

# Clouds in the Climate System:

*Why is this such a difficult problem,  
and where do we go from here?*

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*CERES Science Team Meeting*

*April 29, 2009*

# Collaborators

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# 4<sup>th</sup> IPCC: Key Uncertainties

- “Cloud feedbacks (particularly from low clouds) remain the largest source of uncertainty [to climate sensitivity].”
- “... processes leading to modification of cloud properties by aerosols [are] not well understood and ... indirect radiative effects are poorly determined.”
- “Surface and satellite observations disagree on total and low-level cloud changes over the ocean.”
- “Large uncertainties remain about how clouds might respond to global climate change.”
- “Cloud feedbacks are the primary source of intermodel differences in equilibrium climate sensitivity...”

# Why is this a difficult problem?

- We have no stable system to monitor global cloudiness and radiation on multidecadal time scales
- Cloud and radiation measurements are insufficiently integrated with associated meteorological processes
- Wrong priorities in climate modeling efforts

# Why is $d\text{Cloud}/dT$ so uncertain?

*Unlike other climate feedbacks, temperature does not exert a direct influence on cloud feedbacks*

- Ice/snow albedo feedback  $\rightarrow$  ice/snow melts for  $T > 0^\circ\text{C}$
- Water vapor feedback  $\rightarrow$  saturation humidity strongly varies with temperature
- Cloud feedback  $\rightarrow$  relative humidity  $> 100\%$  is under dynamical control

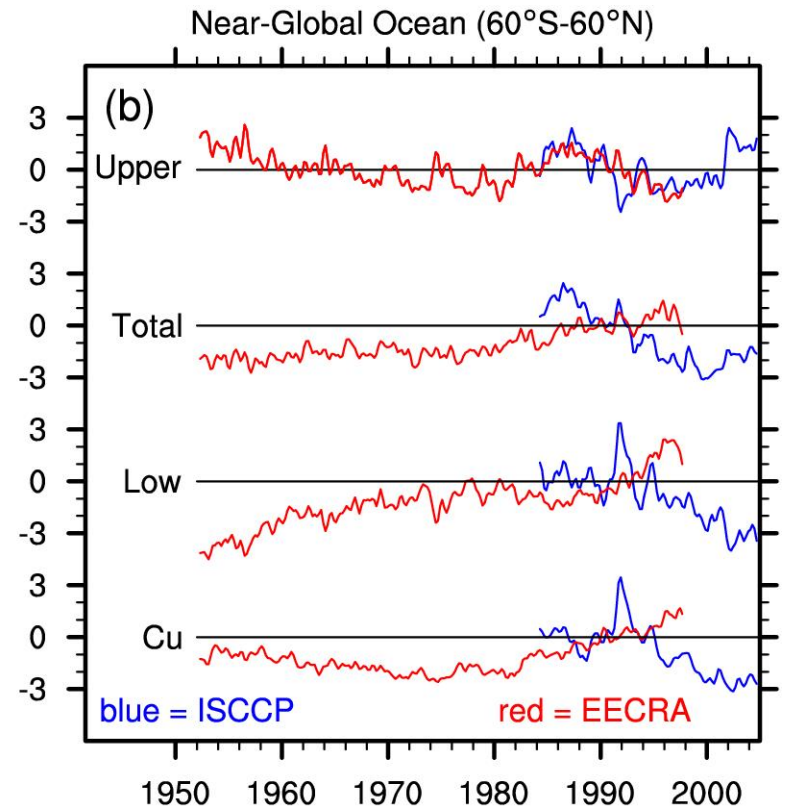
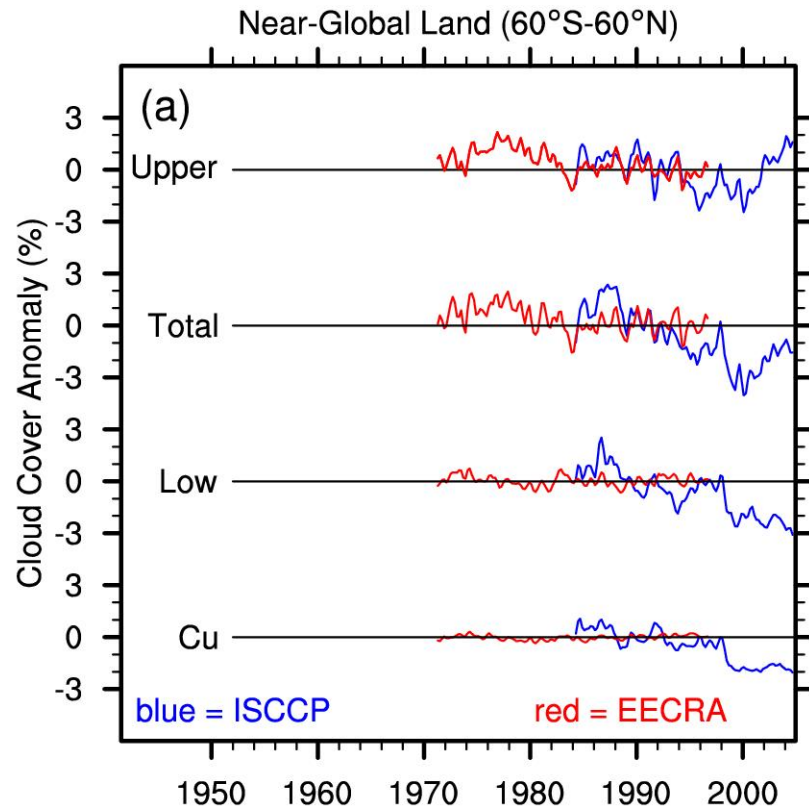
# How to determine $d\text{Cloud}/dT$ ?

Approximate as change in cloud during recent decades of rapid warming

*Some weaknesses...*

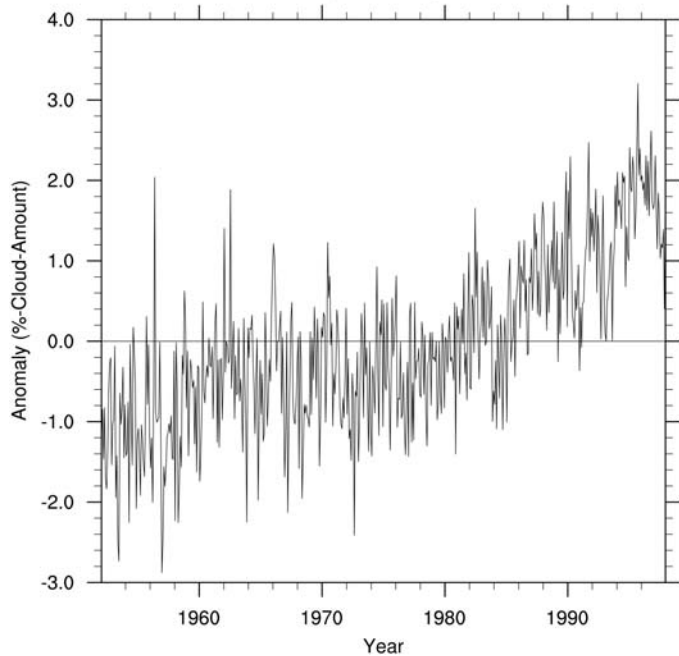
- Not an equilibrium response
- Cloud changes may be influenced by unforced dynamical variability instead of solely temperature
- Lack of a homogeneous observational record

