

How close is close enough?

The role of bulk surface fluxes in regulating tropical clouds and circulations in models

Charlotte A. DeMott
Colorado State University

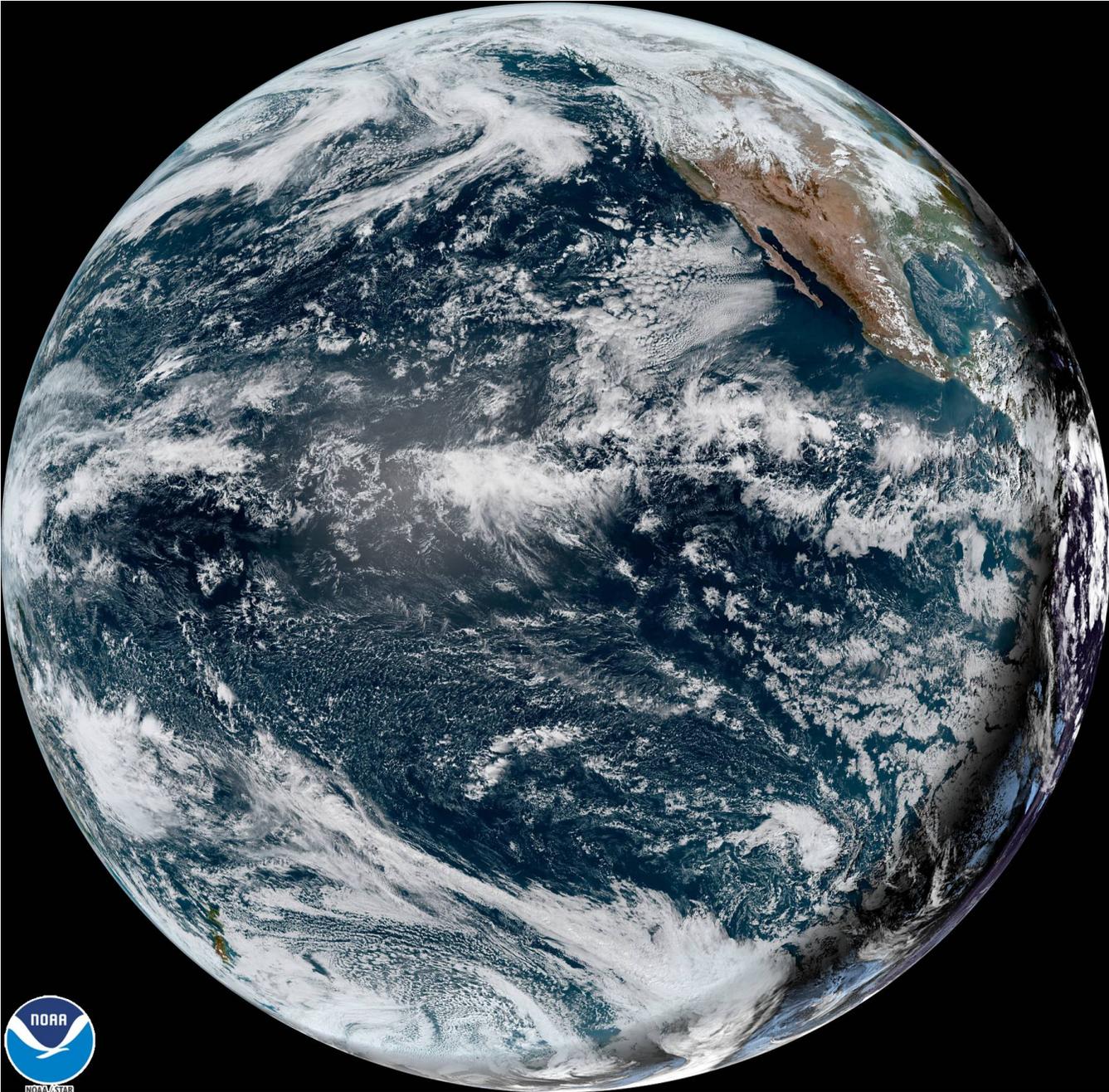
Xuanyu Chen (NOAA/CIRES)
Juliana Dias (NOAA PSL)
Robert Pincus (Columbia)
Brandon Wolding (NOAA PSL)

Carol Anne Clayson (WHOI)
Steve Woolnough (U. Reading)
Mark Branson (CSU)
Chia-Wei Hsu (NOAA PSL)

with additional input from Xubin Zeng, Jack Reeves Eyre, Kyle Shackelford, Elizabeth Thompson, Jeremiah Brown



the atmosphere and ocean from above

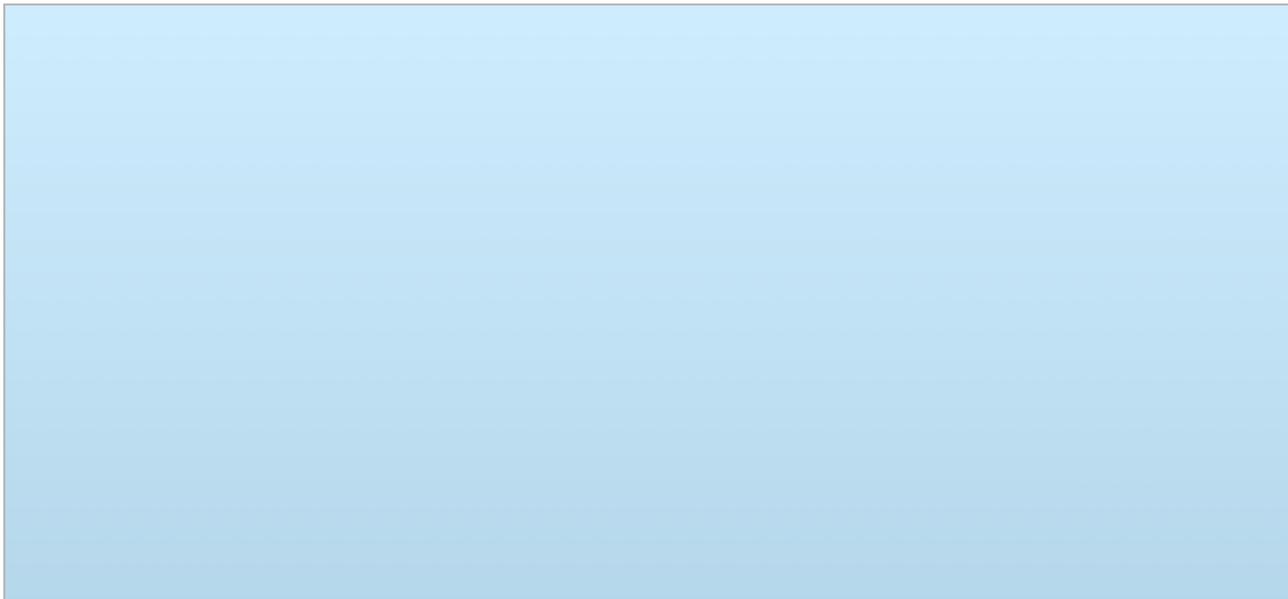
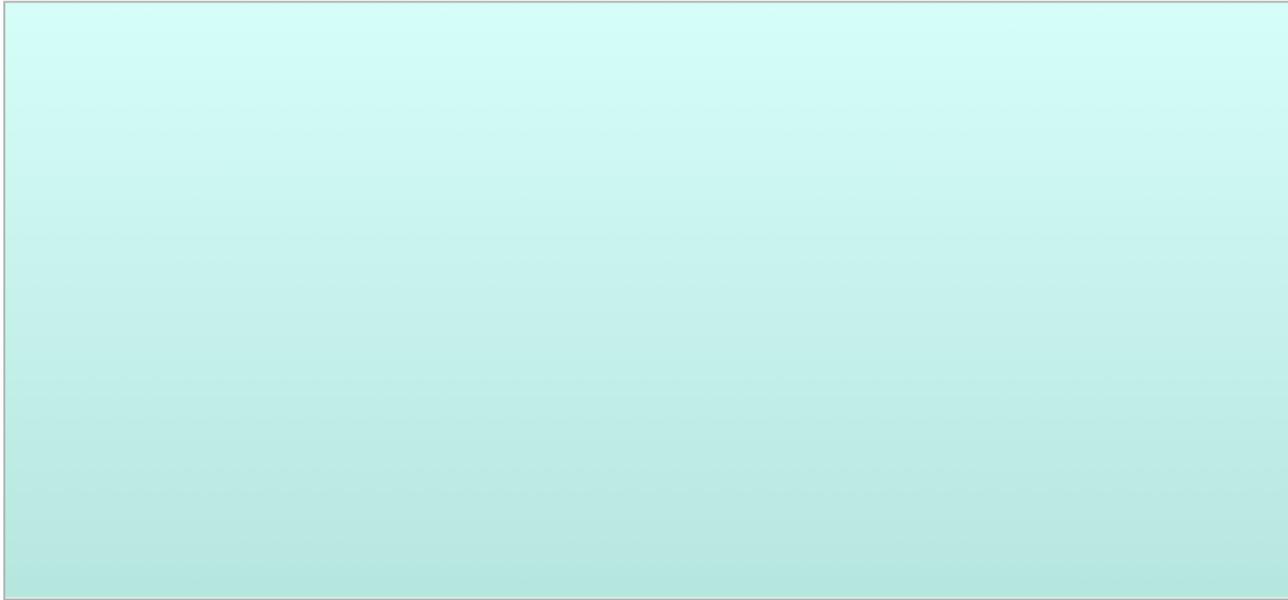


