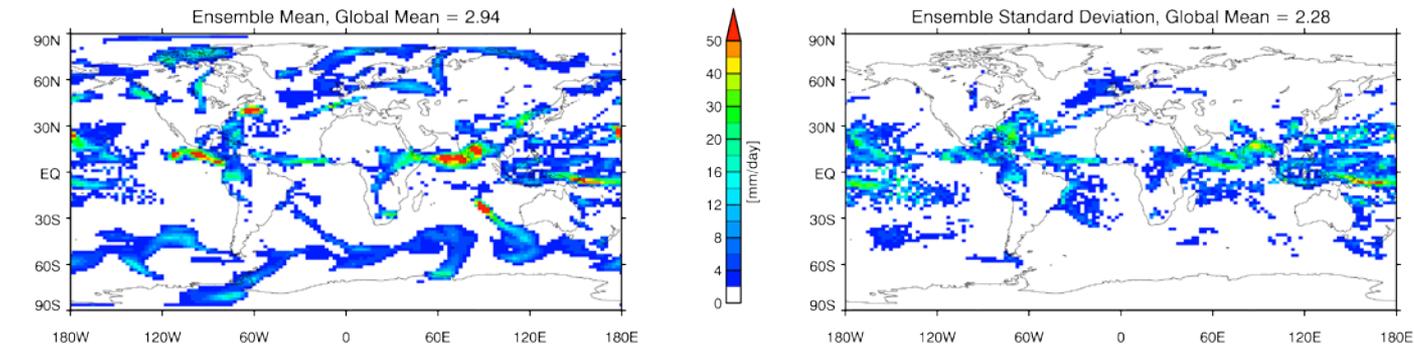
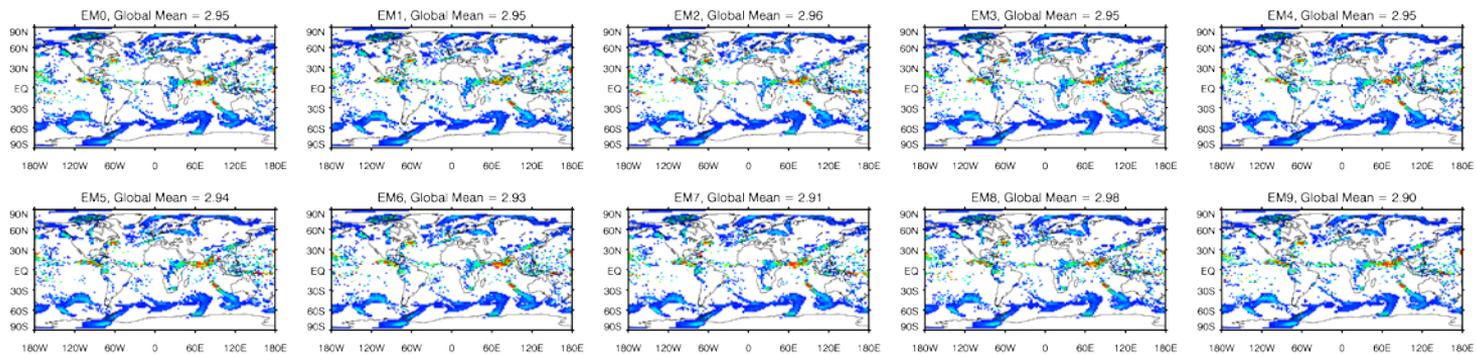


See the same GCM state - Slightly different initial conditions - Running independently

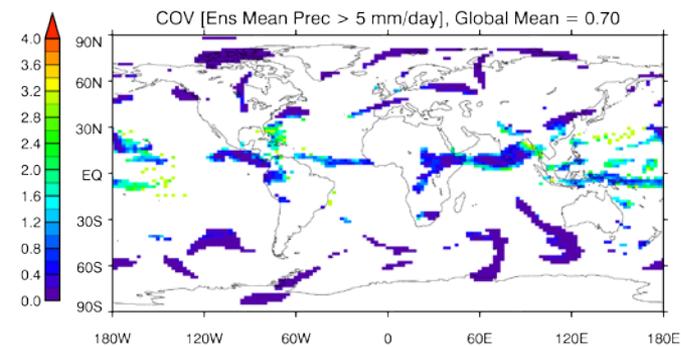
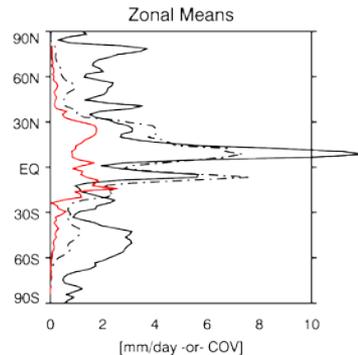
Experimental Design

- CAM-FV with $2.5^{\circ} \times 1.9^{\circ}$ longitude-latitude grid
- Climatological SSTs
- SP-CAM, 30 years
 - 32×4-km curtain, oriented N-S
- MP-CAM, ~23 years
 - 10 clone CPMs
 - 4 years of 3-hourly data
 - 2 days every GCM time step
- SP-CAM Ensemble “Forecast,” 10 days





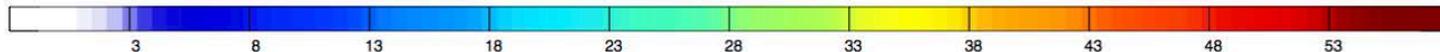
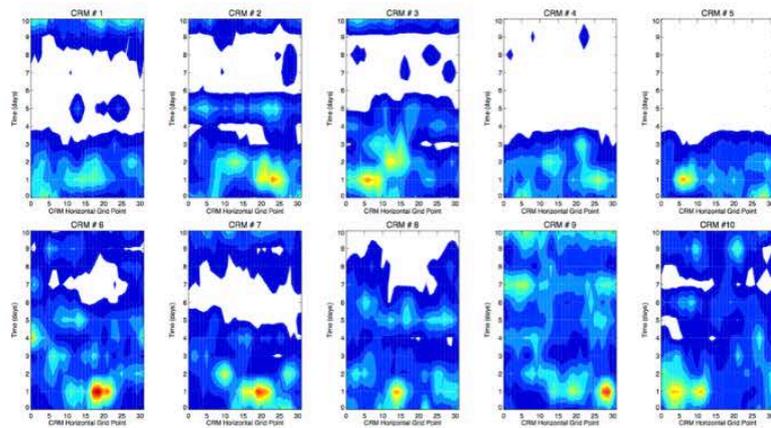
Single Day
Oct 3
Precipitation