# Surface and Satellite Cloud

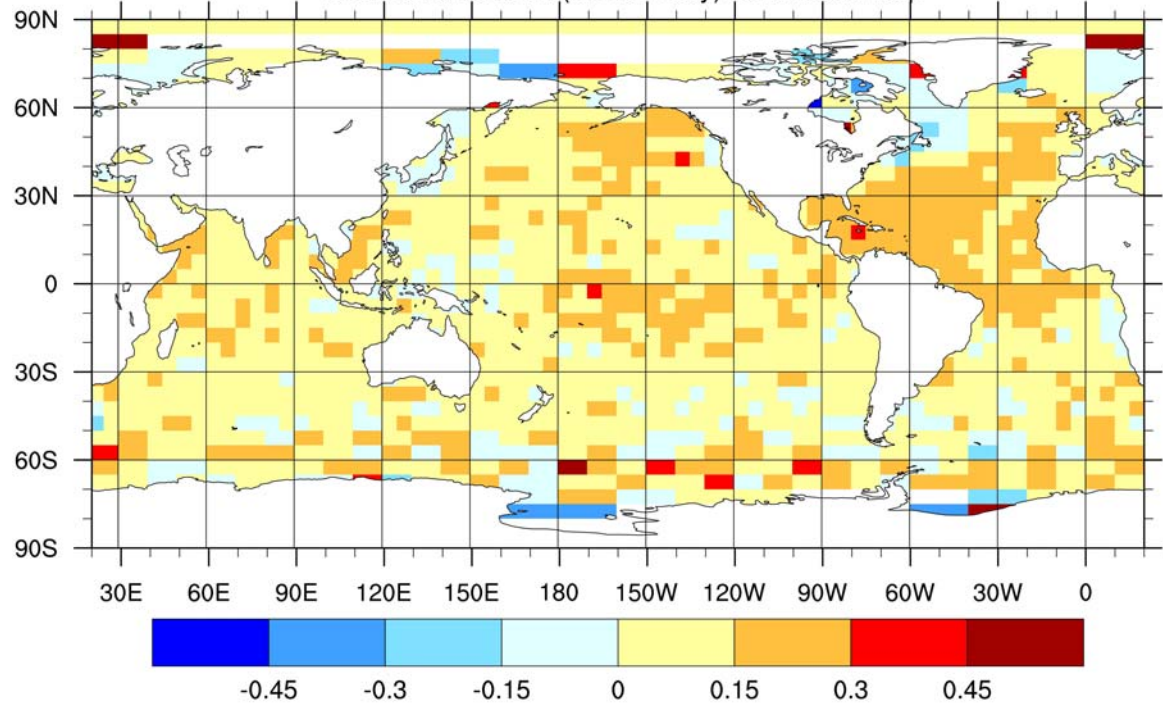


# Surface Cloud Record

Total Cloud Amount (Ocean Only) Global Time Series



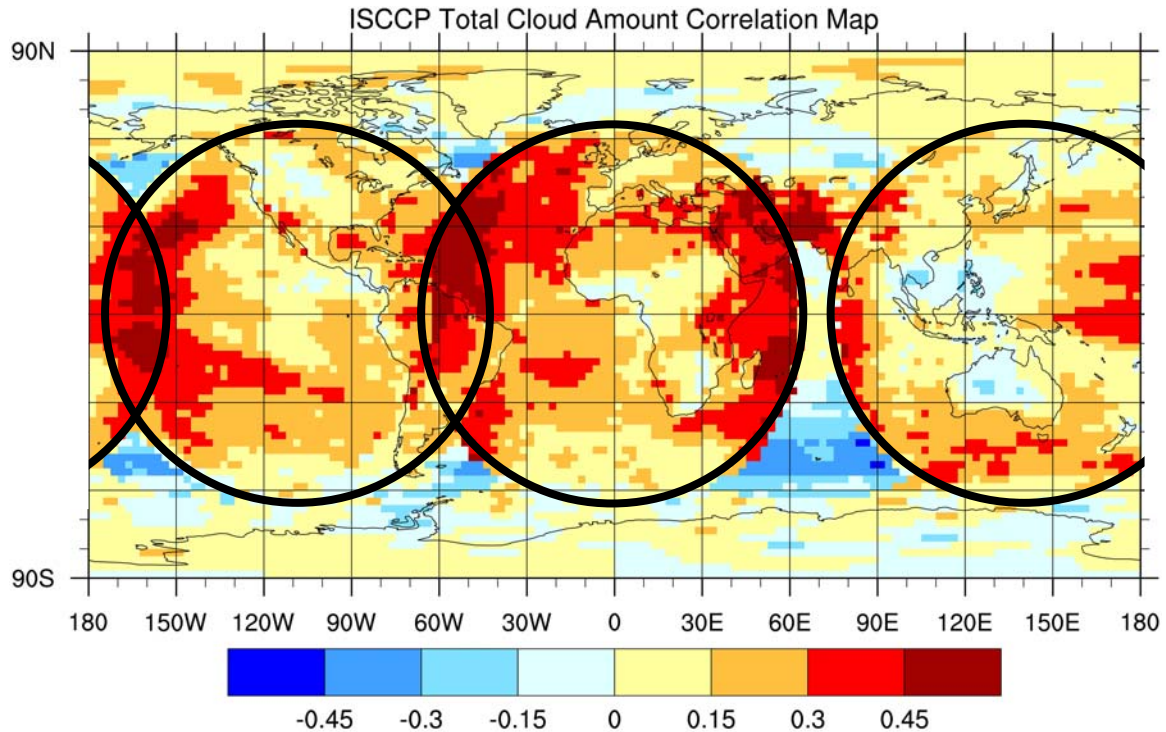
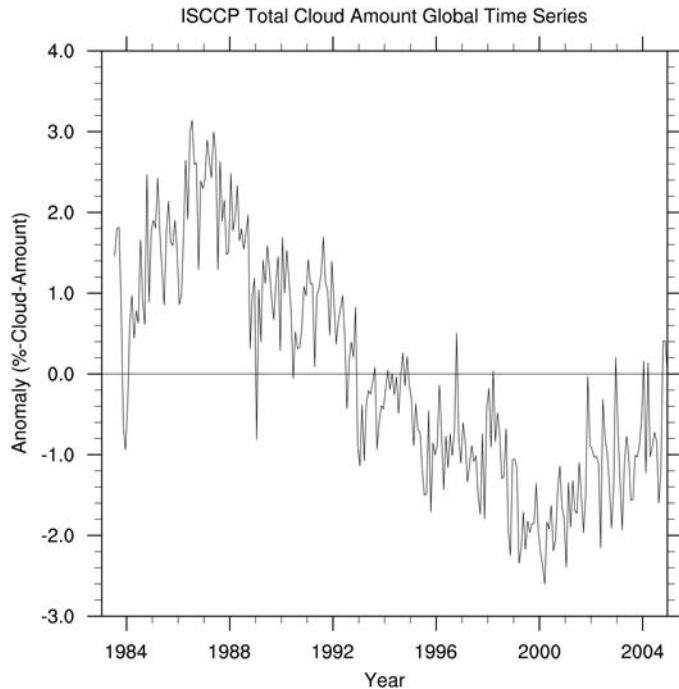
Total Cloud Amount (Ocean Only) Correlation Map



Low-level and especially cumulus cloud types are the greatest contributors to the upward trend in total cloud cover.



# Satellite Cloud Record



Low-level cloudiness is the largest contributor to the apparent artifact in total amount (not shown).

# Another method for $d\text{Cloud}/dT$

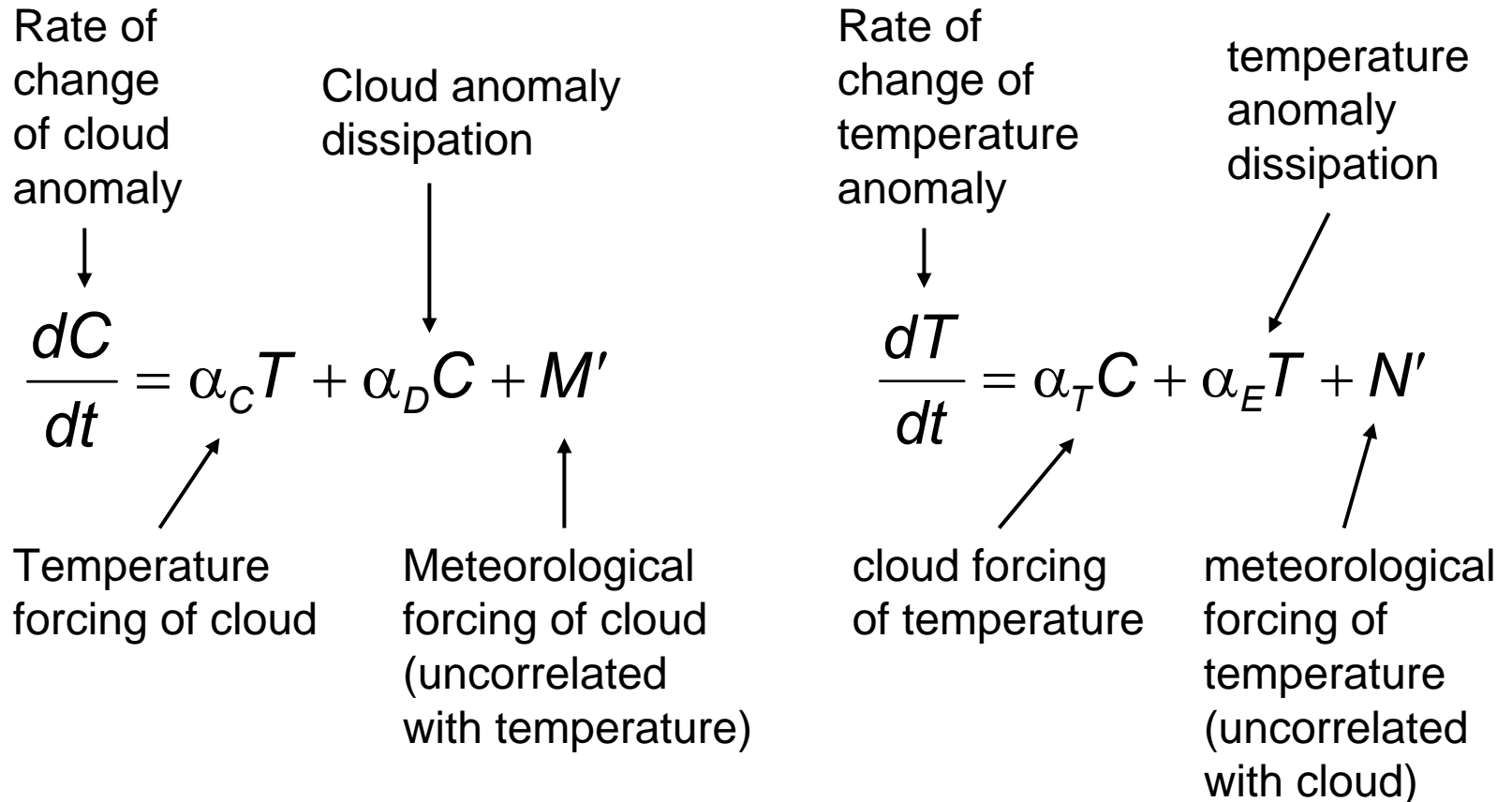
Cloud change associated with temperature change on short time scales (daily to monthly)

*Some weaknesses...*

- Not an equilibrium response
- Processes dominant on short time scales may not be dominant on long time scales
- Cloudiness and temperature are strongly and jointly influenced by dynamical variability

# Conceptual Model

## *Simple Cloud-Temperature-Meteorology System*



# Conceptual Model

*Discretize...*

Set  $\Delta t$  to  $-\alpha_D^{-1}$  (cloud anomaly damping time scale)

$$C_{t+\Delta t} = \beta_C T_t + M_t$$

$$T_{t+\Delta t} = (1 + \beta_E) T_t + \beta_T C_t + N_t$$

$\beta_C$  cloud response factor ( $\beta_C < 0$  for Sc)

$\beta_T$  cloud radiative forcing factor ( $\beta_T < 0$  for Sc)

$\beta_E$  temperature damping factor ( $-1 < \beta_E < 0$ )

# Cloud-Temperature Regression

Calculate regression of  $C_t$  on  $T_t \dots$

For simplicity, set  $\beta_E = 0$  and  $\text{cov}(T, N) = 0$

$$\frac{\text{cov}(C, T)}{\text{var}(T)} = \overset{\text{Term 1}}{\frac{\beta_C}{1 - \beta_C \beta_T}} + \overset{\text{Term 2}}{\frac{\beta_T}{1 - \beta_C \beta_T} \frac{\text{cov}(C, M)}{\text{var}(T)}} + \overset{\text{Term 3}}{\frac{1}{1 - \beta_C \beta_T} \frac{\text{cov}(M, N)}{\text{var}(T)}}$$

# Cloud-Temperature Regression

Term 1  $\frac{\beta_C}{1 - \beta_C \beta_T}$

If  $\beta_T = 0$       no cloud radiative forcing

→ Regression of  $C$  on  $T = \beta_C$

If  $\beta_T \neq 0$       yes cloud radiative forcing

→ |Regression of  $C$  on  $T$ |  $> |\beta_C|$

*Overestimation of cloud response factor magnitude*  
*Overestimation of cloud feedback*

# Cloud-Temperature Regression

Term 2

$$\frac{\beta_T}{1 - \beta_C \beta_T} \frac{\text{cov}(C, M)}{\text{var}(T)}$$

If  $\text{cov}(C, M) \neq 0$        $C_{t+\Delta t}$  and  $C_t$  are autocorrelated  
through long  $M$  timescale

→ Even if  $\beta_C = 0$ , coincident  $C$ - $T$  relationship because  
previous cloud radiatively forced current temperature

→ Regression of  $C$  on  $T$  has additional negative factor

*Cloud response factor appears more negative*

*Cloud feedback appears more positive*

# Cloud-Temperature Regression

Term 3

$$\frac{1}{1 - \beta_C \beta_T} \frac{\text{cov}(M, N)}{\text{var}(T)}$$

If  $\text{cov}(M, N) \neq 0$  meteorology influencing cloud is correlated with meteorology influencing temperature

→ Even if  $\beta_C = 0$  and  $\beta_T = 0$ , coincident  $C$ - $T$  relationship due to joint forcing by meteorology

*Effect on apparent cloud response factor and cloud feedback depends on nature of meteorological forcing*



# Observed Cloud Feedback

- Meteorological memory can mix cloud radiative impact on temperature with cloud response to temperature
- Averaging over time (e.g., monthly means) will exacerbate the above effect
- Is the above effect important? Need better quantification
- *It is essential to consider joint meteorological forcing of cloud and temperature*

# Bias due to Meteorology

For simplicity, set  $\beta_T = 0$  and  $\text{cov}(C, M) = 0$

(no cloud radiative forcing of temperature, no memory)

$$\beta_c = \frac{\text{cov}(C, T)}{\text{var}(T)} - \frac{\text{cov}(M, N)}{\text{var}(T)}$$