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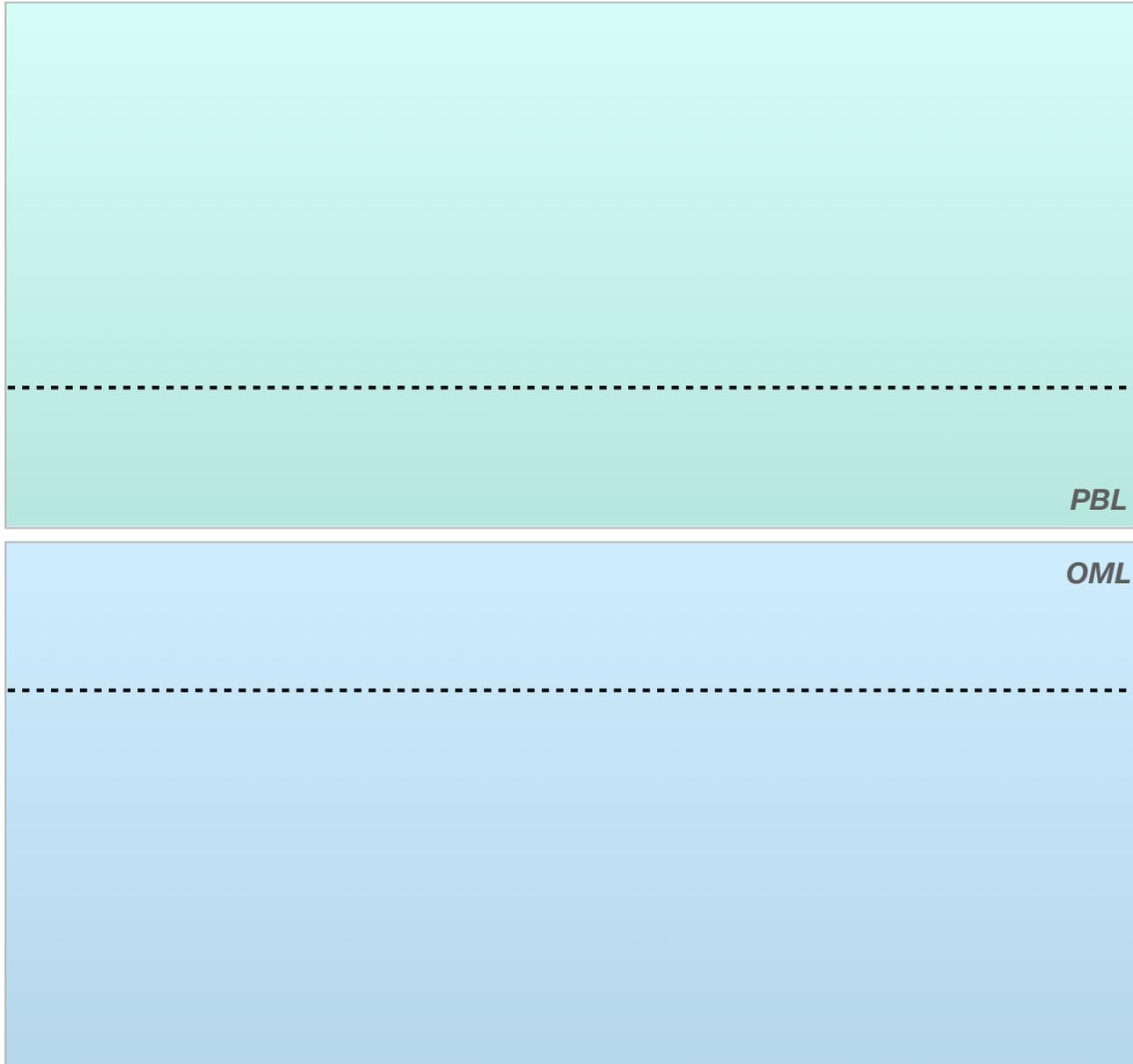
- SST variability can affect cloudiness
- fluxes tell the atmosphere about SST
- flux parameterizations can be biased



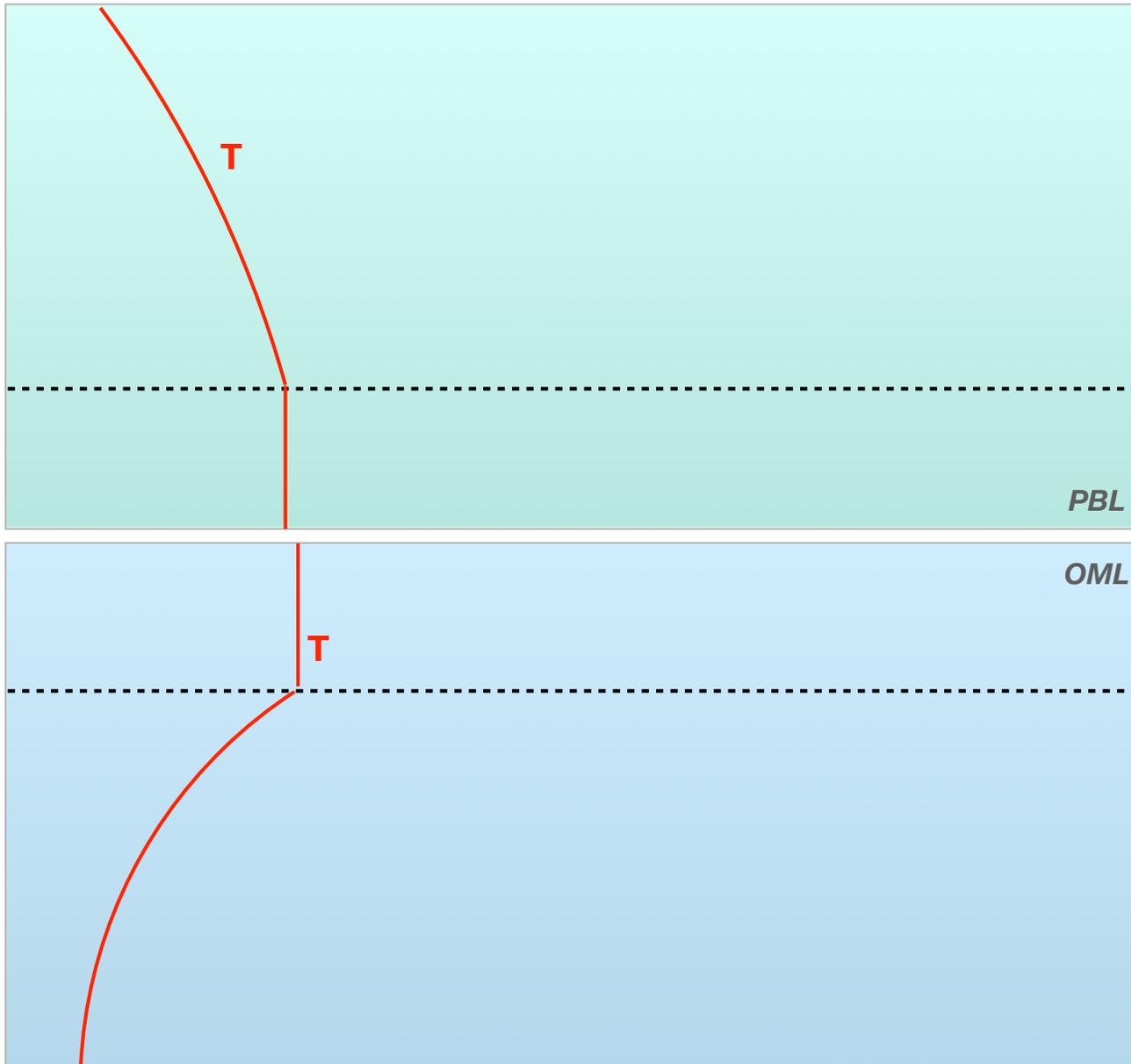
ocean-atmosphere symmetries (similarities)