Ten realizations at a tropical point



Time

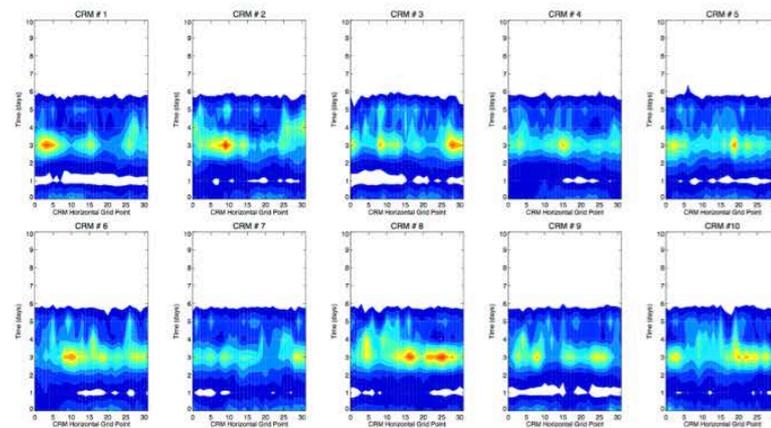


mm day⁻¹

Horizontal



Ten realizations at a midlatitude point



A more deterministic superparameterization

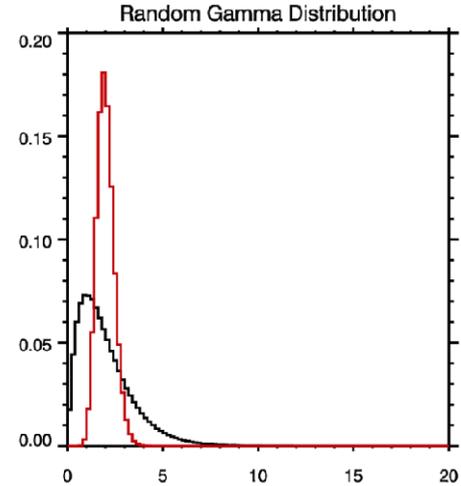
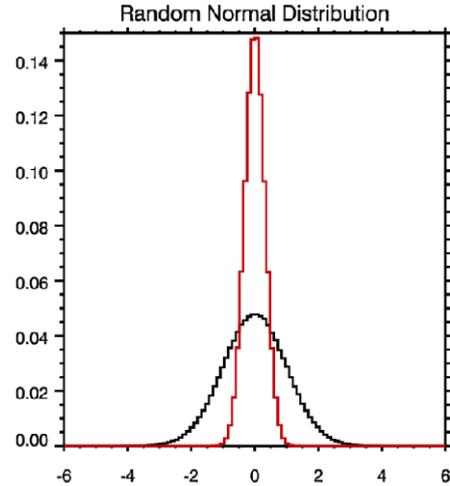
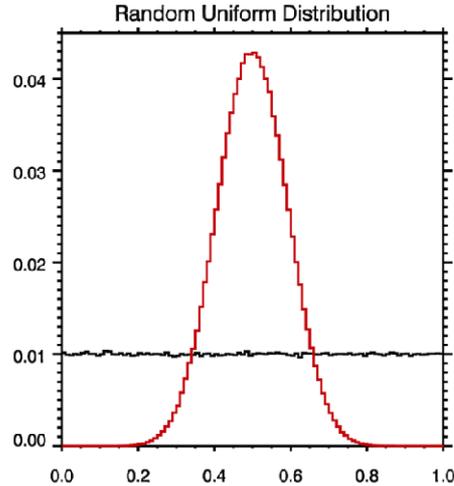
Did we really create one?

A more deterministic superparameterization

Standard deviations of averages are smaller than standard deviations of individual time series.

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

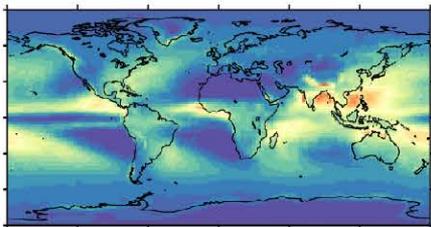
PDFs with 10 random time series



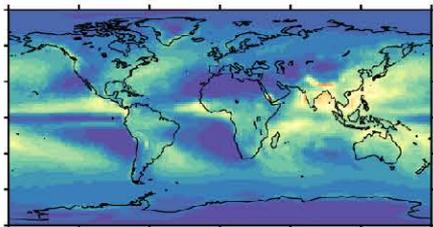
...But our data is not random...

Standard deviation of local time series

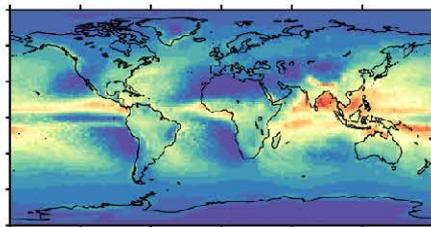
Control



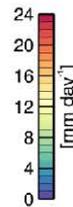
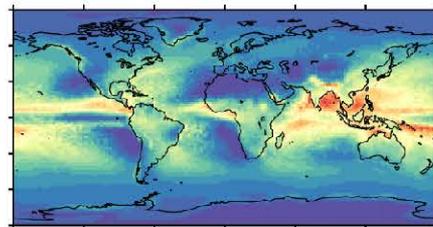
Ensemble



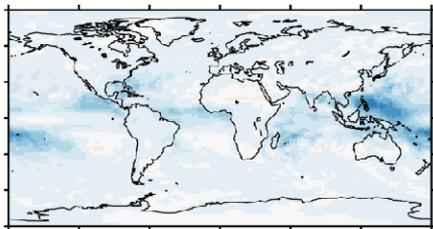
Member 1



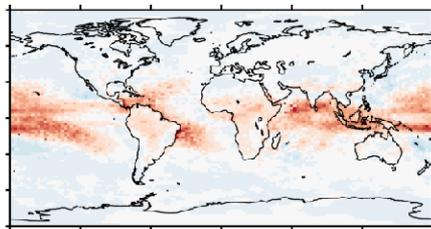
Member 2



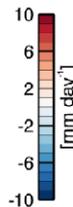
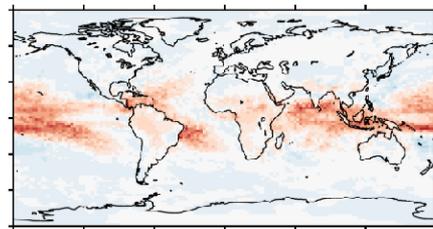
Ensemble



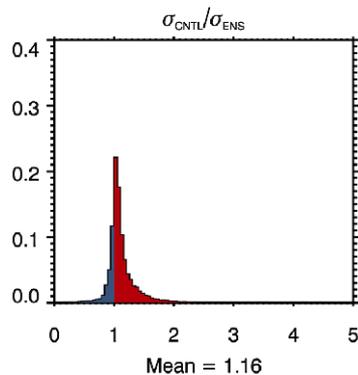
Member 1



Member 2

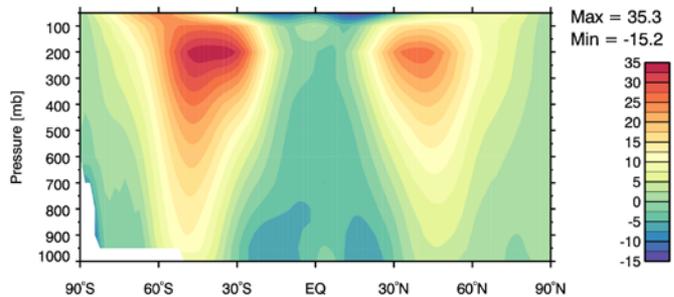


Difference from control →

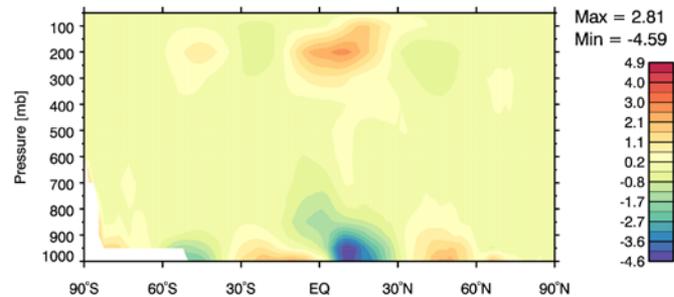


How do the climates of SP and MP differ?