The diagram illustrates the components of the equation for  $\beta_c$ . An arrow points from the text "what we want" to  $\beta_c$ . Another arrow points from "what we measure" to the  $\text{var}(T)$  term in the first fraction. A third arrow points from "what we don't know" to the  $\text{var}(T)$  term in the second fraction.

what we want

what we measure

what we don't know

# Bias due to Meteorology

*How can the impact of  $\text{cov}(M, N)$  be reduced?*

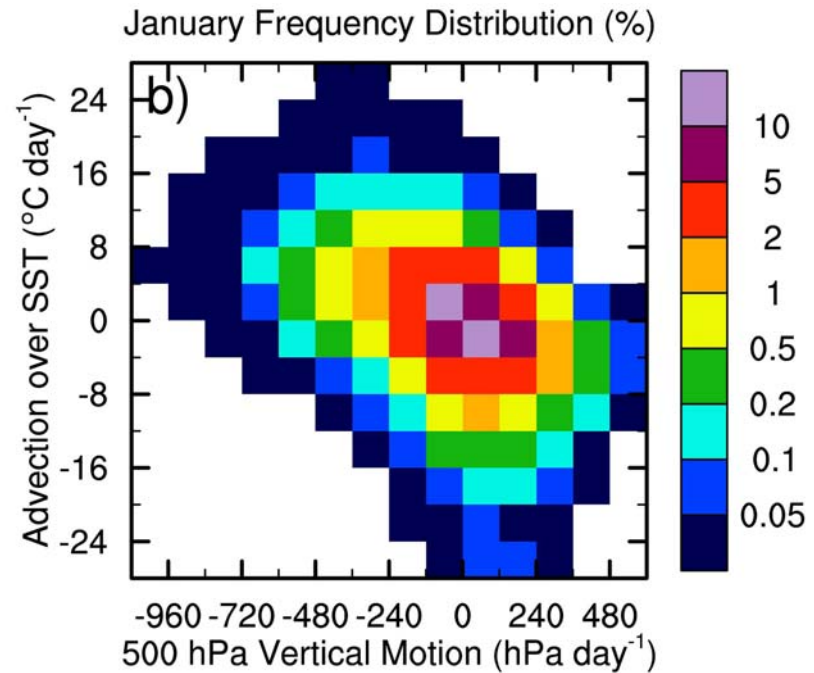
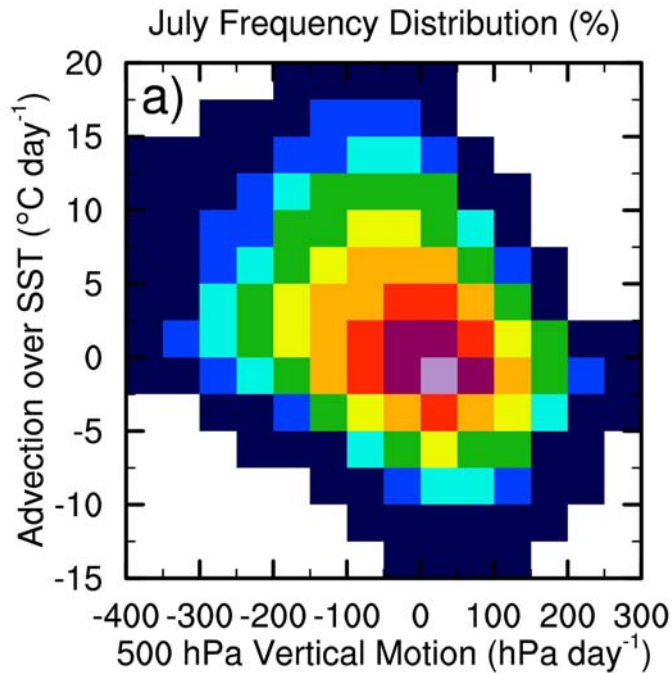
- Select relevant parameters to represent influential meteorological processes
- Bin cloud and temperature data into small intervals of the parameters (e.g., hold meteorology “constant”)
- Examine  $\text{cov}(C, T)$  separately for each bin
- *What if an important parameter is left out?*
- *What if a process cannot be fully represented by a simple parameter?*

Meteorological influence will always be underestimated!

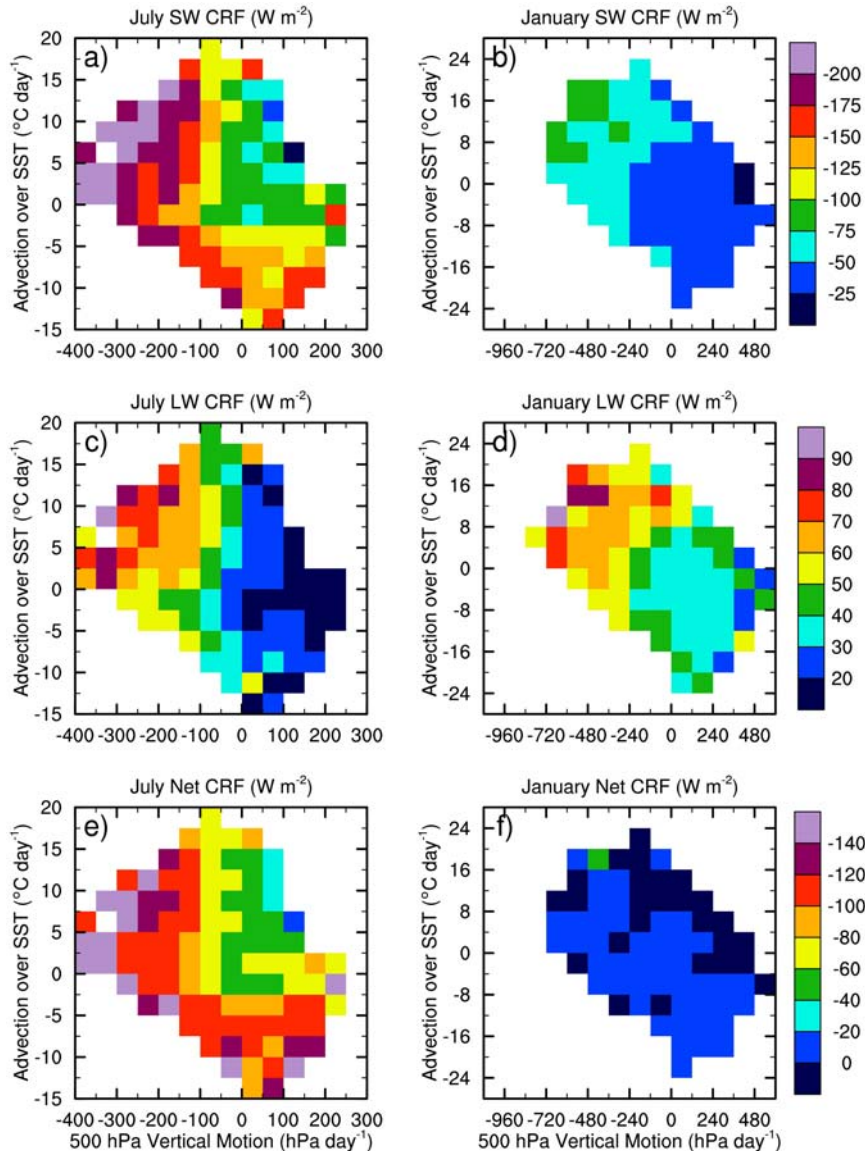
# Cloud, SST, and Advection

- Synoptic variability causes atmospheric flow over the North Pacific SST gradient to frequently change.
- Horizontal advection and vertical motion have large impacts on cloud and temperature
- Bin daily cloud and CRF on according to  $\omega_{500}$  and SST advection (defined as  $-\mathbf{V}_{1000} \cdot \nabla \text{SST}$ )
- Examine composite difference in cloud and CRF between warm and cold temperature for each bin

# SST Advection- $\omega_{500}$ Histograms (Freq)



# SST Advection- $\omega_{500}$ Histograms (CRF)

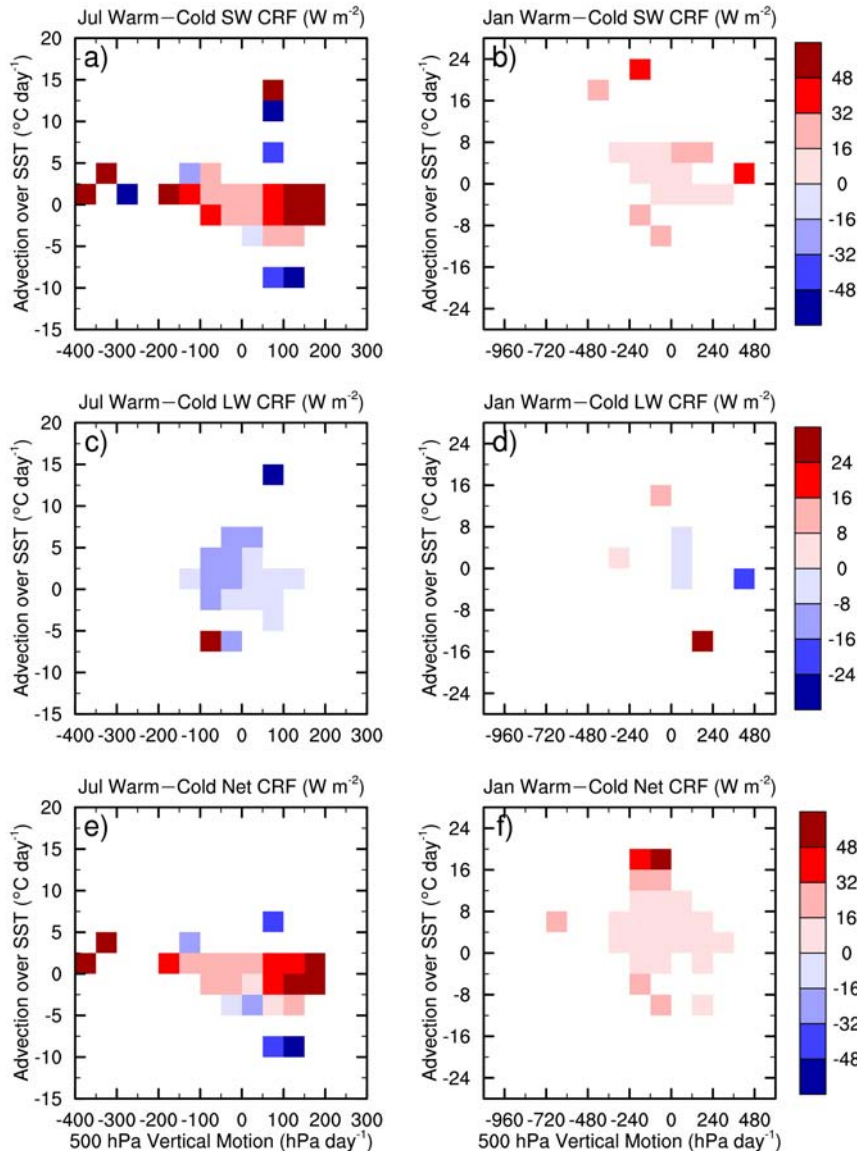


large SW CRF for upward and cold/down quadrants (latter only for July)