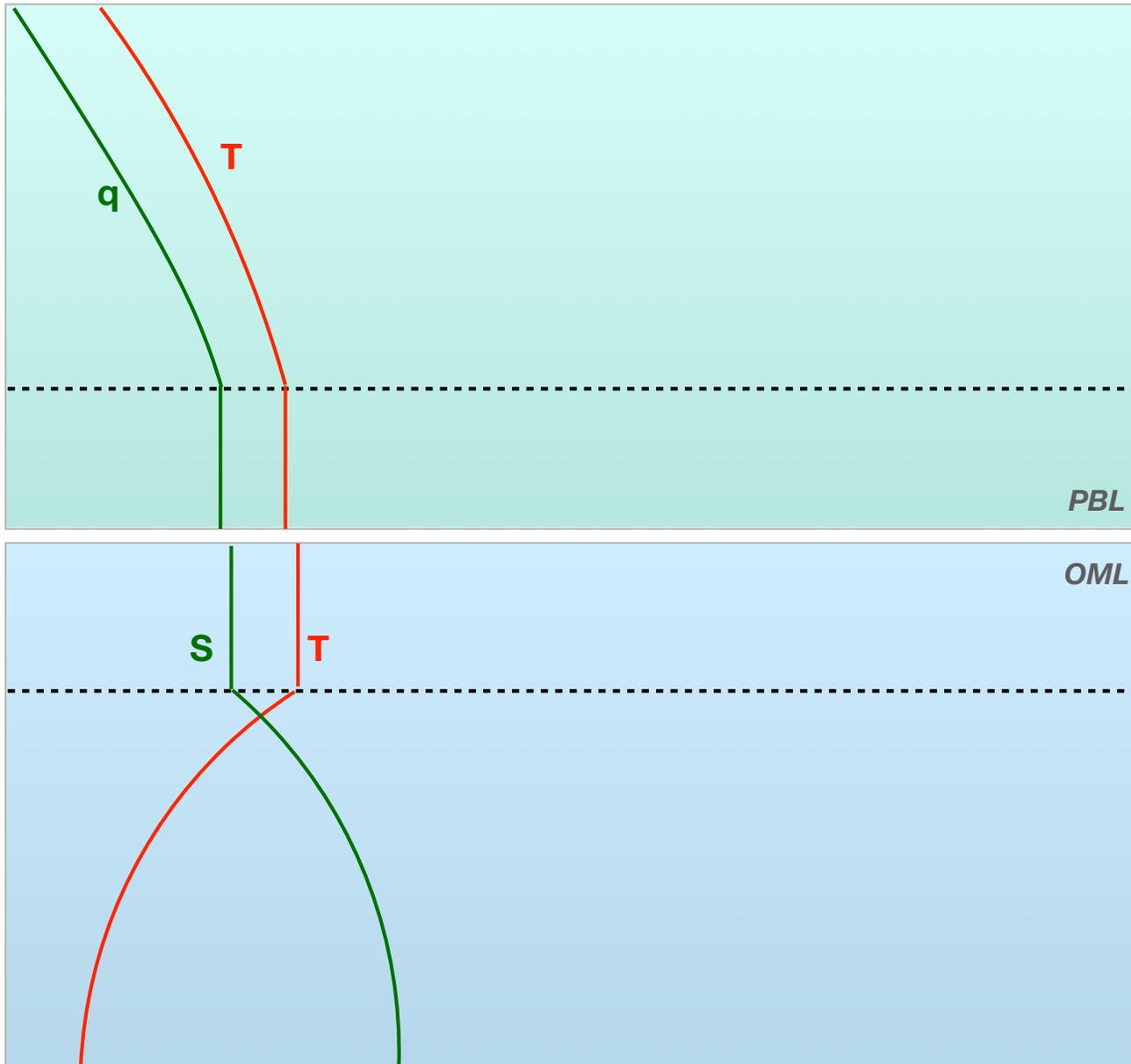
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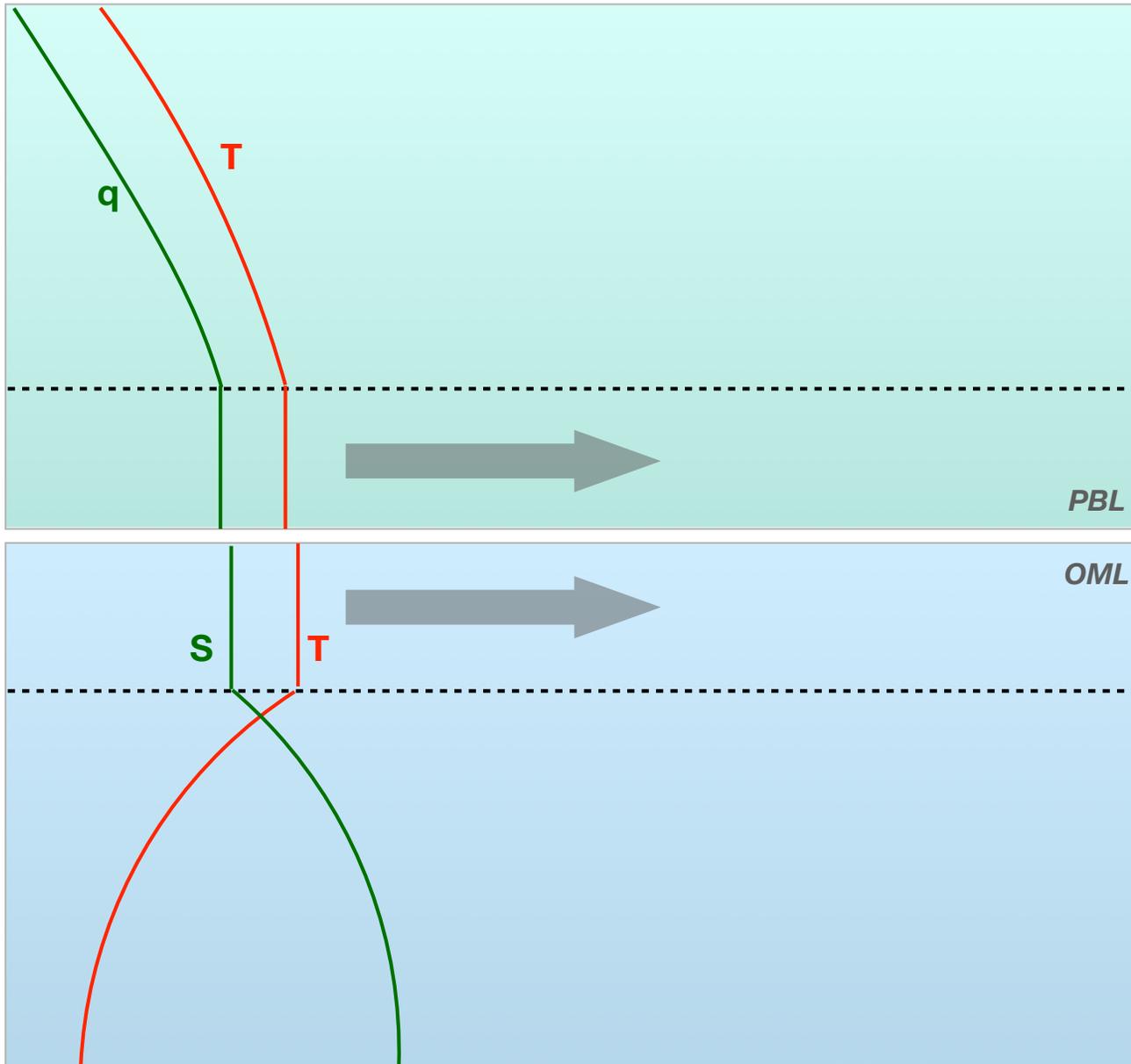
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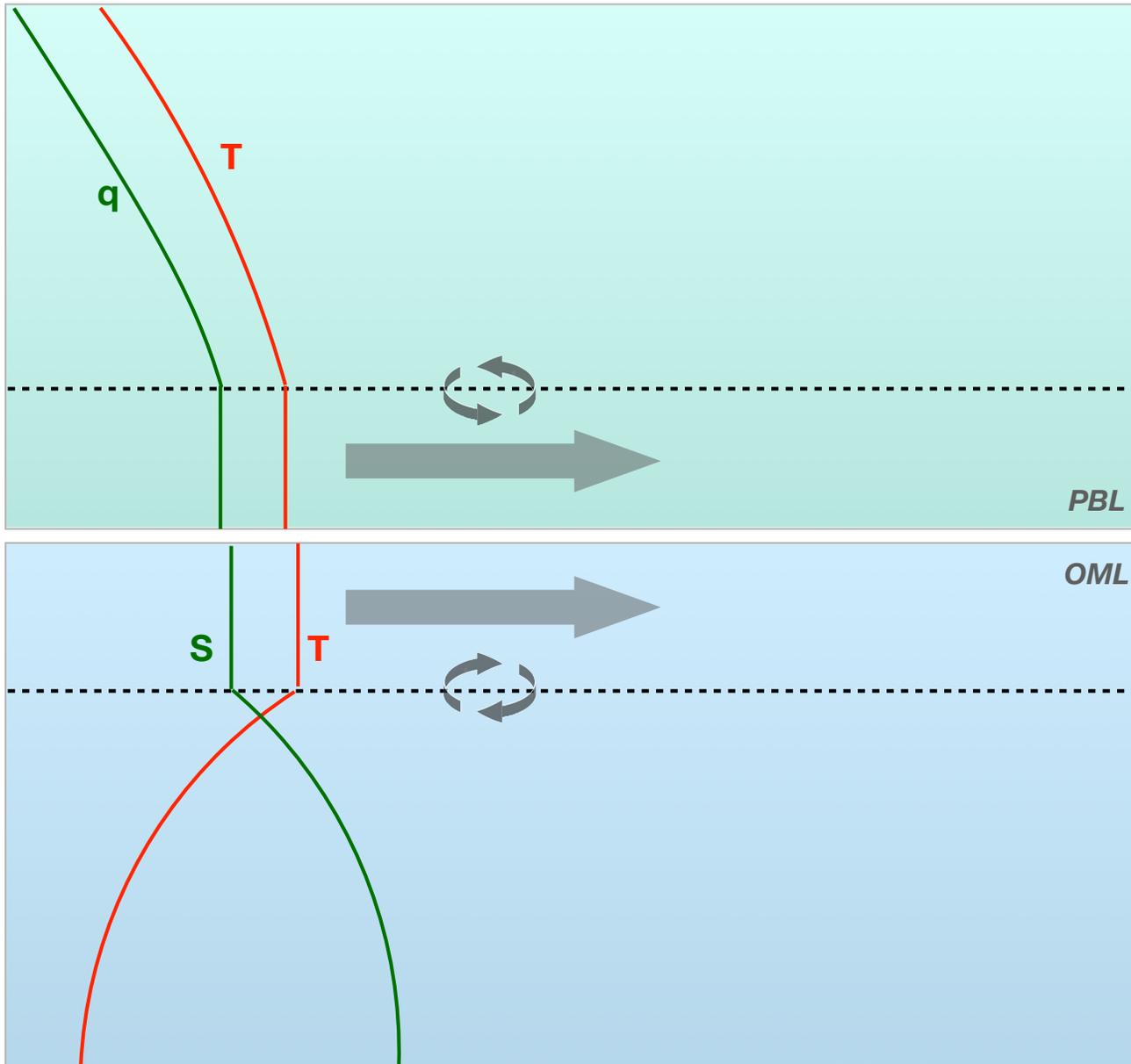
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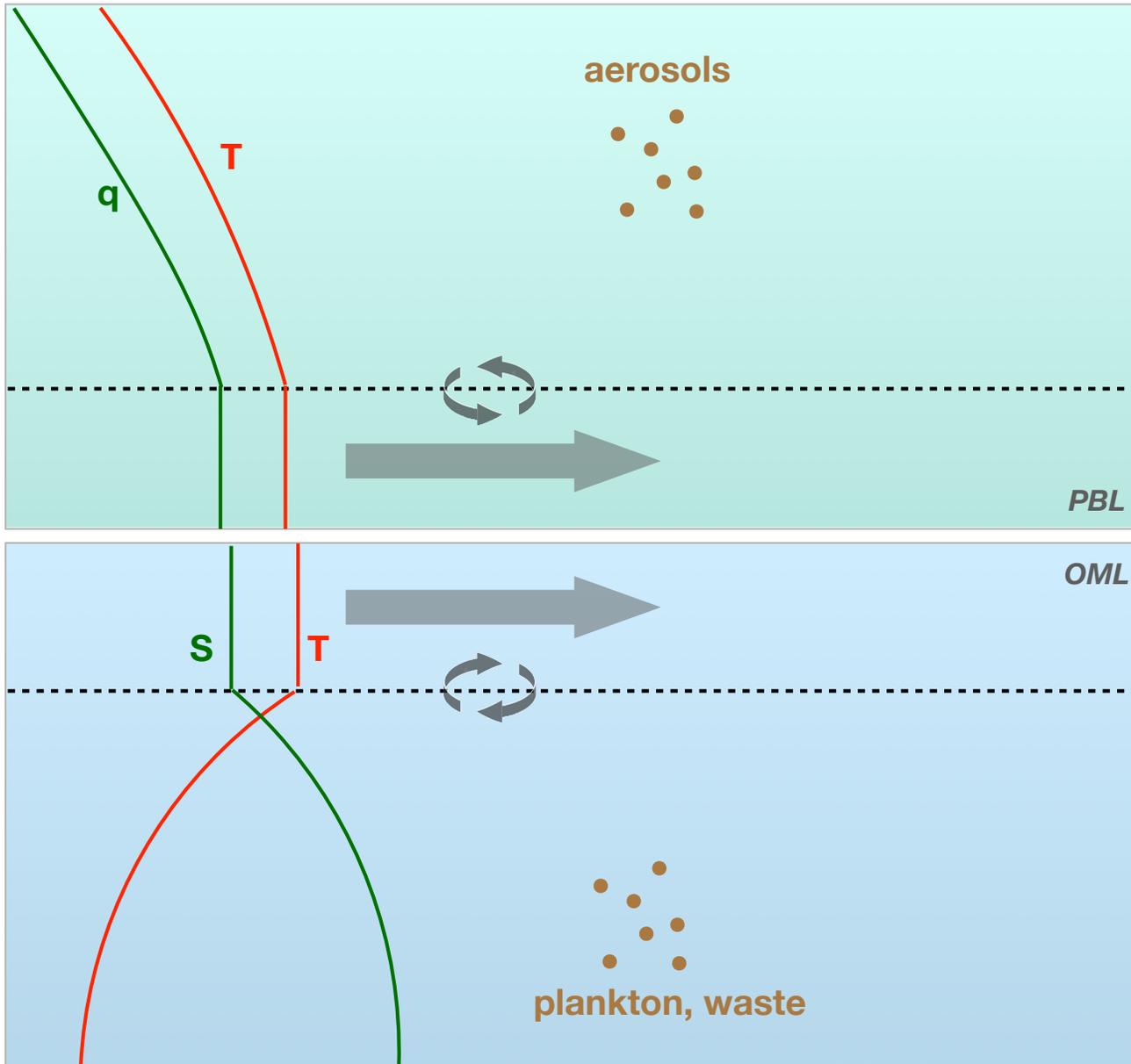