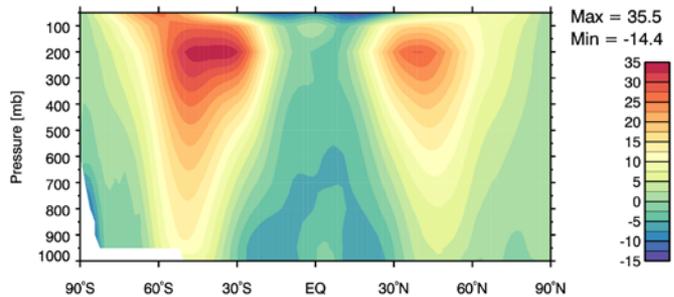
Zonal wind [m/s]
ANN
SP



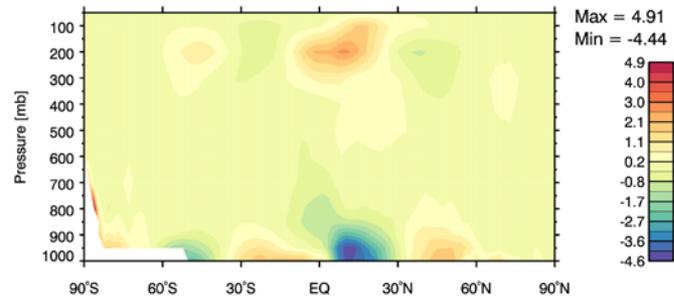
Meridional wind [m/s]
DJF
SP



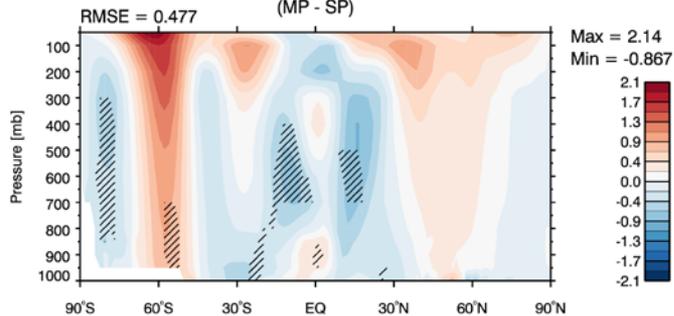
MP



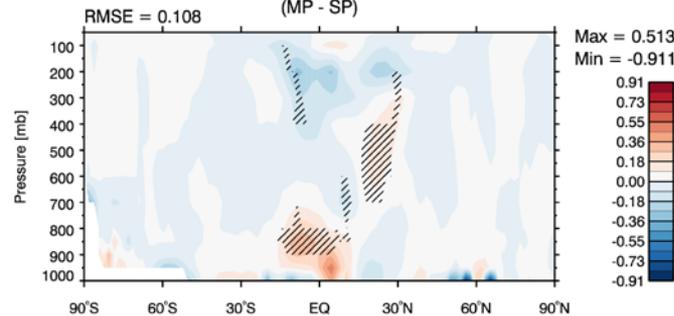
MP

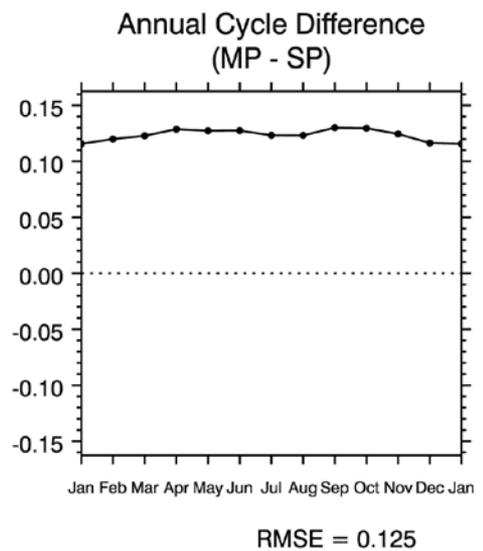
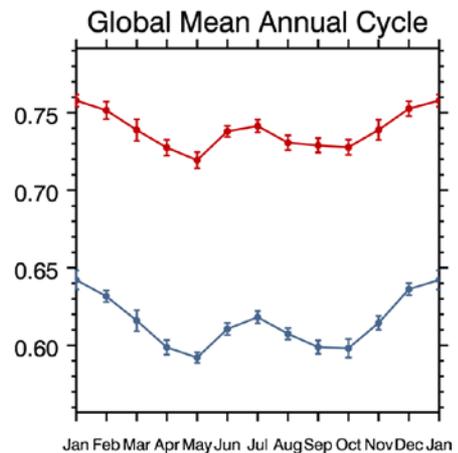
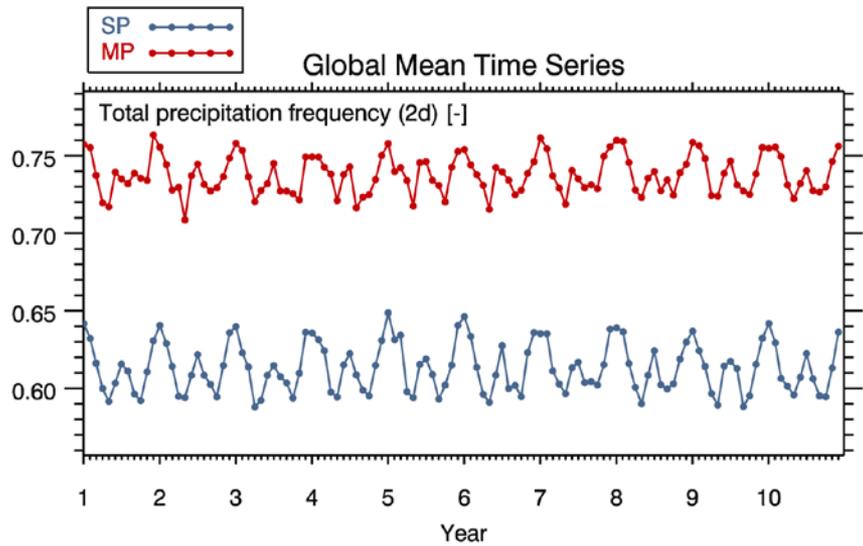


Difference
(MP - SP)

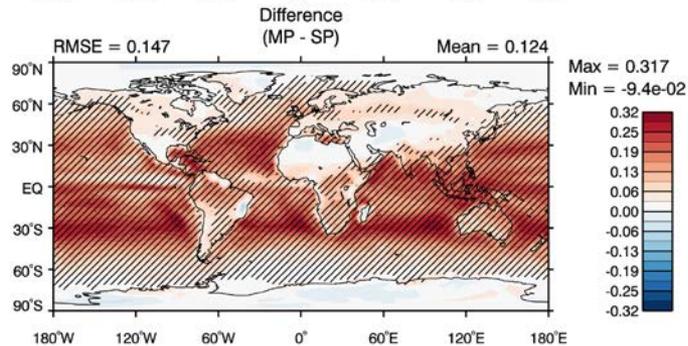
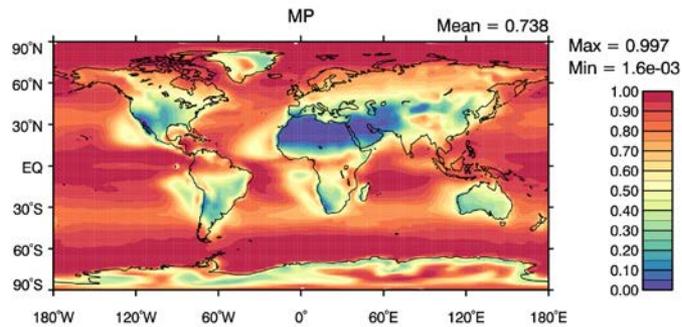
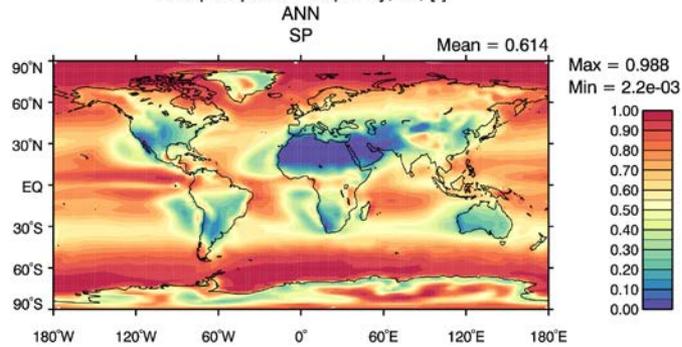


Difference
(MP - SP)

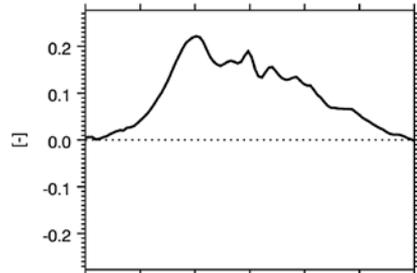
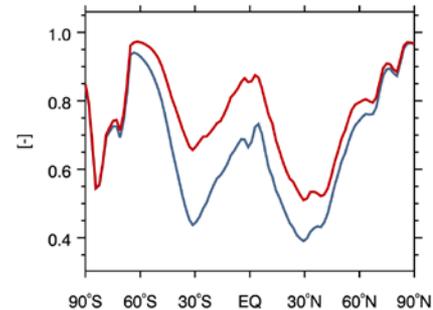