Large LW CRF for upward motion

large net CRF for upward and cold/down quadrants in July

# Adv- $\omega$ Histograms (Warm–Cold CRF)



SW CRF more positive  
(weaker negative) for warm  
conditions under most  
dynamical states

LW CRF more negative  
(weaker positive) for warm  
conditions under most  
dynamical states

net CRF more positive  
(weaker negative) for warm  
conditions under most  
dynamical states

# Average Warm-Cold CRF

$\omega_{500}$ ,  $-\mathbf{V}_{1000} \cdot \nabla \text{SST}$ , and vertical stratification held constant (as much as possible)

Month	SW CRF (W m <sup>-2</sup> per K)	LW CRF	Net CRF
January	<b>+4.4</b>	<b>-1.0</b>	<b>+3.4</b>
July	<b>+9.4</b>	<b>-2.5</b>	<b>+6.9</b>

Cloud response to temperature suggests a positive cloud feedback

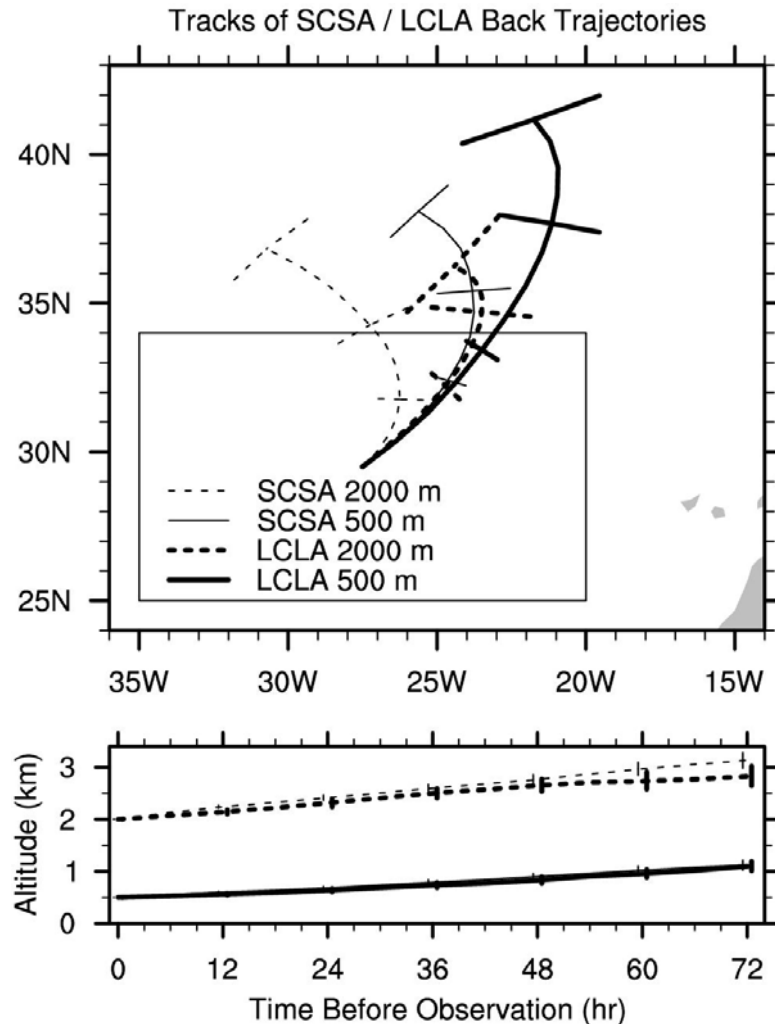
*But are there any additional meteorological processes that produce less cloud and warmer temperature?*



# Aerosol Influence on Cloud?

- Previous studies have reported a positive correlation between satellite-retrieved AOD and cloud fraction
- Does greater AOD mean more CCN, smaller cloud droplets, less precipitation loss, and more cloud?
- Or is greater AOD associated with greater cloud fraction due to meteorological conditions?
- Since clouds have a non-instantaneous response time, it is essential to consider *meteorological history*

# Aerosol Influence on Cloud?

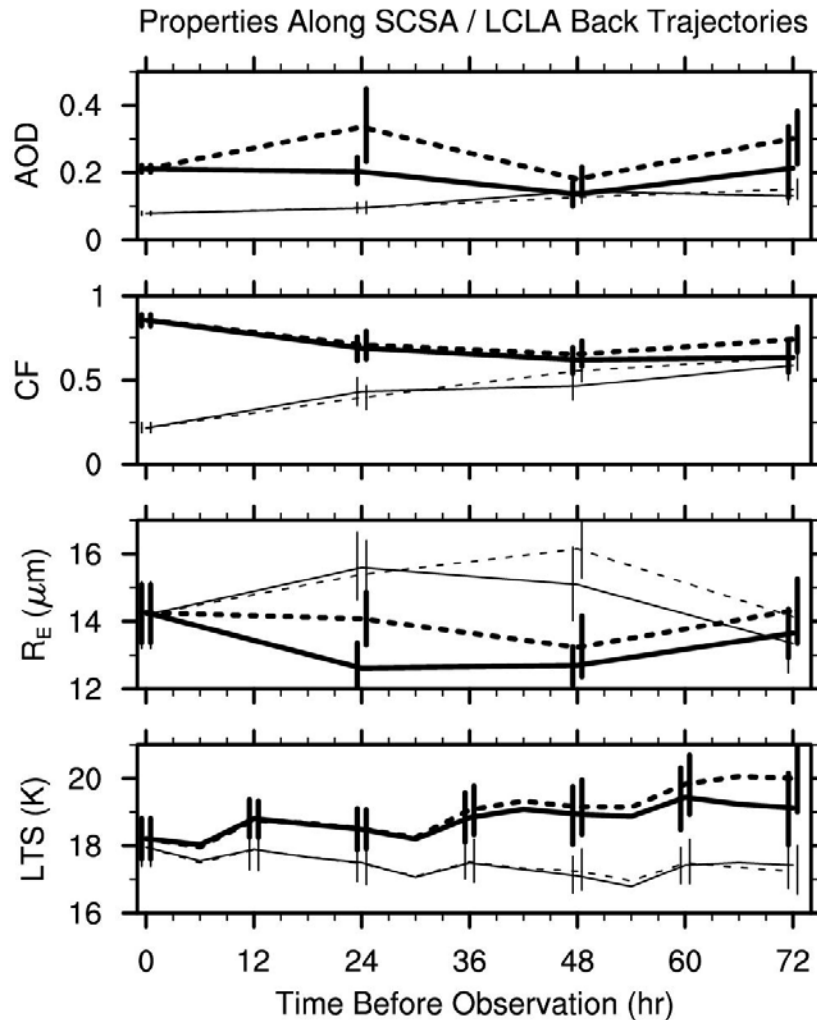


SCSA – Small Cloud,  
Small Aerosol

LCLA – Large Cloud,  
Large Aerosol

LCLA trajectories come  
from locations that are  
systematically closer to  
Europe

# Aerosol Influence on Cloud?



SCSA – thin lines

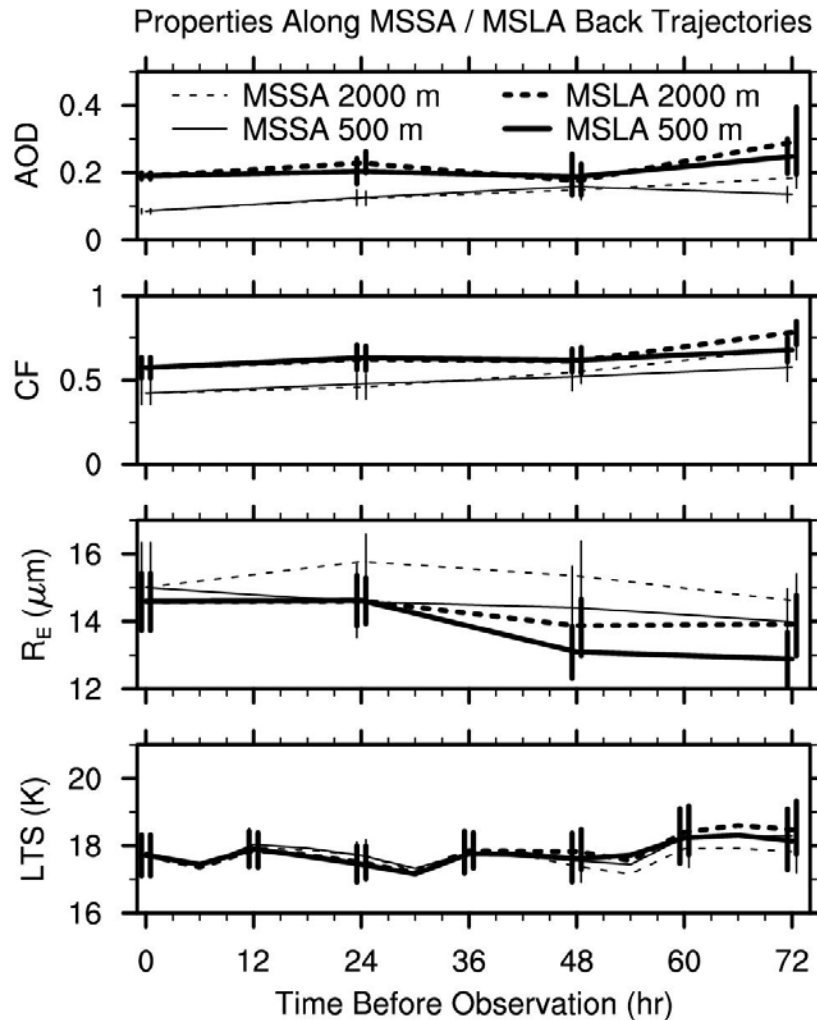
LCLA – thick lines

LTS – lower tropospheric static stability ( $\theta_{700} - \theta_{sfc}$ )

Previous studies show larger LTS promotes more cloud fraction

LCLA has larger LTS at -72 hours but not 0 hours

# Aerosol Influence on Cloud?



MSSA – Median Stability,  
Small Aerosol

MSLA – Median Stability,  
Large Aerosol

When LTS history is the  
same, much smaller  
cloud difference between  
small and large aerosol