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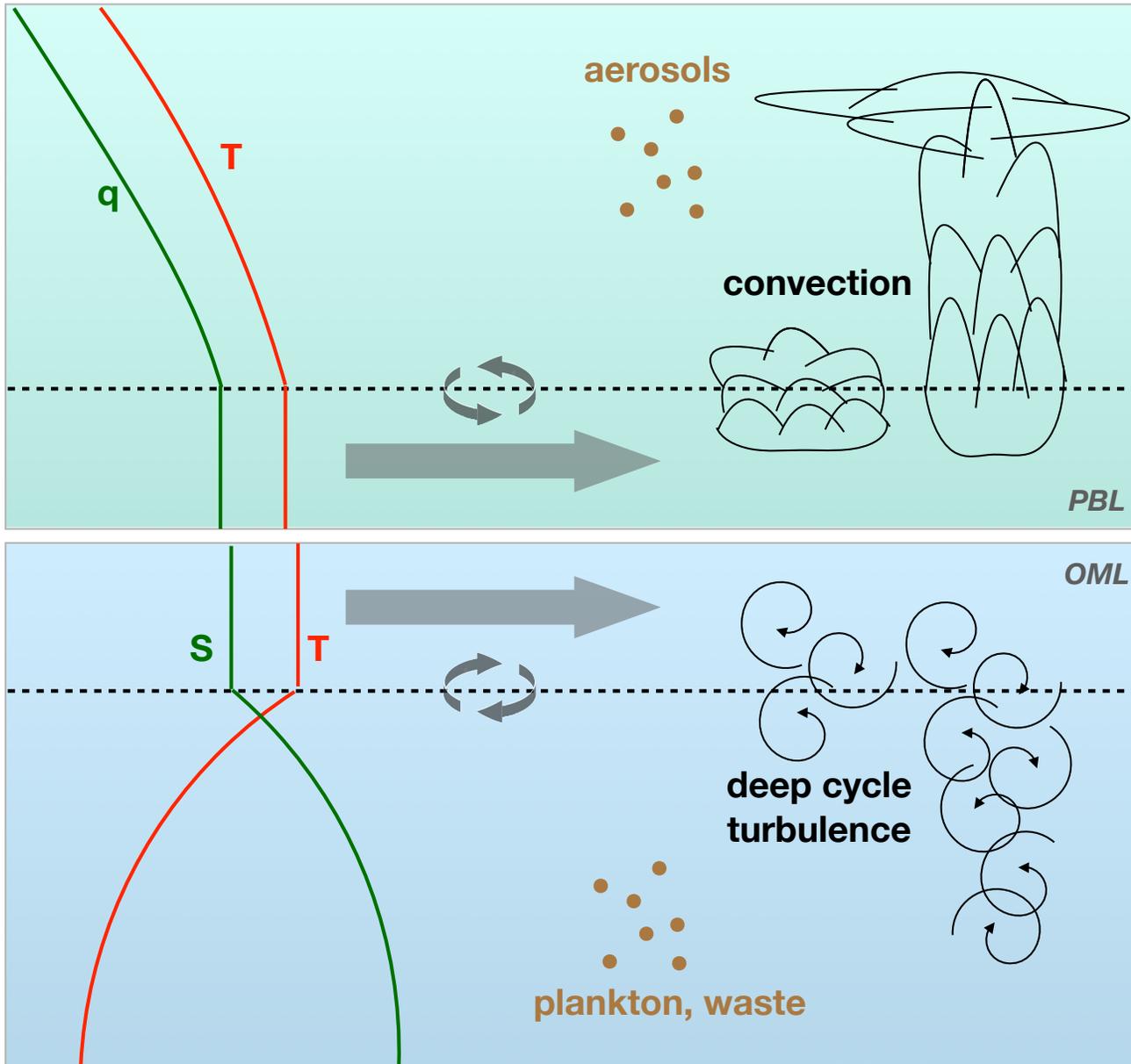
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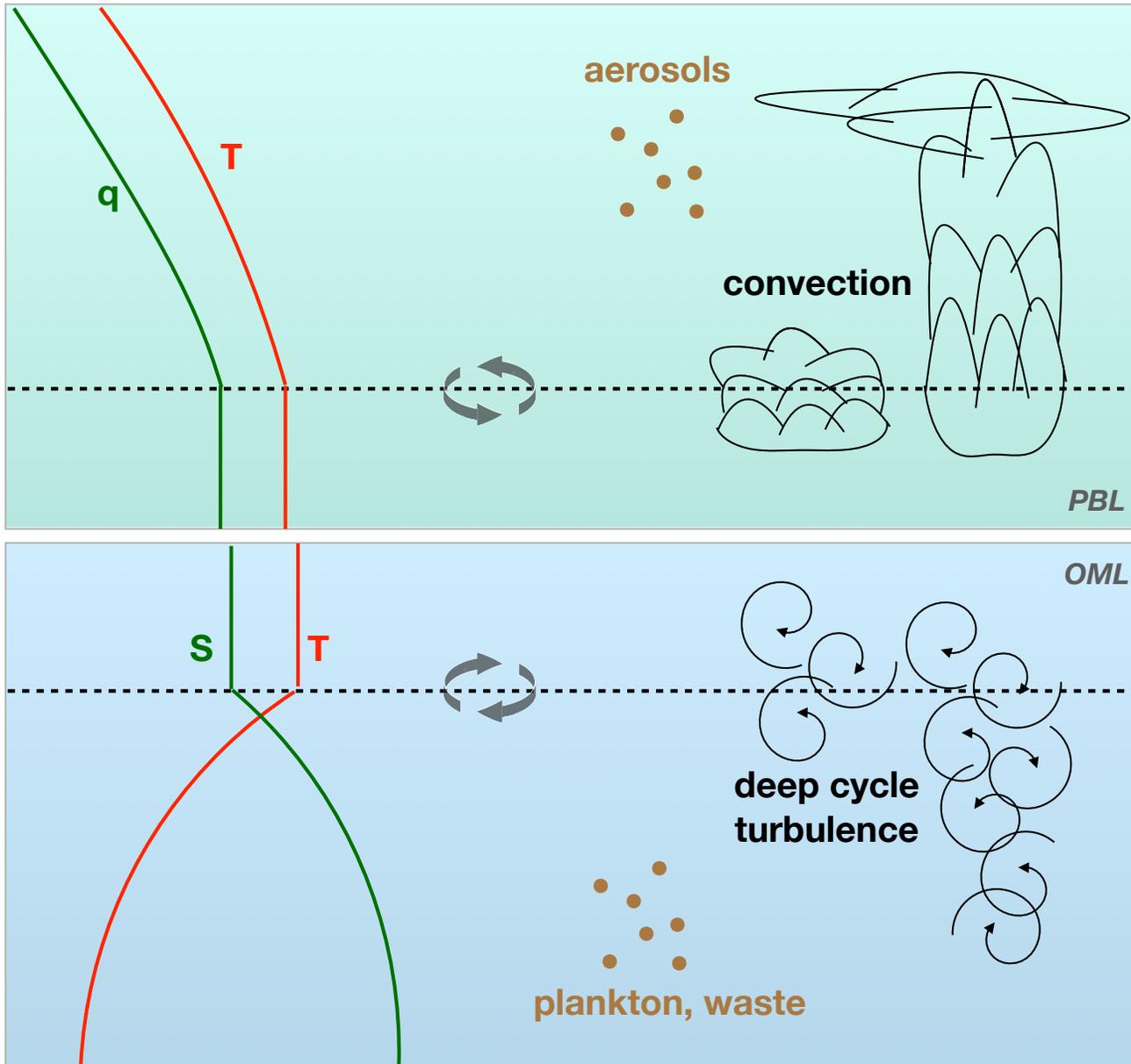
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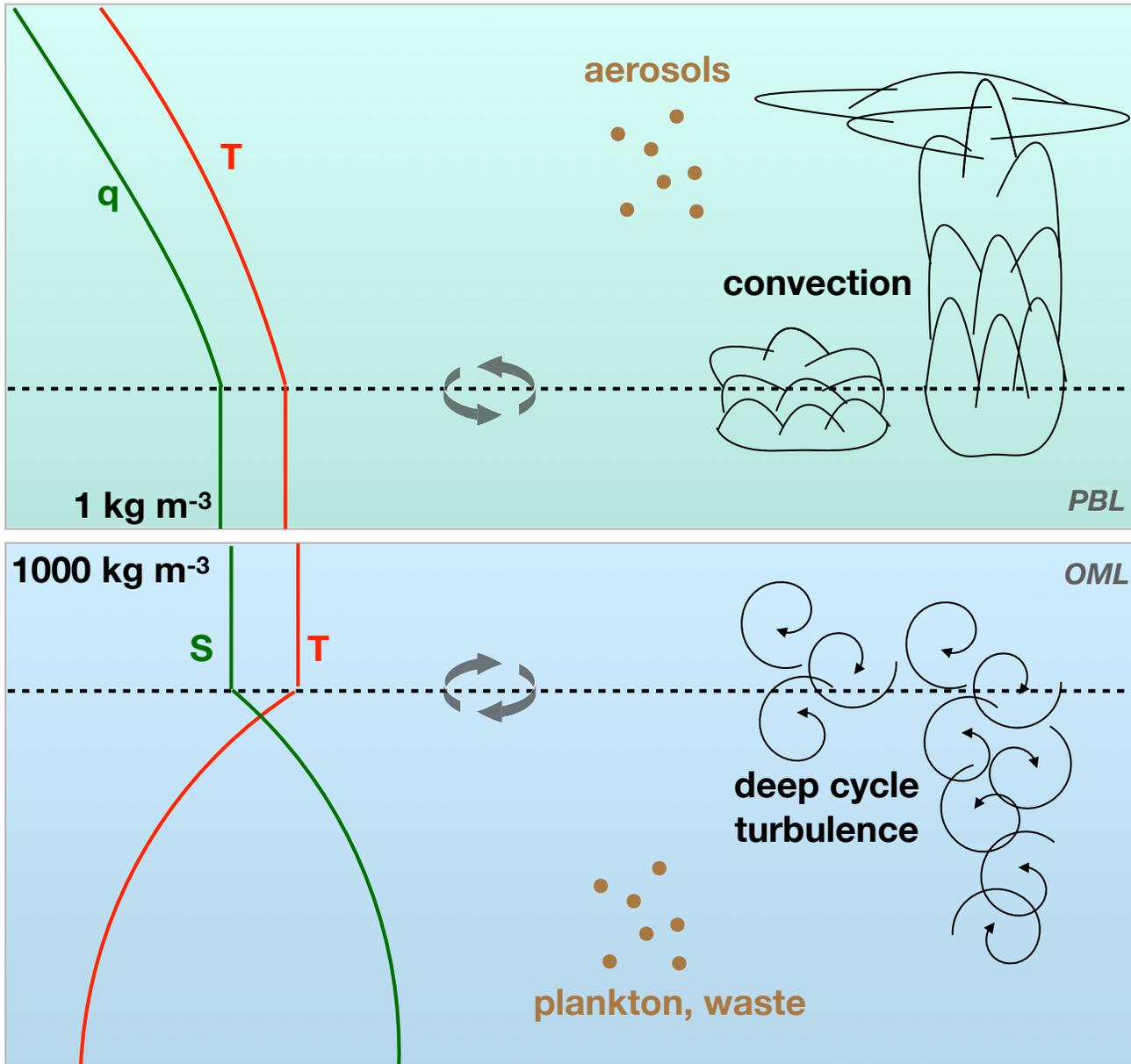
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ocean-atmosphere asymmetries (differences)