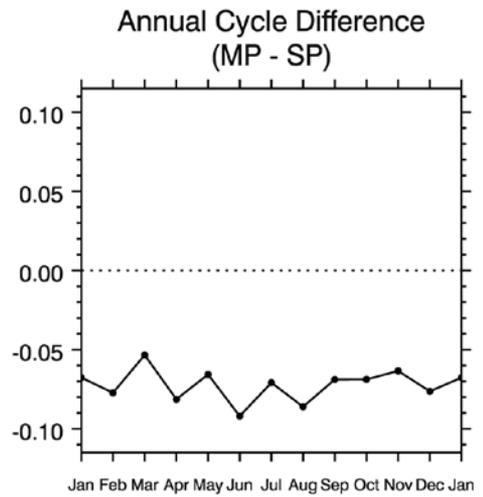
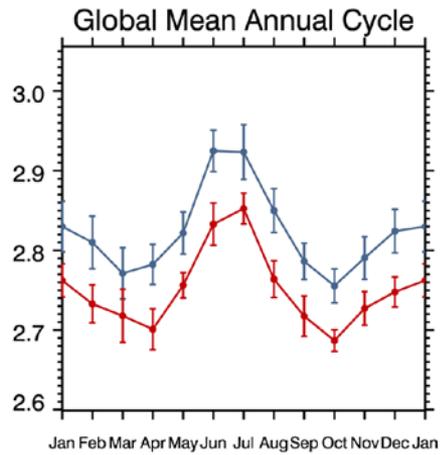
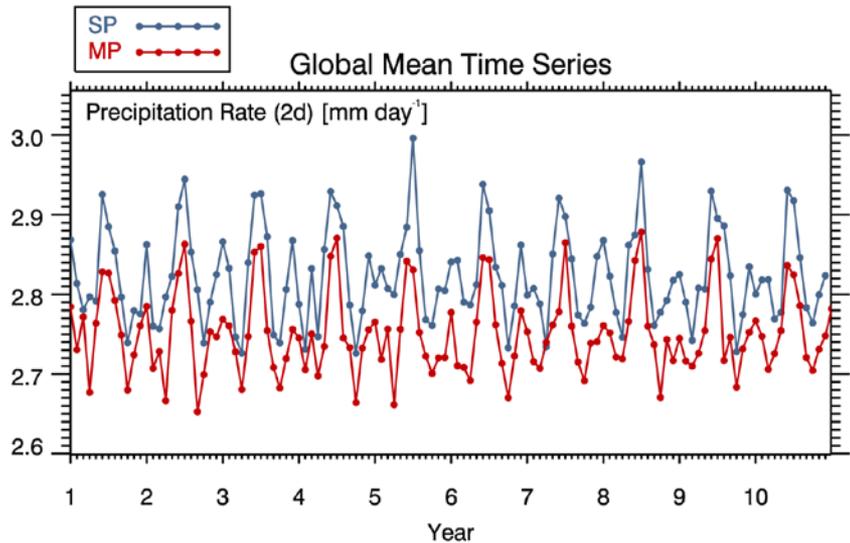


Total precipitation frequency, 2d, [-]

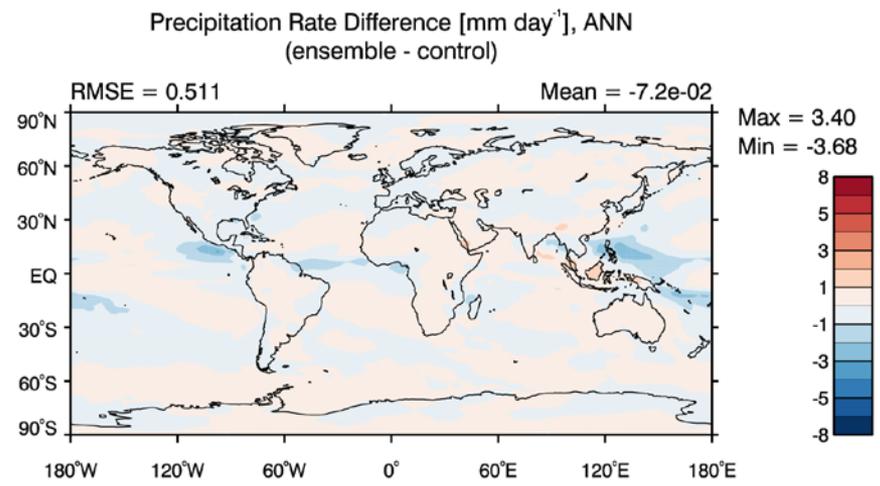


Total precipitation frequency
ANN

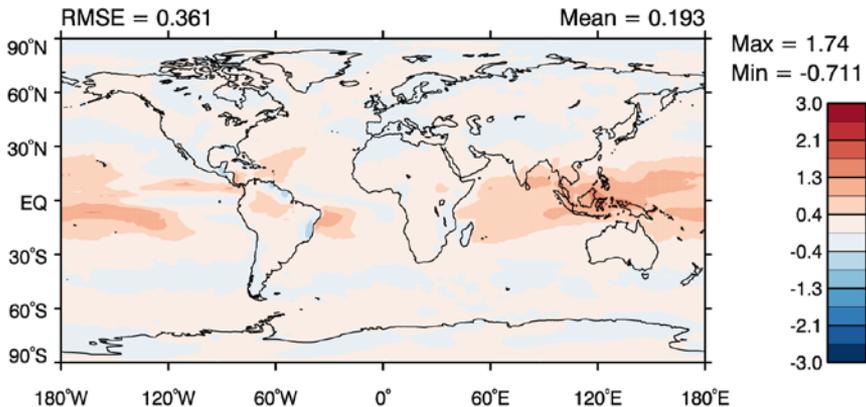




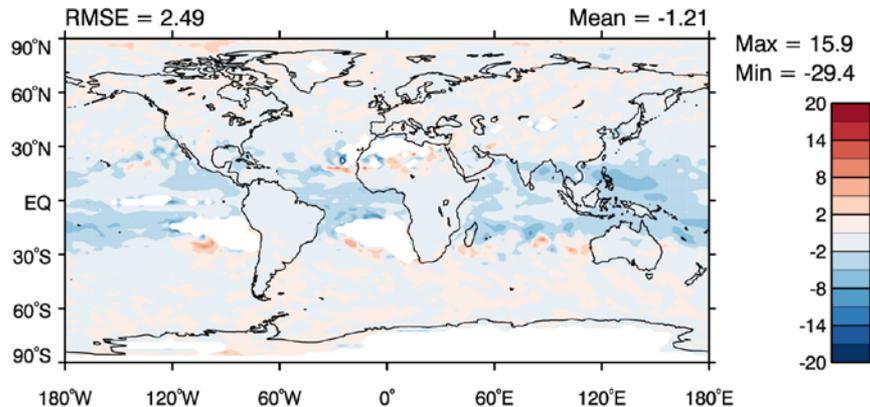
RMSE = 8.3×10^{-2}



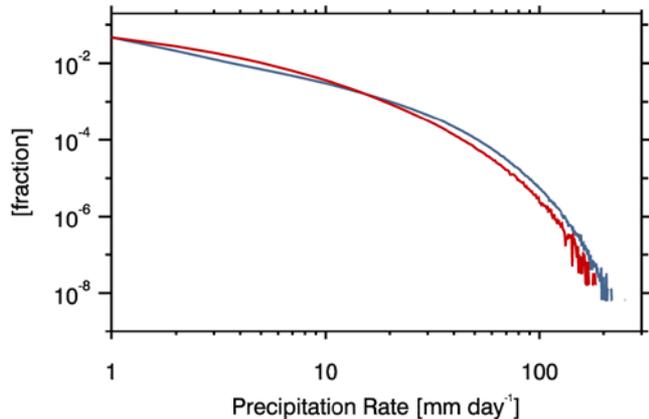
Precipitation Rate Difference [mm day^{-1}], (< 10), ANN
(ensemble - control)