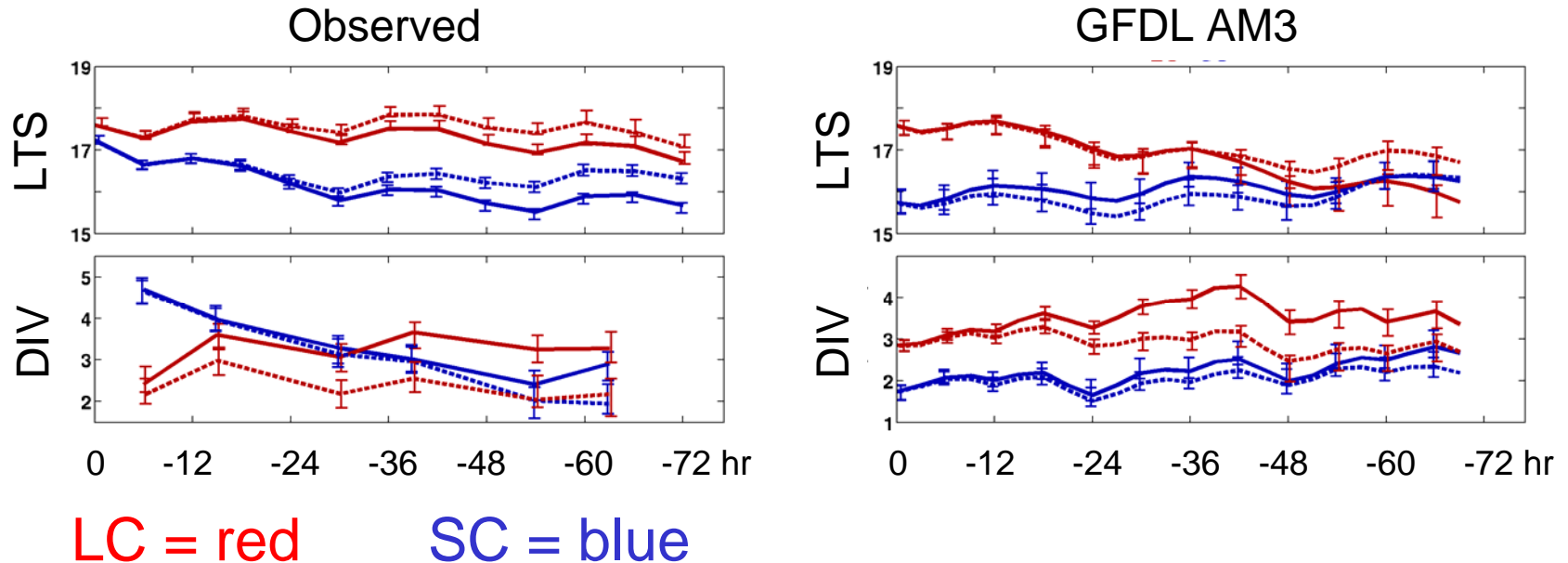
# Aerosol Influence on Cloud?

- Air mass source region is related to history of meteorological conditions experienced by a parcel
- This creates an apparent correlation between aerosol (from source region) and cloud (from meteorological history)
- The correlation between meteorological influence and cloud may be near-zero at  $t = 0$
- The preceding results are a lower limit for the confounding impact of meteorology and an upper limit for the influence of aerosol

# Evaluation of GCM Cloud

- Calculate trajectories for observed and model large cloud fraction (LC) and small cloud fraction (SC)
- Compare observed and model meteorological history for LC and SC composites
- Substantial differences are seen in the sign and timing of observed and simulated cloud relationships for the GFDL AM3

# Evaluation of GFDL AM3



Observed LC has strongest LTS at  $t = -36$  hr

Model LC has strongest LTS at  $t = 0$  hr

Observed LC has weak  $DIV_{sfc}$  at  $t = -6$  hr

Model LC has strong  $DIV_{sfc}$  at  $t = 0$  hr and  $t = -36$  hr

# Circulation and Cloud Feedbacks

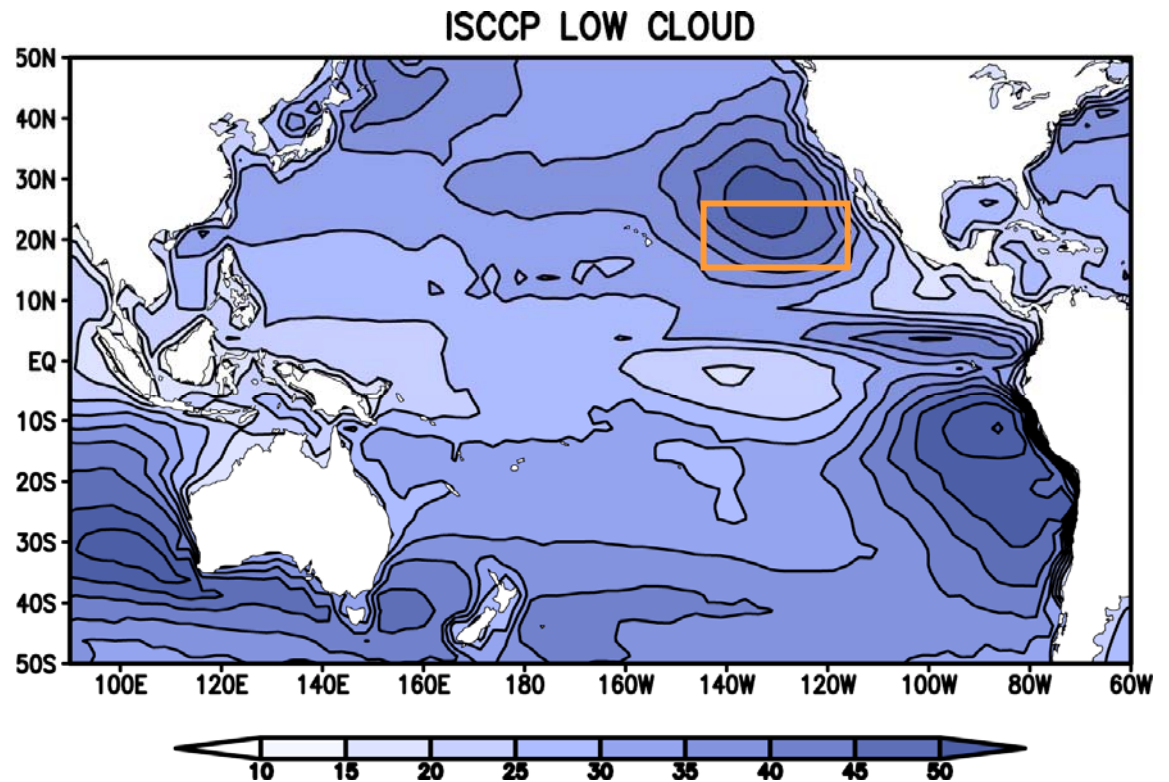
*What is the primary direct driver of cloud feedbacks in climate change?*

- Previous work has likely overestimated the impact of “thermodynamics” (temperature and lapse rate change)
- Atmospheric circulation change associated with global warming may instead play a leading role



# NE Pacific Decadal Variability

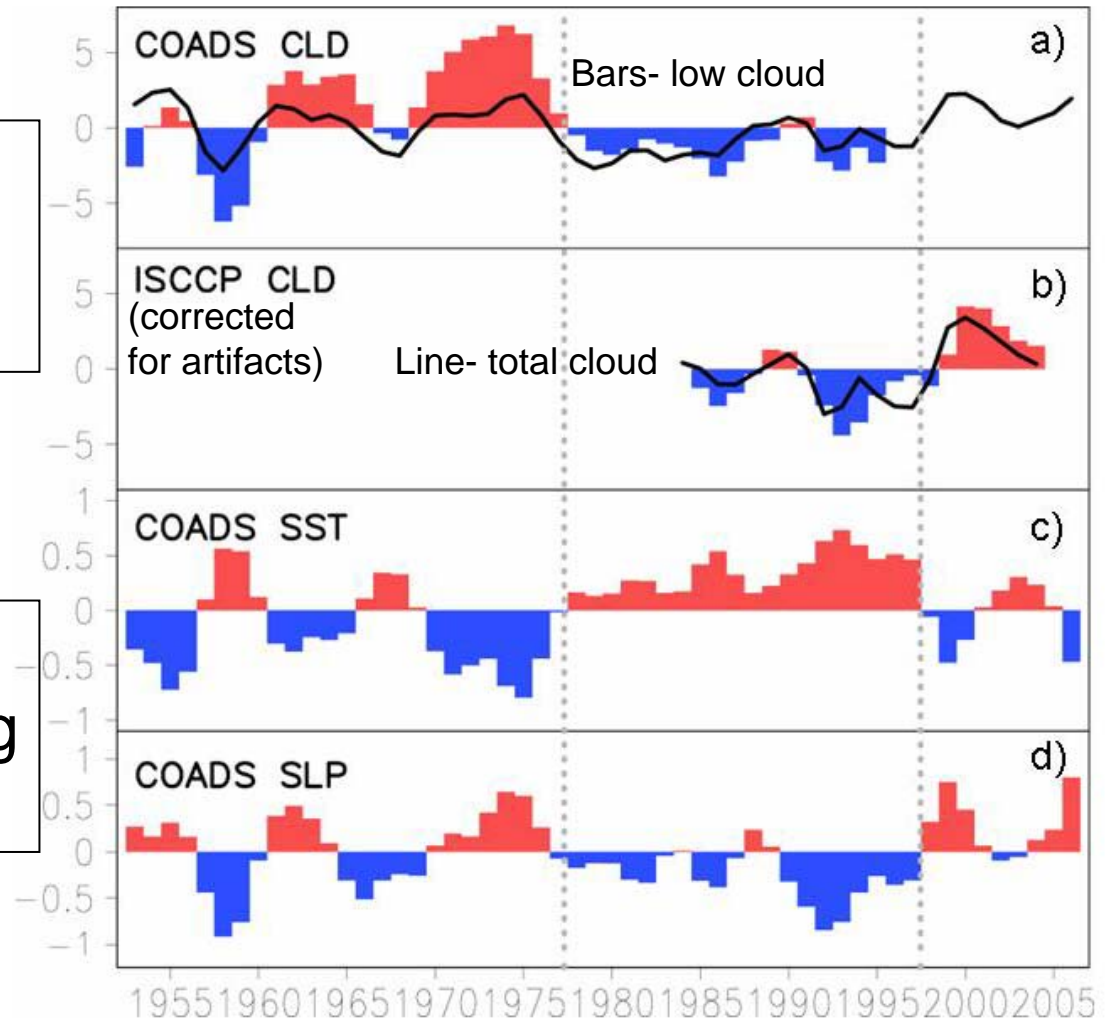
Does a cloud feedback promote decadal variability in SST and circulation?