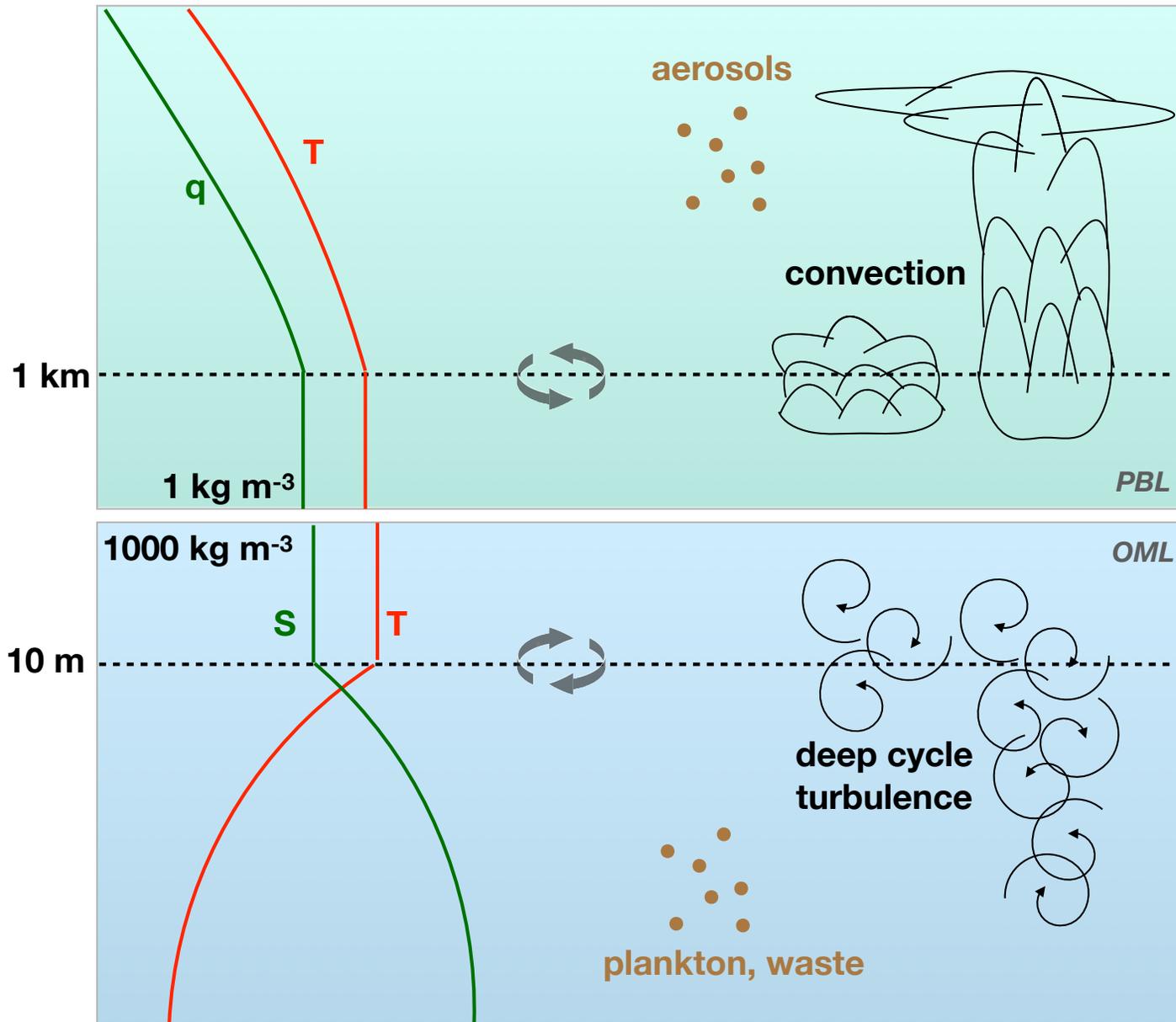


ocean-atmosphere asymmetries (differences)