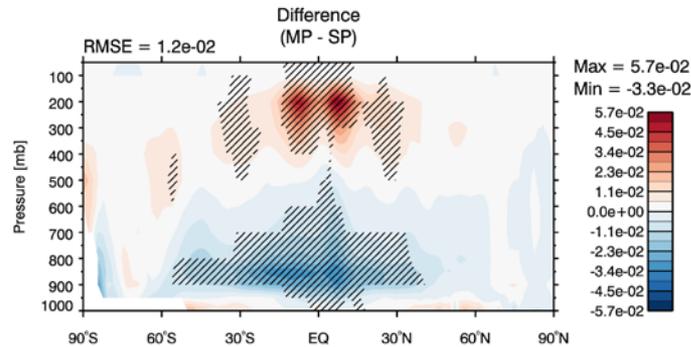
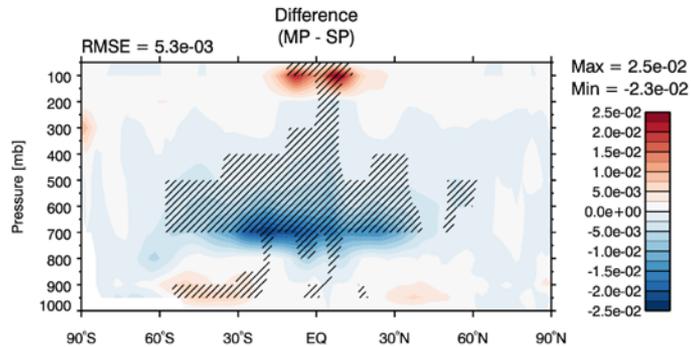
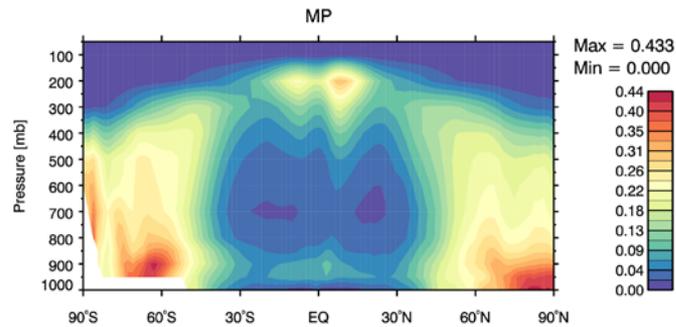
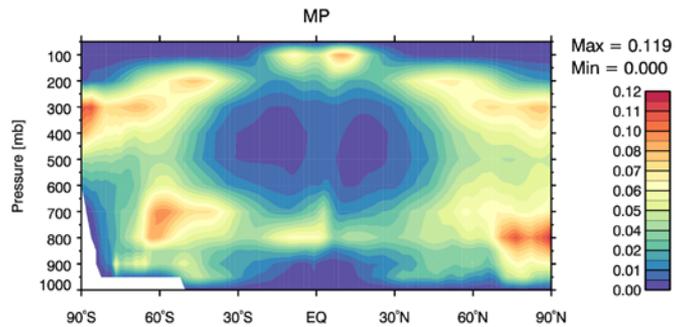
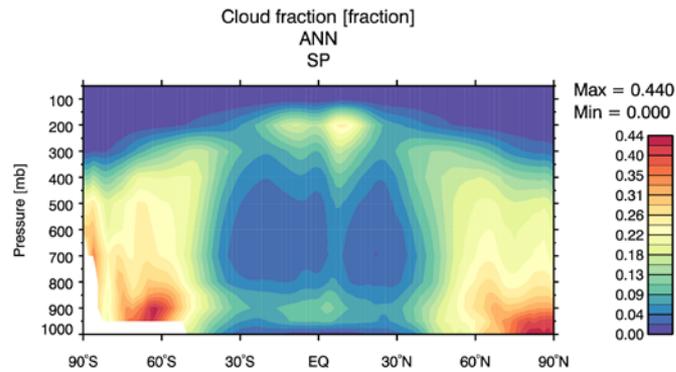
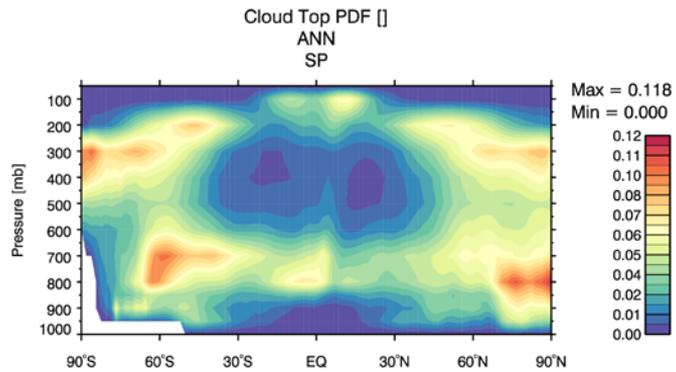
Precipitation Rate Difference [mm day^{-1}], (> 10), ANN
(ensemble - control)



Daily Mean Precipitation PDF
Tropics, ANN



- More frequent precipitation
- Increased light precipitation
- Decreased heavy precipitation

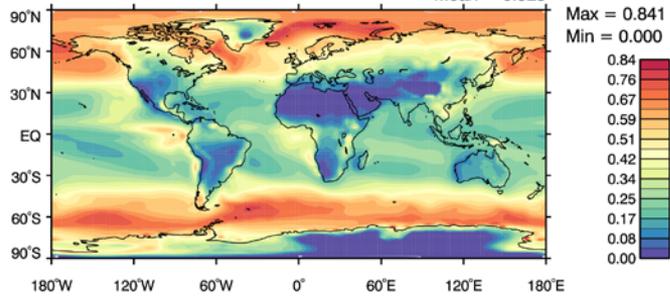


Vertically-integrated low cloud, 2d, [fraction]

ANN

SP

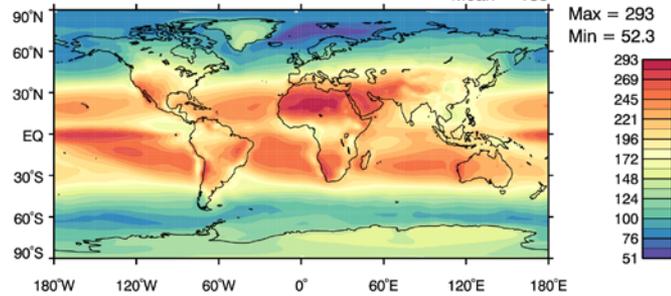
Mean = 0.323

Downwelling solar flux at surface, 2d, [W/m²]

ANN

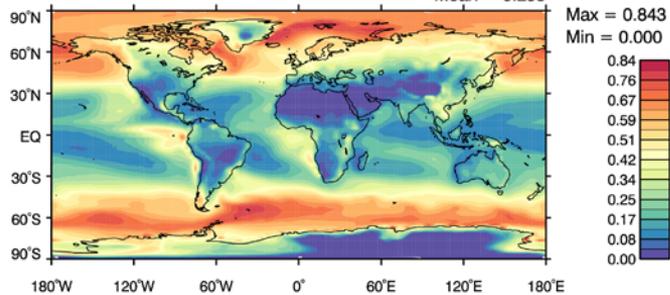
SP

Mean = 183



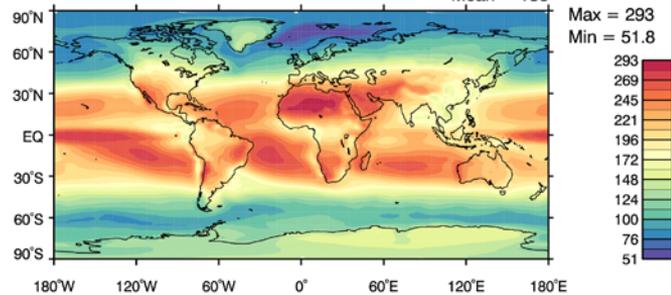
MP

Mean = 0.299



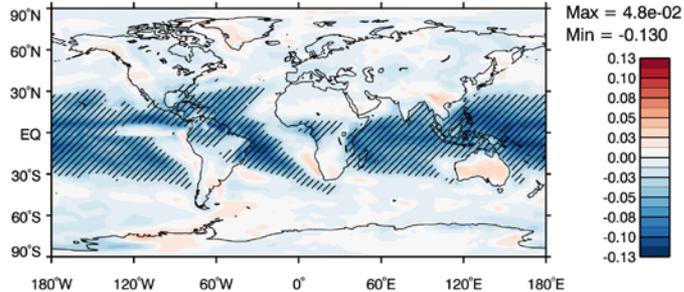
MP

Mean = 186

Difference
(MP - SP)