# NE Pacific Decadal Variability

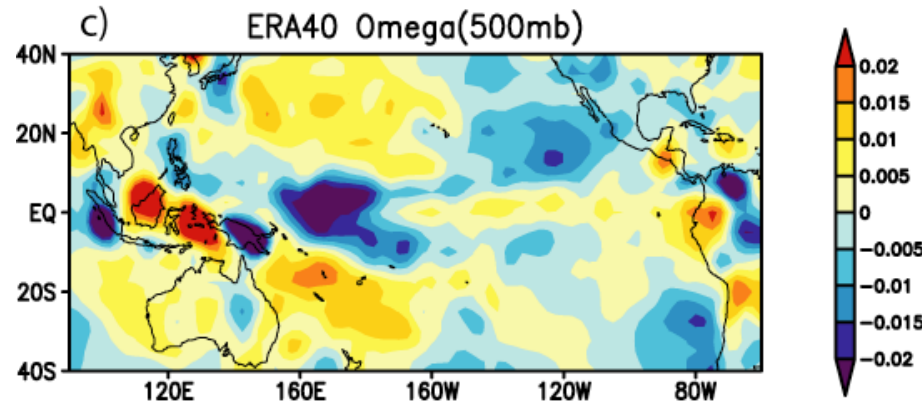
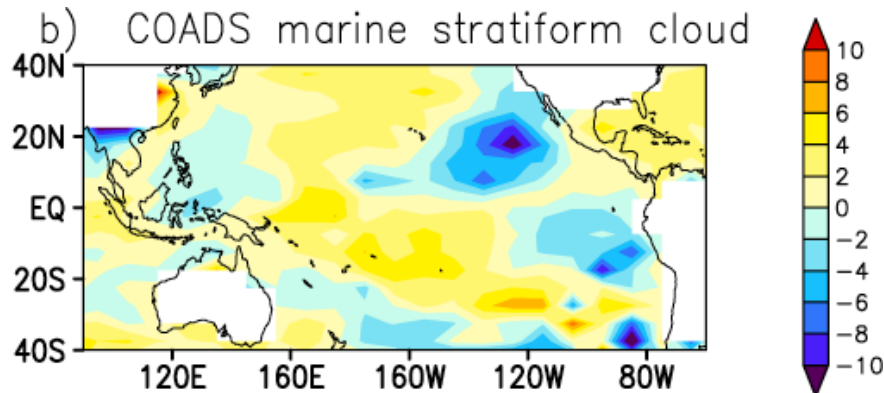
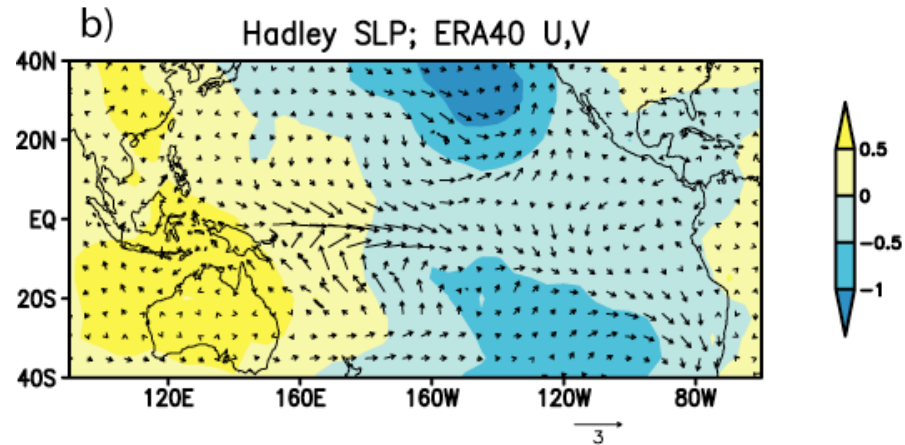
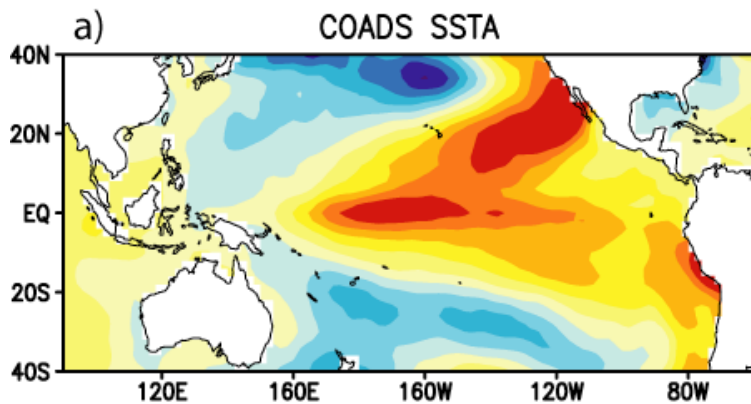
warm SST  
weak SLP  
weak wind

less stratocumulus  
more ocean heating  
less BL cooling



# NE Pacific Decadal Variability

*Basin-wide regression on NE Pacific SST time series*



# Is this feedback present in IPCC AR4 models?

Correct sign  
 $r$  and robust  
simulation

wrong sign  
 $r(\text{cloud}, \omega_{500})$

wrong sign  
 $r(\text{cloud}, \text{SLP})$

models with  
wrong sign  
 $r(\text{cloud}, \text{LTS})$

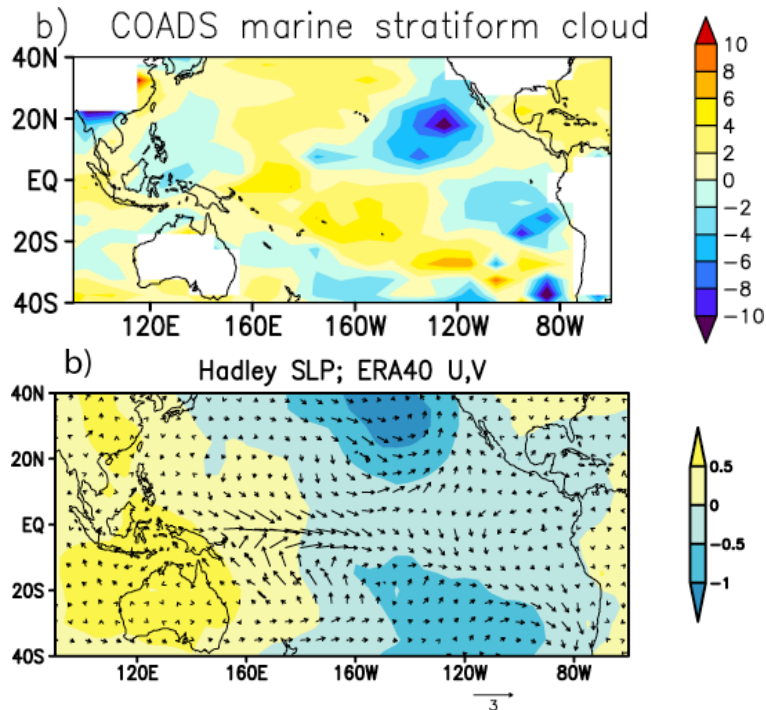
models with  
wrong sign  
 $r(\text{cloud}, \text{SST})$

	SST	LTS	SLP	Vs	$\omega$ 500
ISCCP low	-0.78	0.51	0.64	-0.56	0.50
COADS msc	-0.92	0.43	0.80	-0.63	0.62
ukmo_hadgem1	-0.81	0.84	0.65	-0.55	0.39
inmcm3_0	-0.77	0.37	0.58	-0.57	0.14
bccr_bcm2_0	-0.60	0.43	0.44	-0.58	NA
mri_cgcm2_3_2a	-0.59	0.14	0.32	-0.31	-0.60
miroc3_2_hires	-0.91	0.54	-0.03	-0.71	-0.10
cccma_cgcm3_1_t63	-0.86	0.01	0.52	-0.44	0.20
gfdl_cm2_0	-0.69	0.06	0.52	-0.73	-0.42
cccma_cgcm3_1	-0.80	-0.08	0.35	-0.51	-0.14
ncar_ccsm3_0	-0.76	-0.22	0.69	NA	-0.23
cnrm_cm3	-0.73	-0.24	0.54	-0.45	-0.54
ipsl_cm4	-0.53	-0.16	0.25	-0.40	-0.32
csiro_mk3_5	-0.48	-0.47	0.20	-0.56	NA
ukmo_hadcm3	-0.44	-0.17	0.33	-0.29	-0.43
miub_echo_g	-0.35	NA	0.13	-0.24	NA
gfdl_cm2_1	-0.31	-0.38	0.05	-0.19	-0.56
mpi_echam5	-0.23	-0.44	-0.06	-0.15	-0.70
ingv_echam4	-0.22	-0.12	-0.16	NA	NA
miroc3_2_medres	-0.13	-0.08	-0.04	-0.28	-0.67
csiro_mk3_0	-0.12	-0.12	-0.23	NA	NA
giss_aom	0.12	-0.63	-0.39	0.32	-0.67
iap_fgoals1_0_g	0.22	-0.43	-0.24	0.47	-0.89
giss_model_e_h	0.34	0.10	0.10	0.27	-0.81
giss_model_e_r	0.39	-0.04	0.003	0.22	-0.58
ncar_pcm1	0.45	-0.61	-0.51	NA	-0.76

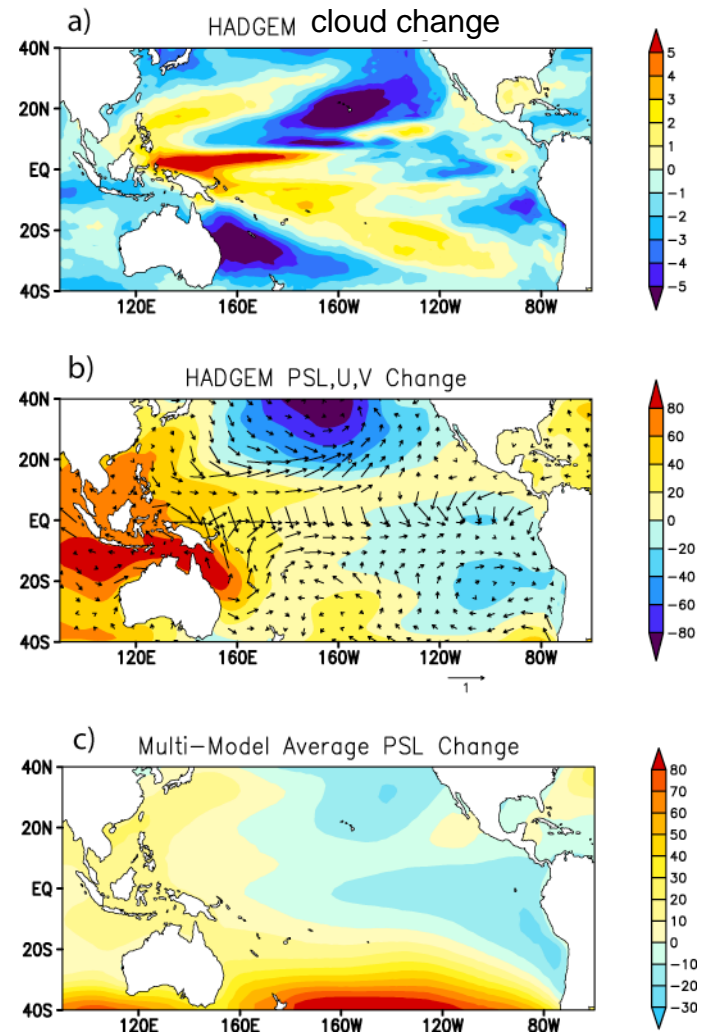
Observed  $r$   
NE Pacific  
cloud and  
meteorology

# HadGEM1 2×CO<sub>2</sub> Change

## Observed Decadal



## 2×CO<sub>2</sub> Simulation



2×CO<sub>2</sub> cloud and circulation changes resemble observed decadal cloud and circulation changes

# Circulation and Cloud Feedbacks

- On decadal time scales, decreased stratocumulus associated with warmer SST and weaker circulation
- Likely positive cloud feedback due to solar warming of ocean and reduced cooling of atmospheric BL
- Only one robust IPCC AR4 model reproduces correct sign for all 5 cloud-meteorological correlations
- This model exhibits stratocumulus decrease and weaker circulation for  $2\times\text{CO}_2$  that resembles observed pattern

# Where do we go from here? (1)

- Develop a stable observational system to monitor global cloudiness and radiation on decadal time scales
- Correct (to the extent possible) the historical cloud and radiation record
  - *this includes reprocessing data long after a mission has ended*
  - *integrate satellite and non-satellite datasets (surface observations, ocean heat content, reanalysis meteorology)*

# Where do we go from here? (2)

- Integrate meteorological conditions with cloud and radiation measurements
  - *detailed information of cloud properties is not sufficient to characterize processes and feedbacks*
  - *daily rather than monthly data is fundamental*
- Understand that the instantaneous cloud and radiation state results from a history of meteorological processes
  - *coincident cloud and meteorological correlations may not show true relationships*



# Where do we go from here? (3)

- Assimilate cloud and radiation measurements into global models for best integration
  - *this is a very difficult task due to model cloud biases*
- Focus on essential cloud, convection, and turbulence parameterization development
  - *it doesn't make sense to add aerosol indirect effects when basic cloud processes are not credible*

Thank You!