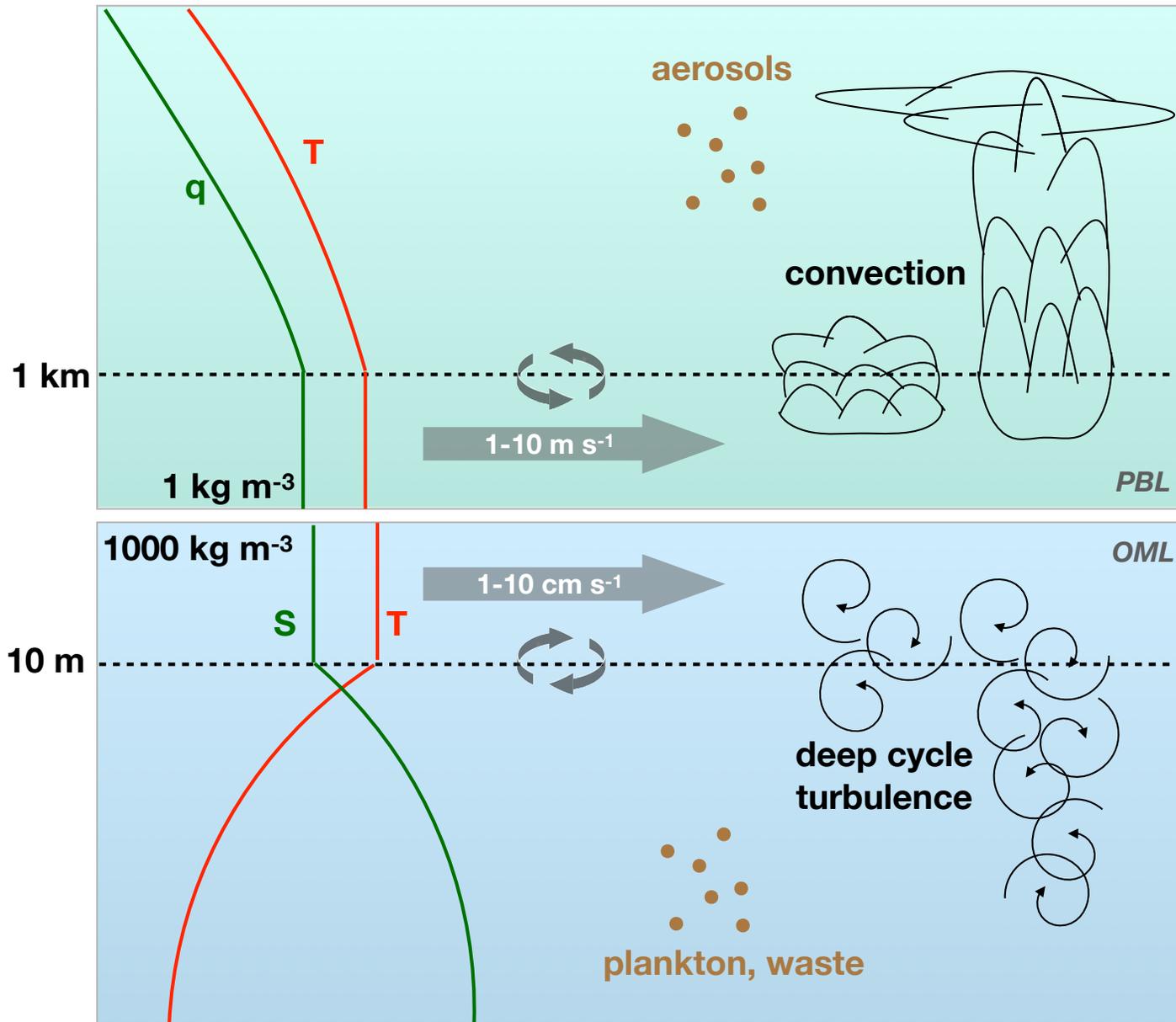
- density: $\delta \sim 10^3$

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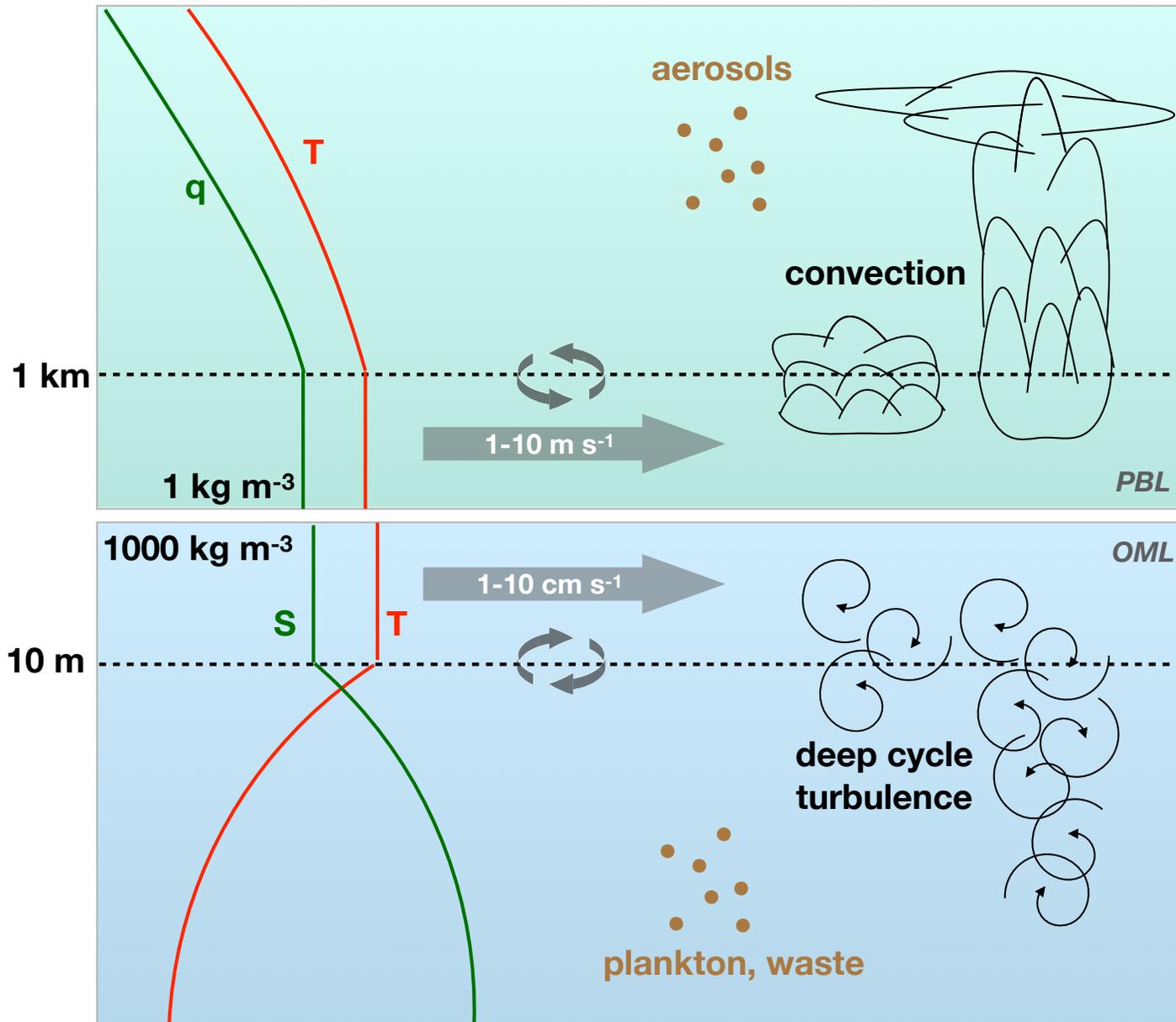
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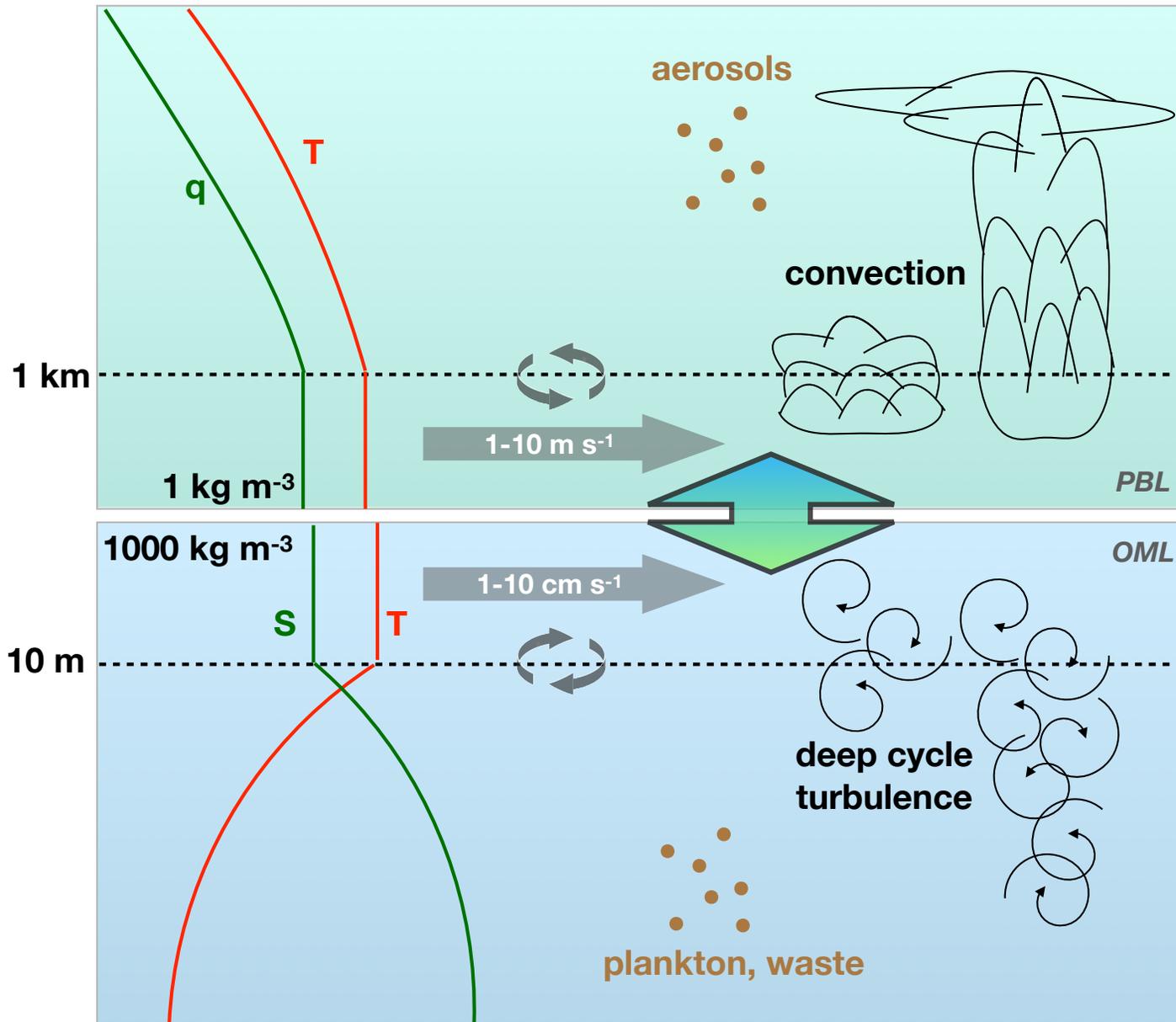
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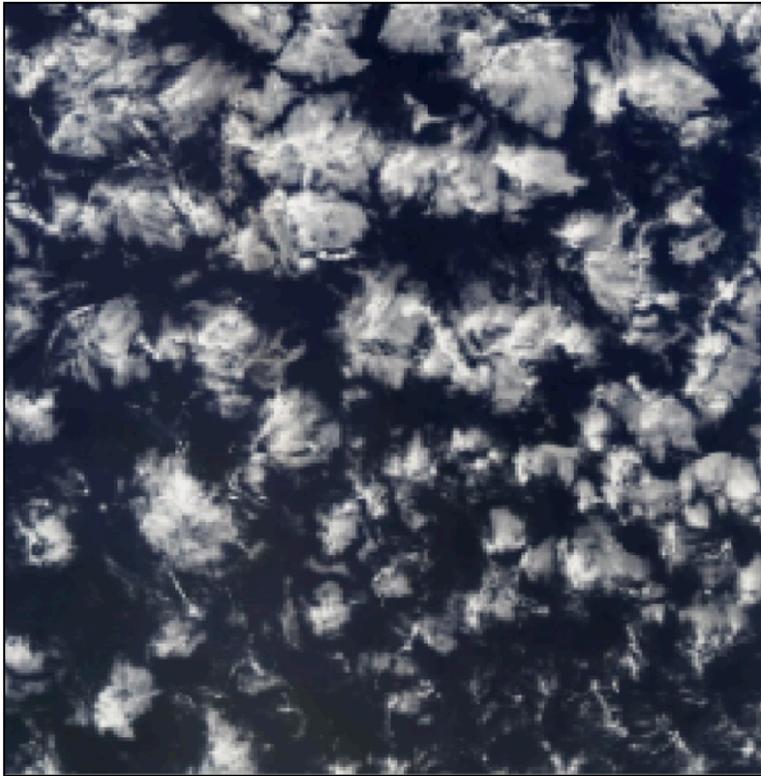
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- clouds are global; DCT mostly on Equator
- clouds regulate radiation; DCT does not