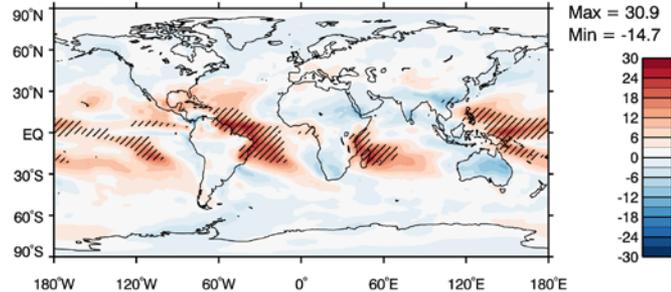
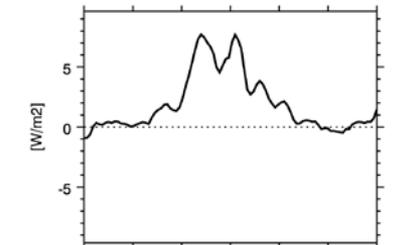
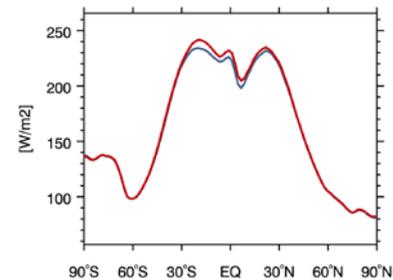
Mean = -2.4e-02

RMSE = 4.0e-02

Difference
(MP - SP)

Mean = 2.81

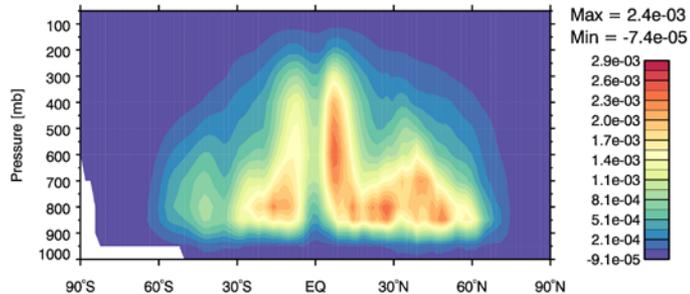
RMSE = 6.34

Downwelling solar flux at surface
ANN

Total mass flux from CRM [kg/m²/s]

ANN

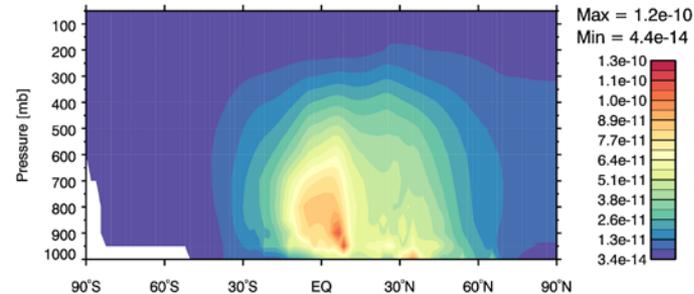
SP



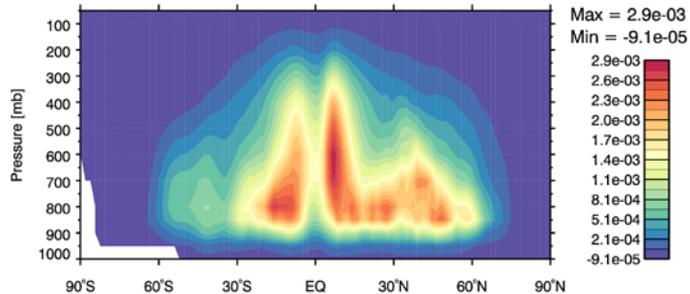
CB2 concentration [kg/kg]

ANN

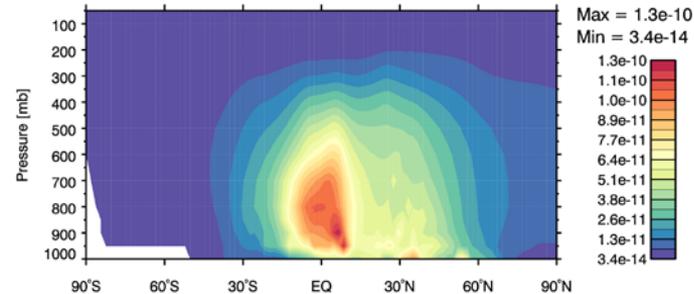
SP



MP

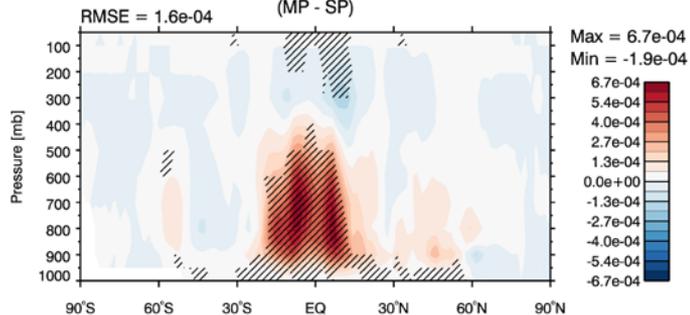


MP



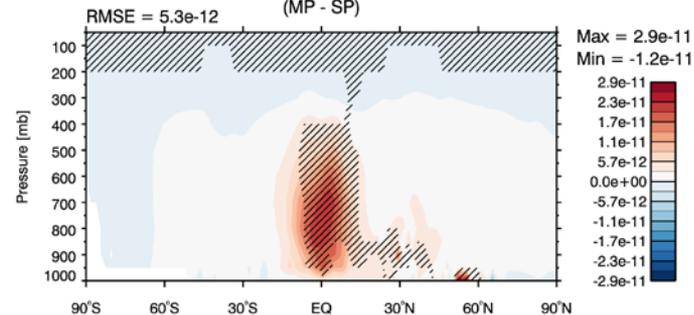
Difference

(MP - SP)

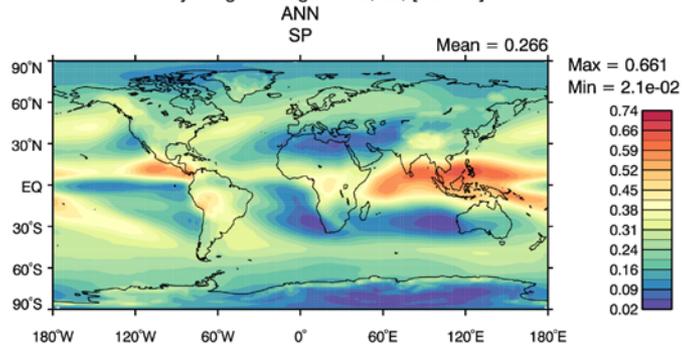


Difference

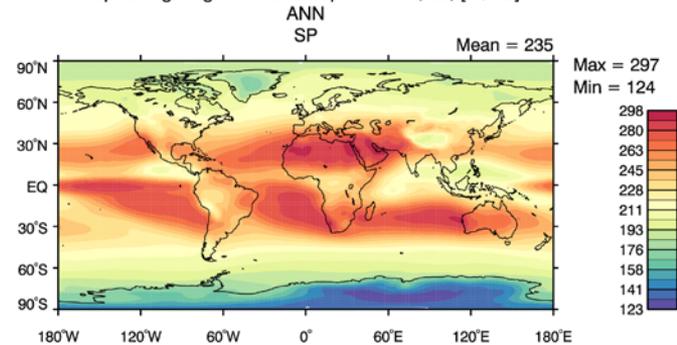
(MP - SP)



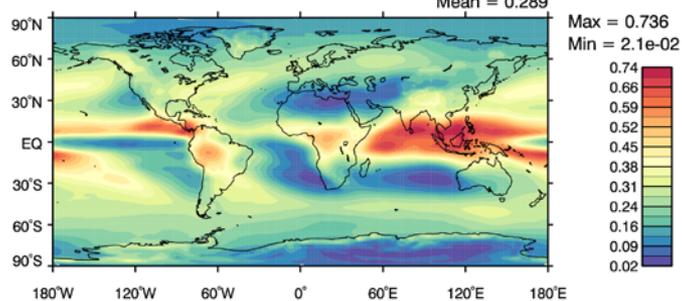
Vertically-integrated high cloud, 2d, [fraction]