# Additional Slides

# Conceptual Model for Climate

*Equilibrium climate response to external radiative forcing*

$$0 = \Delta R + \frac{1}{\lambda_{BB}} \Delta T_s + \sum_k \frac{\partial F}{\partial I_k} \frac{dI_k}{dT_s} \Delta T_s$$

$\Delta R$  external radiative forcing change

$\Delta T_s$  surface temperature change

$1/\lambda_{BB}$  increase of blackbody emission

$F_k$  radiation flux from internal parameter  $I_k$

# Conceptual Model for Climate

*Equilibrium climate response to external radiative forcing*

$$\Delta T_s = \frac{-\Delta R}{\frac{1}{\lambda_{BB}} + \sum_k \frac{1}{\lambda_k}} \quad \text{where} \quad \frac{1}{\lambda_k} = \frac{\partial F}{\partial I_k} \frac{dI_k}{dT_s}$$

$$\Delta T_s = -\Delta R \lambda_{clim} \quad \text{where} \quad \frac{1}{\lambda_{clim}} = \frac{1}{\lambda_{BB}} + \sum_k \frac{1}{\lambda_k}$$

$\lambda_{clim}$  climate sensitivity

# What is $\lambda_{cloud}$ ?

$$\frac{1}{\lambda_{cloud}} = \frac{\partial F}{\partial I_{cloud}} \frac{dI_{cloud}}{dT_s}$$

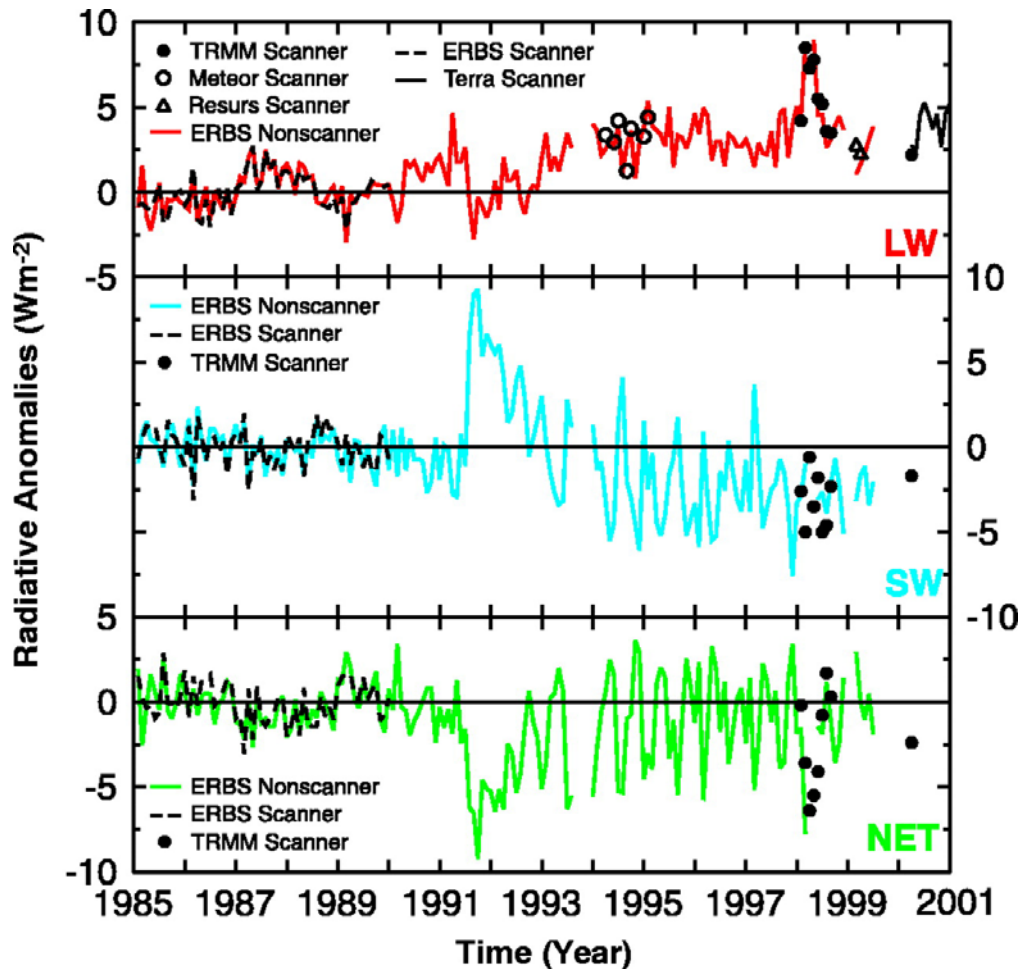
$$\frac{\partial F}{\partial I_{cloud}}$$

straightforward for specified cloud properties

$$\frac{dI_{cloud}}{dT_s}$$

very uncertain, many different cloud types

# Tropical Mean Radiation Flux (Satellite)



1985-1999 tropical mean time series of all-sky SW, LW, and net radiation flux from the Earth Radiation Budget Satellite (ERBS)

Created by  
B. Wielicki group

*from Wielicki et al. (2002)*

# Cloud-Temperature Regression

*Calculate regression of cloud on coincident temperature...*

$$\begin{aligned} \frac{\text{cov}(C,T)}{\text{var}(T)} = & \overset{\text{Term 1}}{\frac{(1+\beta_E)\beta_C}{1-\beta_C\beta_T}} + \overset{\text{Term 2}}{\frac{\beta_T}{1-\beta_C\beta_T} \frac{\text{cov}(C,M)}{\text{var}(T)}} \\ & + \underset{\text{Term 4}}{\frac{\beta_C}{1-\beta_C\beta_T} \frac{\text{cov}(T,N)}{\text{var}(T)}} + \underset{\text{Term 3}}{\frac{1}{1-\beta_C\beta_T} \frac{\text{cov}(M,N)}{\text{var}(T)}} \end{aligned}$$



# Cloud-Temperature Regression

$$\text{Term 1} \quad \frac{(1 + \beta_E)\beta_C}{1 - \beta_C\beta_T}$$

If  $\beta_E \neq 0$       temperature damping  
 $\beta_T = 0$       no cloud radiative forcing

$$\rightarrow |\text{Regression of } C \text{ on } T| < |\beta_C|$$

*Underestimation of cloud response factor magnitude*

*Underestimation of cloud feedback*

# Cloud-Temperature Regression

$$\text{Term 4} \quad \frac{\beta_C}{1 - \beta_C \beta_T} \frac{\text{cov}(T, N)}{\text{var}(T)}$$

If  $\text{cov}(T, N) \neq 0$   $T_{t+\Delta t}$  and  $T_t$  are autocorrelated through long  $N$  timescale

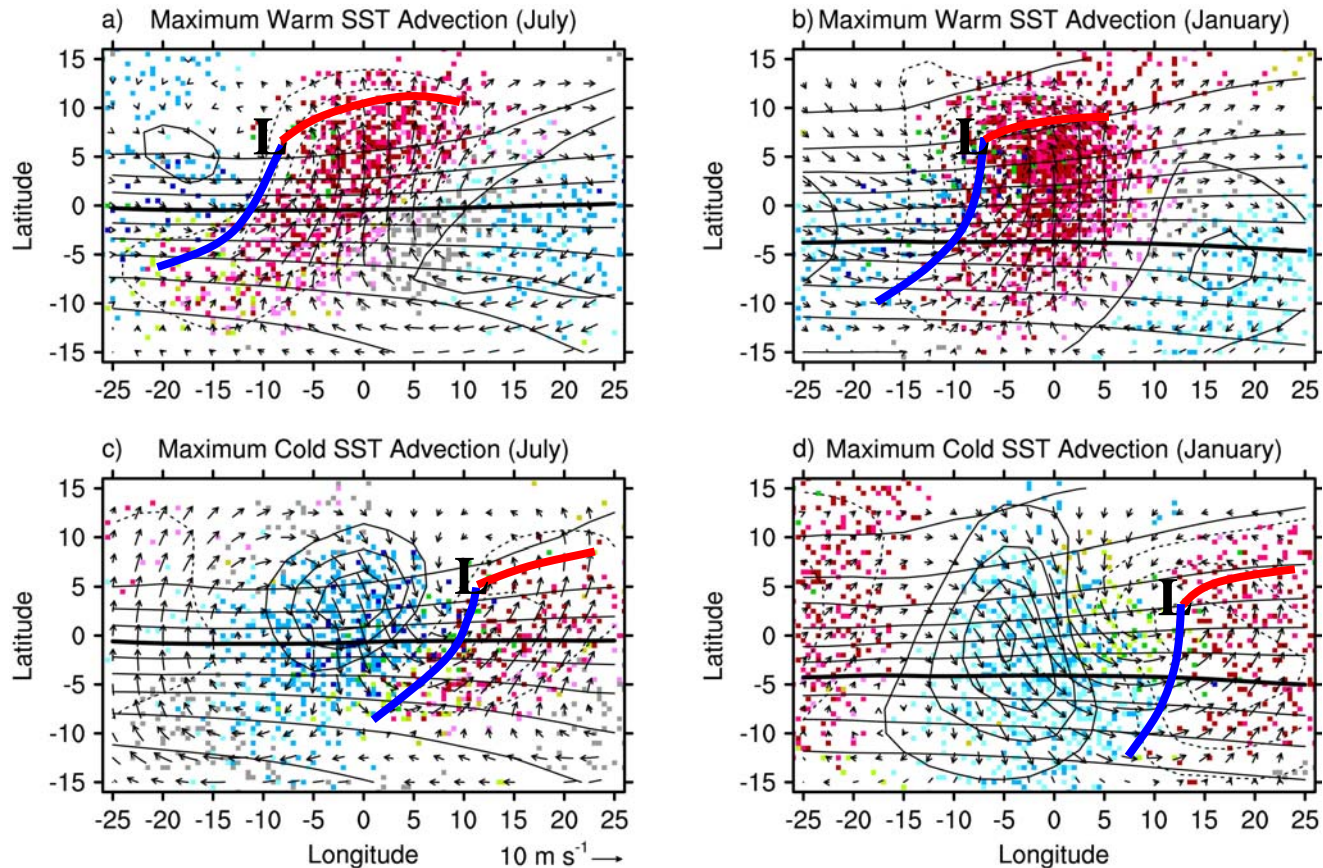
→ Coincident  $C$ - $T$  relationship because current temperature related to previous forcing of cloud