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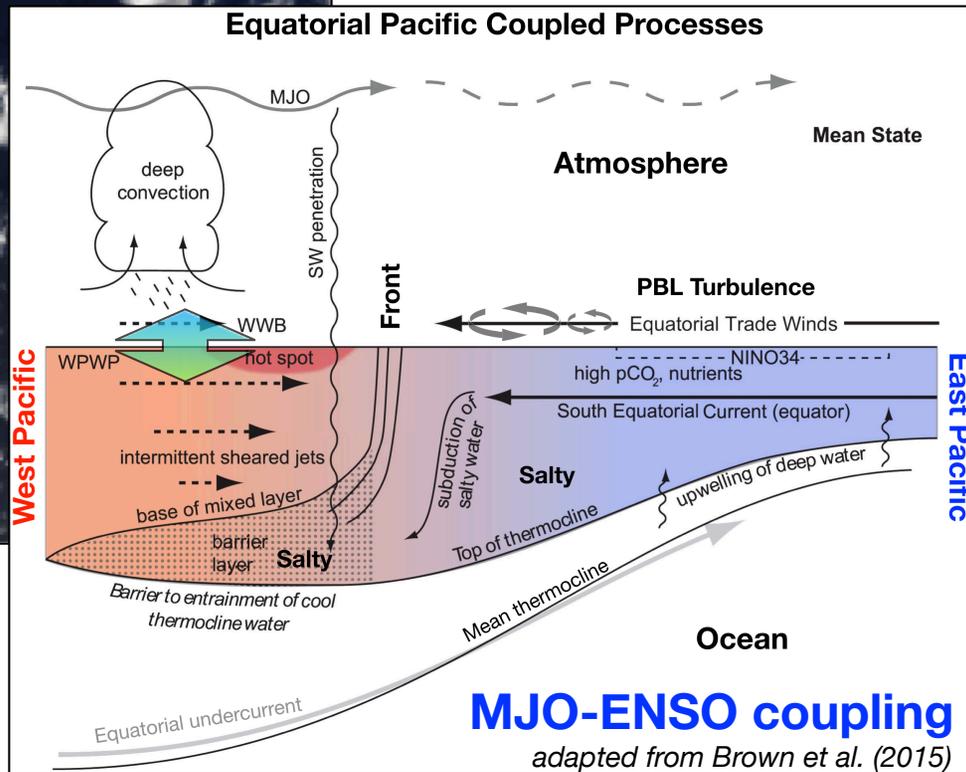
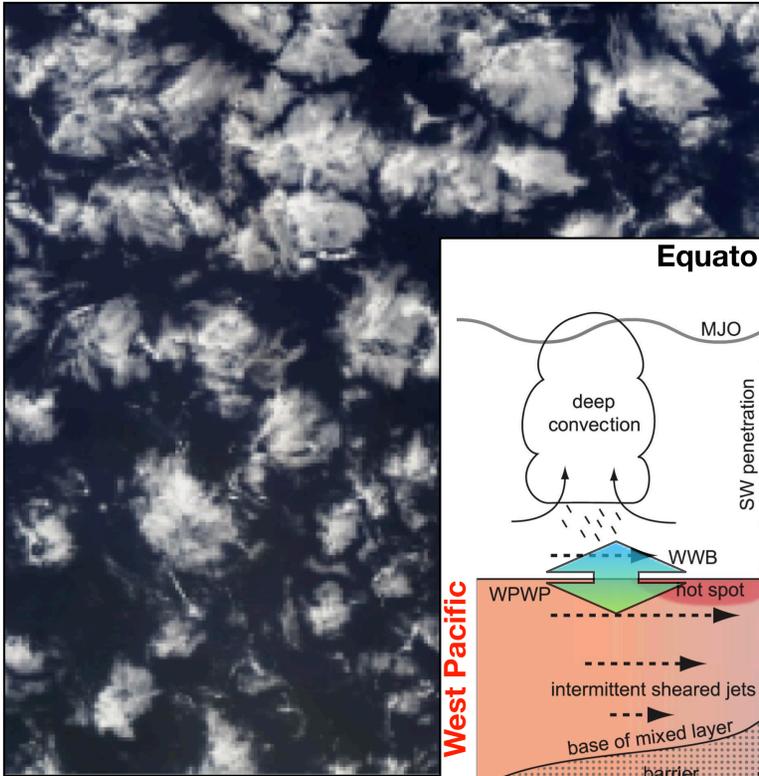