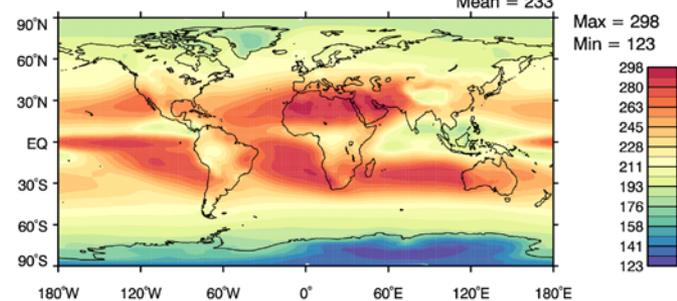
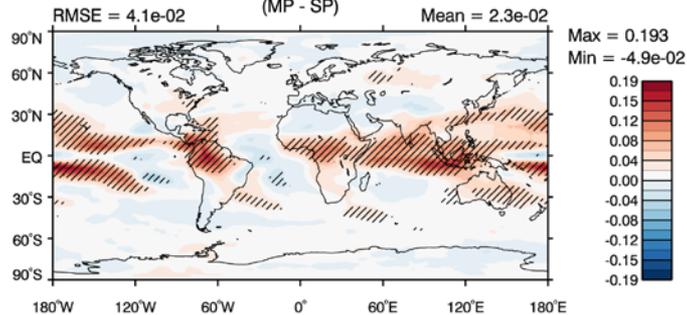
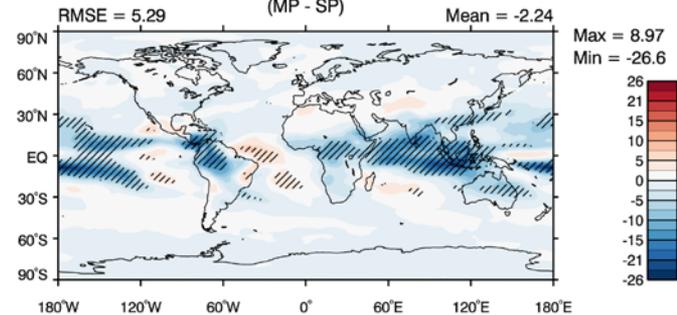
Upwelling longwave flux at top of model, 2d, [W/m2]

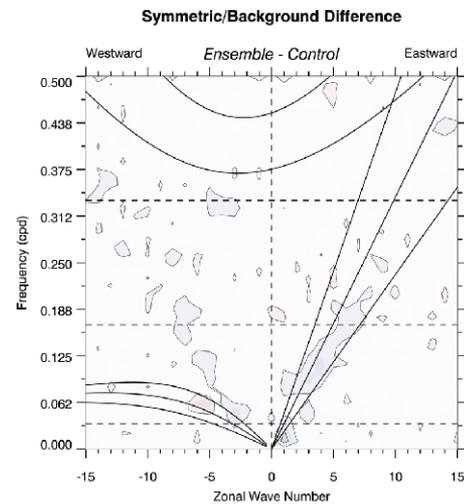
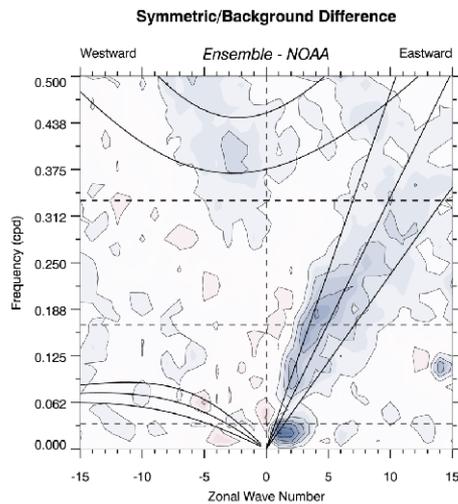
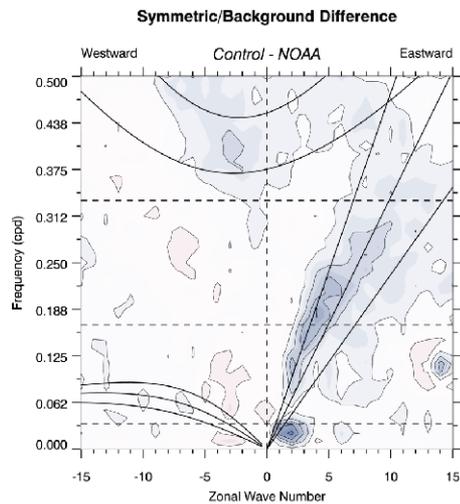
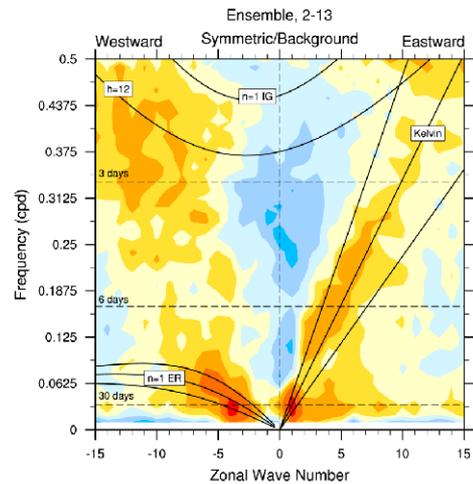
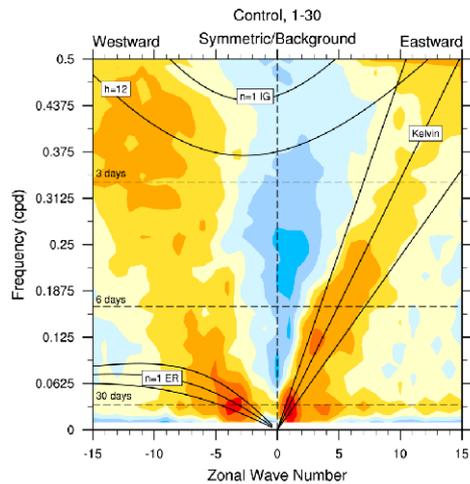
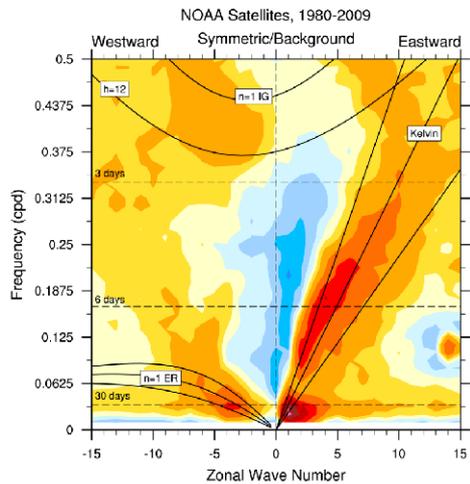


MP



MP

Difference
(MP - SP)Difference
(MP - SP)



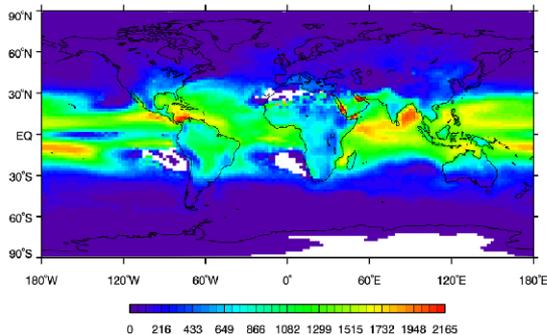
Locating Strong, Unpredictable Convection

- Exploring, and potentially exploiting, the range of possible realizations using identical large-scale states.
- Use spread measures as indicators of predictability
 - Coefficient of variation (standard deviation/mean, noise/signal)
 - Proportional variability (mean ratio of all combinations of data pairs)
- Look for associated large-scale properties
 - Since we can better predict the large-scale, perhaps we can predict potential predictability.