→ Regression of  $C$  on  $T$  has additional negative factor

*Cloud response factor appears more negative*

*Cloud feedback appears more positive*

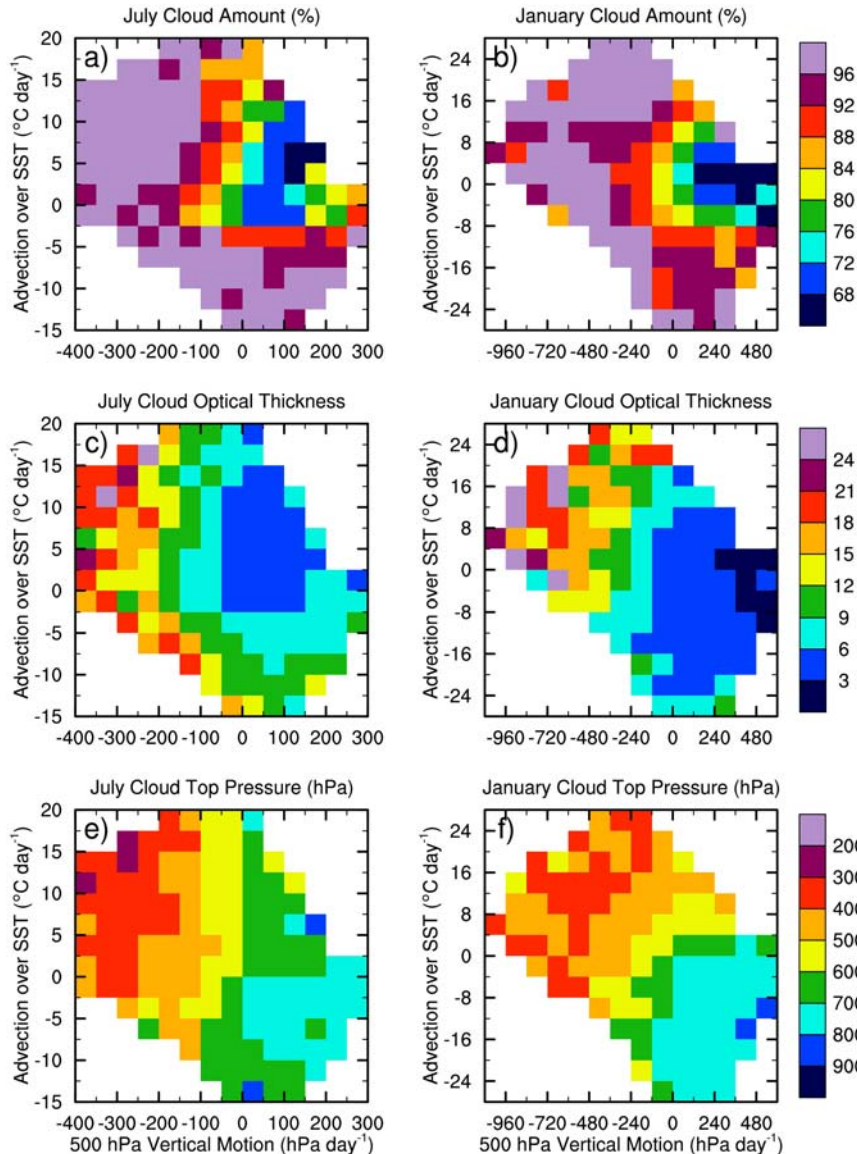
# Advection over SST Gradient



Thin High-Top    ■ ■ ■    Thick High-Top  
 Thin Mid-Top    ■ ■ ■    Thick Mid-Top  
 Thin Low-Top    ■ ■ ■    Thick Low-Top  
 ■ Clear Sky

*color: ISCCP cloud types*  
*black straight lines: SST*  
*black curved lines:  $\omega_{500}$*   
*black arrows:  $V_{1000}$*

# SST Advection- $\omega_{500}$ Histograms (Cloud)



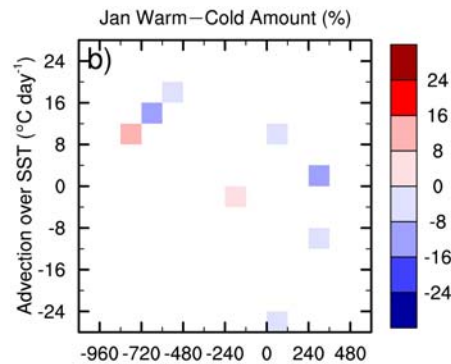
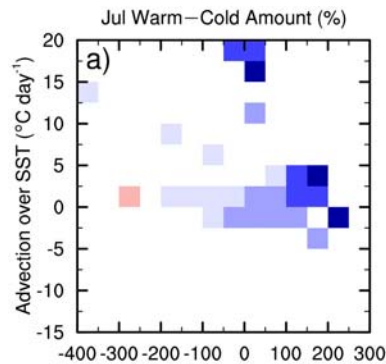
large cloud amount except  
for warm/down quadrant

largest optical thickness for  
upward and cold/down  
quadrants (latter only for  
July)

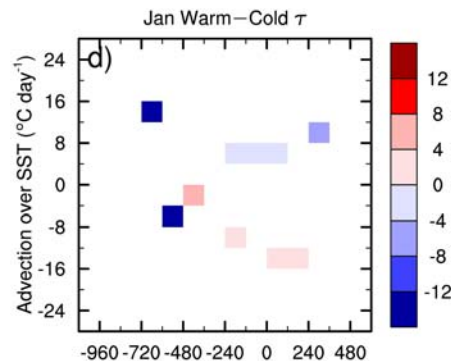
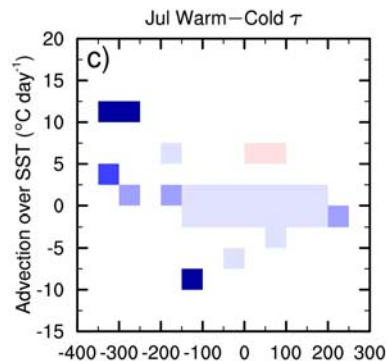
lowest cloud top pressure  
for upward motion

some cirrus clouds occur in  
warm/down quadrant

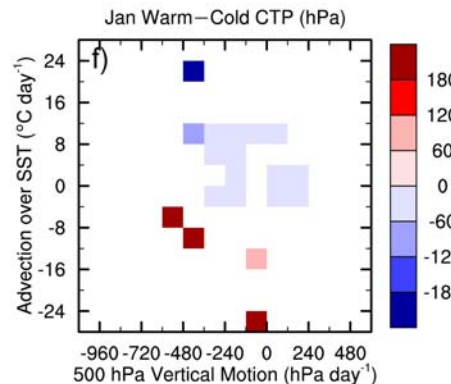
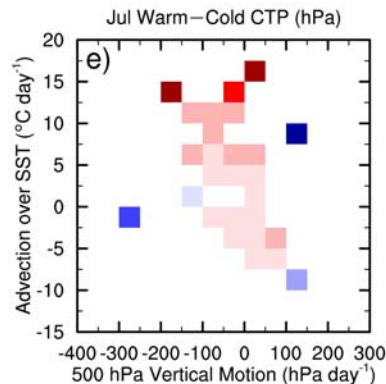
# Adv- $\omega$ Histograms (Warm–Cold Cloud)



less cloud amount for warm conditions under most dynamical states



less cloud optical thickness for warm conditions under most dynamical states



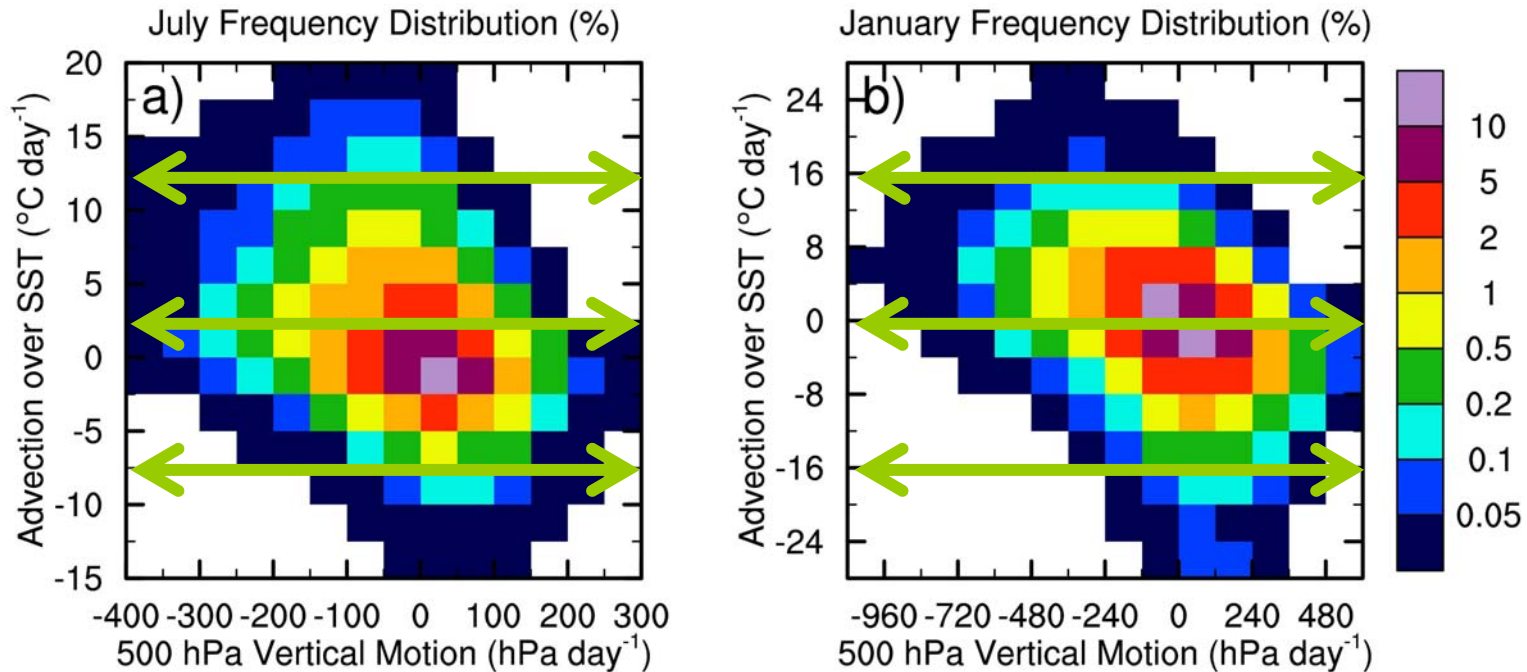
mixed cloud top pressure response for warming across seasons and dynamical states

## Average Warm–Cold Cloud

Month	Cloud Amount (% per K)	Optical Thickness (per K)	Cloud Top Pressure (hPa per K)
January	-0.1	0.0	-6
April	-1.1	-0.1	-4
July	-2.6	-0.3	+6
October	-1.6	0.0	-2



# Cloud Response to Dynamical Changes



Increase standard deviation of vertical motion and average cloud properties and CRF with new frequency distribution.

## 20% Decrease in $\omega_{500}$ Variability

Month	SW CRF (W m <sup>-2</sup> per K)	LW CRF	Net CRF
January	+0.7	-0.8	-0.1
July	+3.6	-1.2	+2.3



## 20% Decrease in SST Advection Variability

Month	SW CRF (W m <sup>-2</sup> per K)	LW CRF	Net CRF
January	+0.4	-0.3	-0.1
July	+1.2	-0.3	+1.0