Locating Strong, Unpredictable Convection

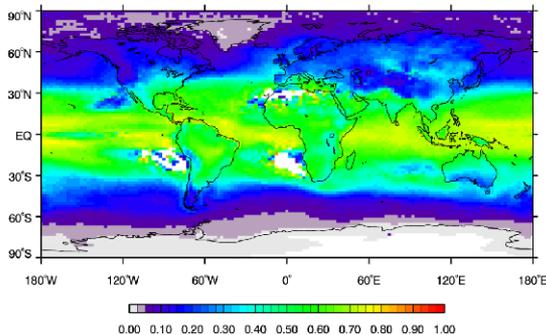
Data: 3648 days

Pseudoadiabatic convective available potential energy [J/kg]



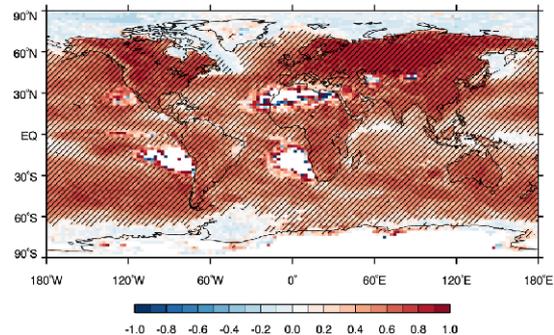
ANN, precip > 5 mm day⁻¹

Proportional Variability



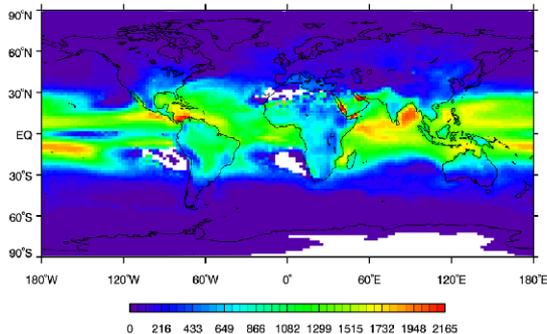
$r_{SIT} = 0.788$, $r_s = 0.907$, $r_{GMT} = 0.465$, $r_{GMT95} = 0.511$

Local Temporal Correlation (sig @ 95%)



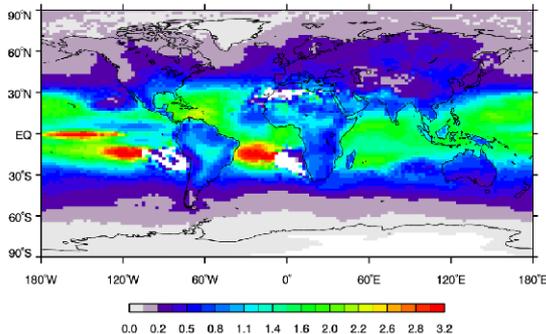
Data: 3648 days

Pseudoadiabatic convective available potential energy [J/kg]



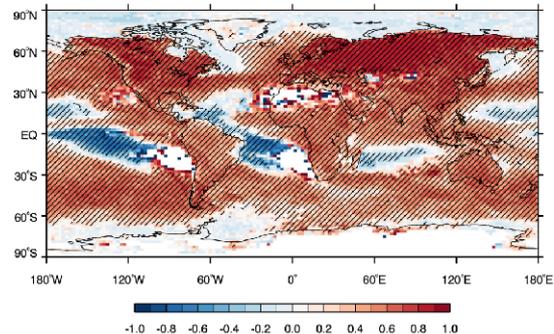
ANN, precip > 5 mm day⁻¹

COV



$r_{SIT} = 0.716$, $r_s = 0.842$, $r_{GMT} = 0.366$, $r_{GMT95} = 0.430$

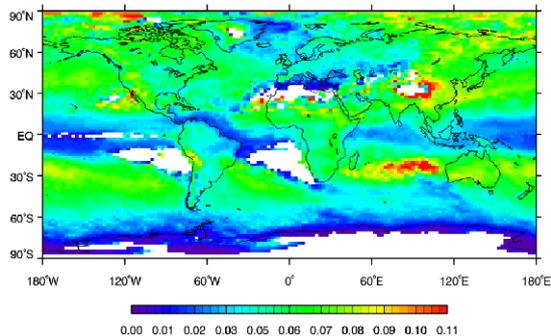
Local Temporal Correlation (sig @ 95%)



Locating Strong, Unpredictable Convection

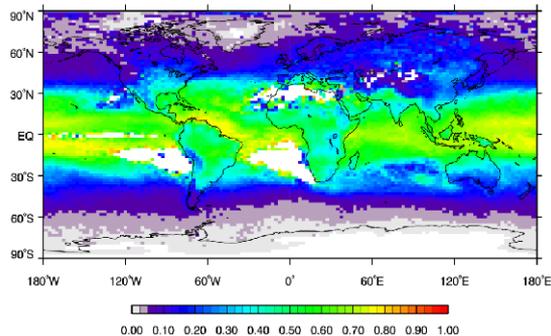
Data: 2133 days

Grid box averaged cloud liquid amount [g kg⁻¹], 600 mb



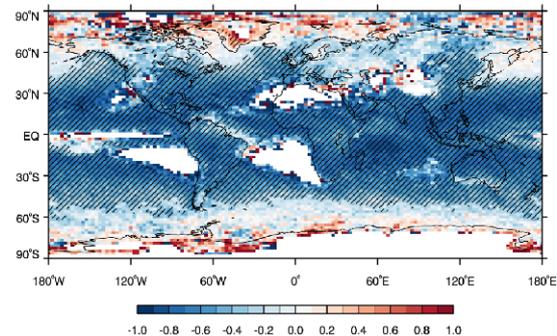
ANN, precip > 10 mm day⁻¹

Proportional Variability



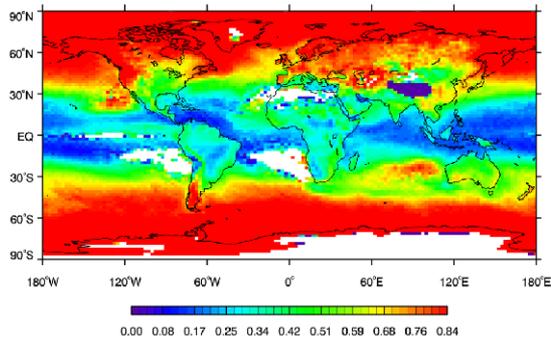
$r_{SIT} = -0.445$, $r_S = -0.083$, $r_{GMT} = -0.452$, $r_{GMT95} = -0.576$

Local Temporal Correlation (sig @ 95%)



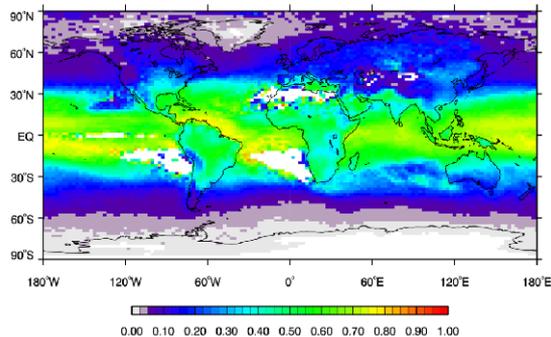
Data: 3648 days

Vertically-integrated low cloud [fraction]



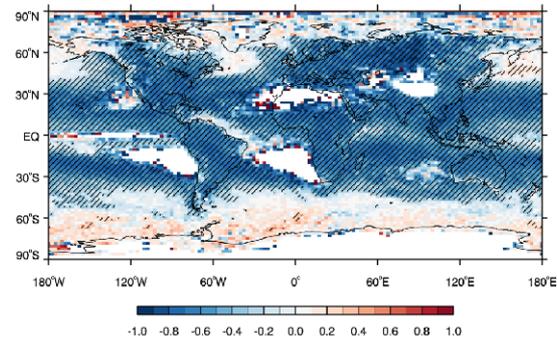
ANN, precip > 10 mm day⁻¹

Proportional Variability



$r_{SIT} = -0.803$, $r_S = -0.931$, $r_{GMT} = -0.413$, $r_{GMT95} = -0.528$

Local Temporal Correlation (sig @ 95%)



Quasi-Conclusions

- Deterministic parameterizations produce weaker, more frequent precipitation.
- This leads to more high cloud, weaker OLR, and lower precipitation amounts.
- Deterministic parameterizations may be able to represent intraseasonal variability. (Even while degrading some statistics)
- Spread around the response is not simply proportional to the mean, has spatial structure, and is associated with L-S features.
- We have useful tools to explore these issues.