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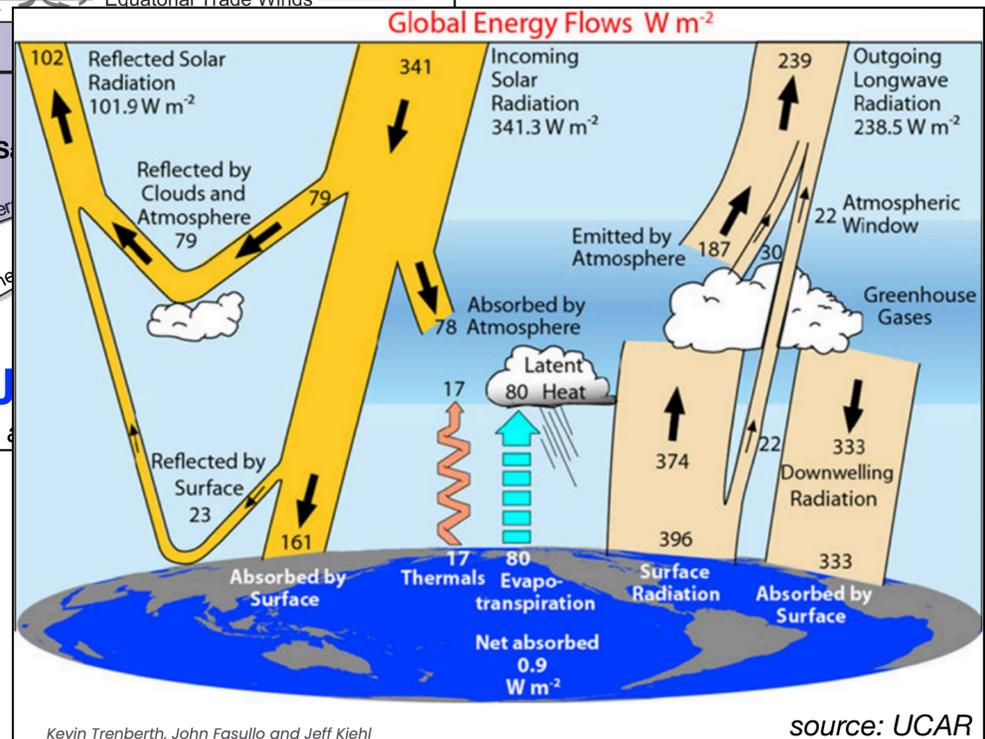
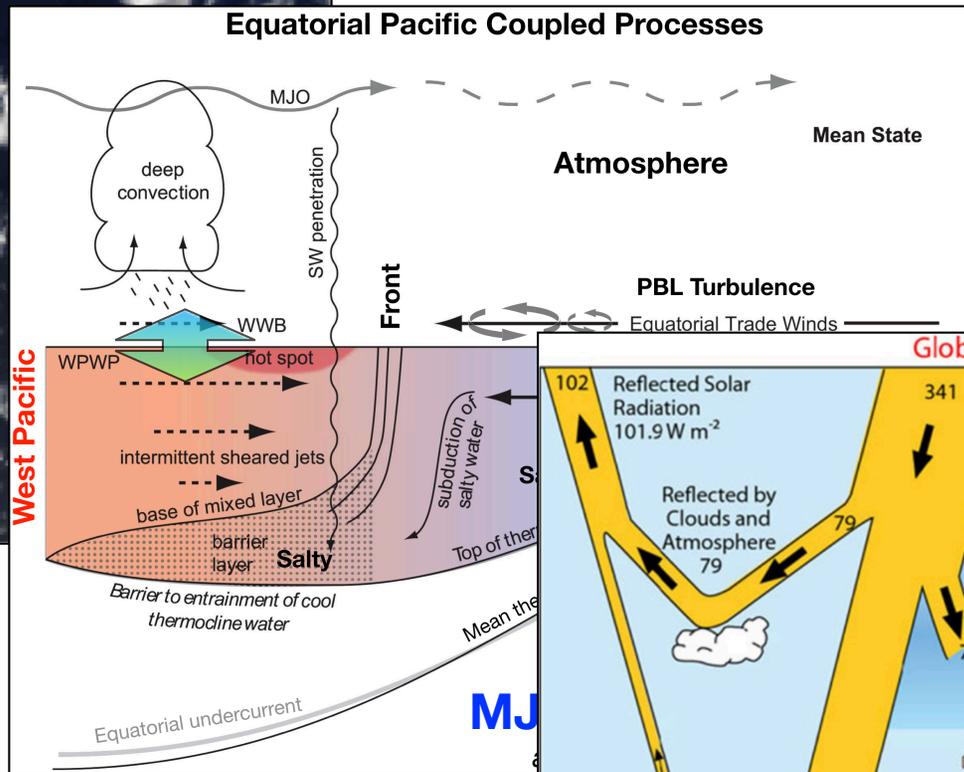
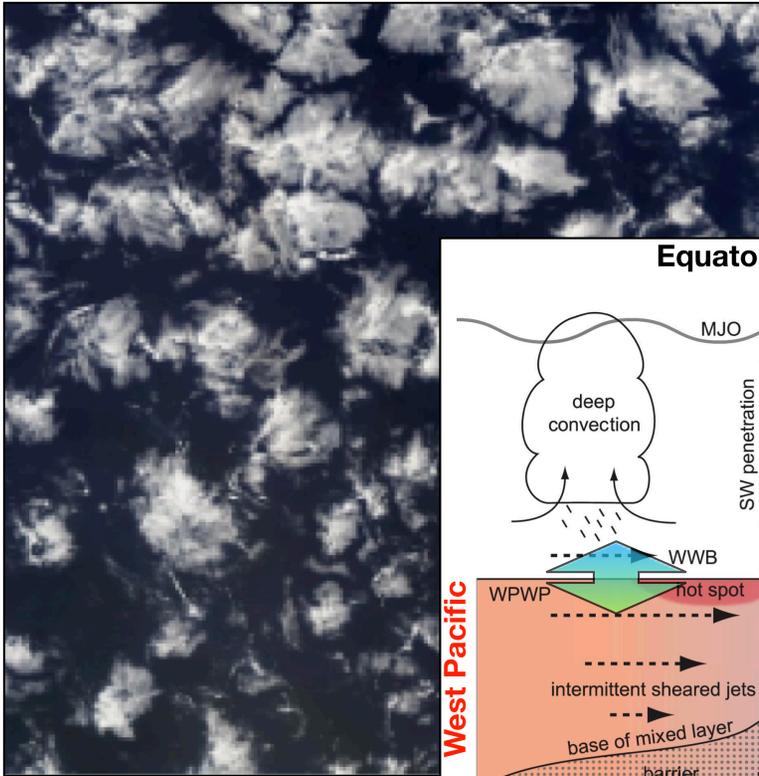
marine surface fluxes across scales



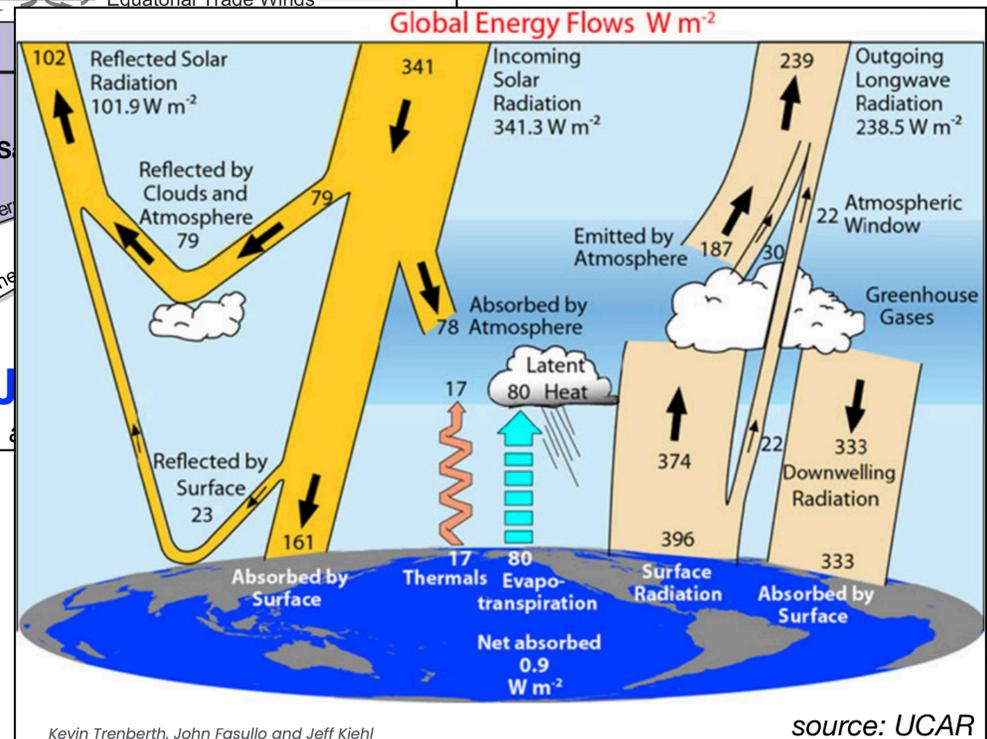
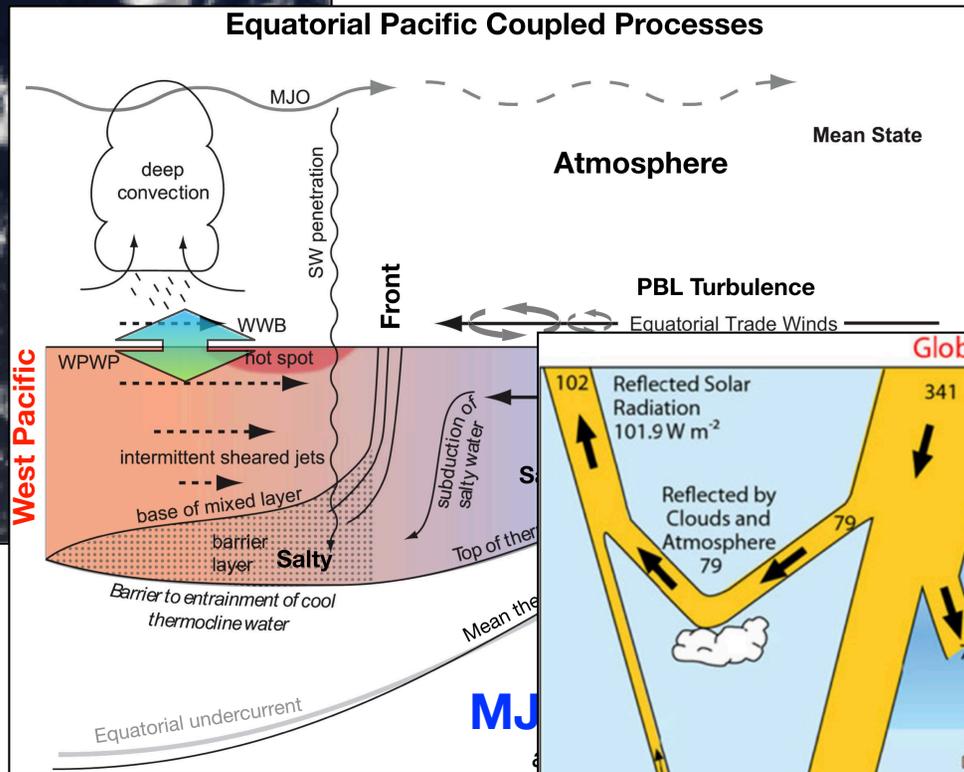
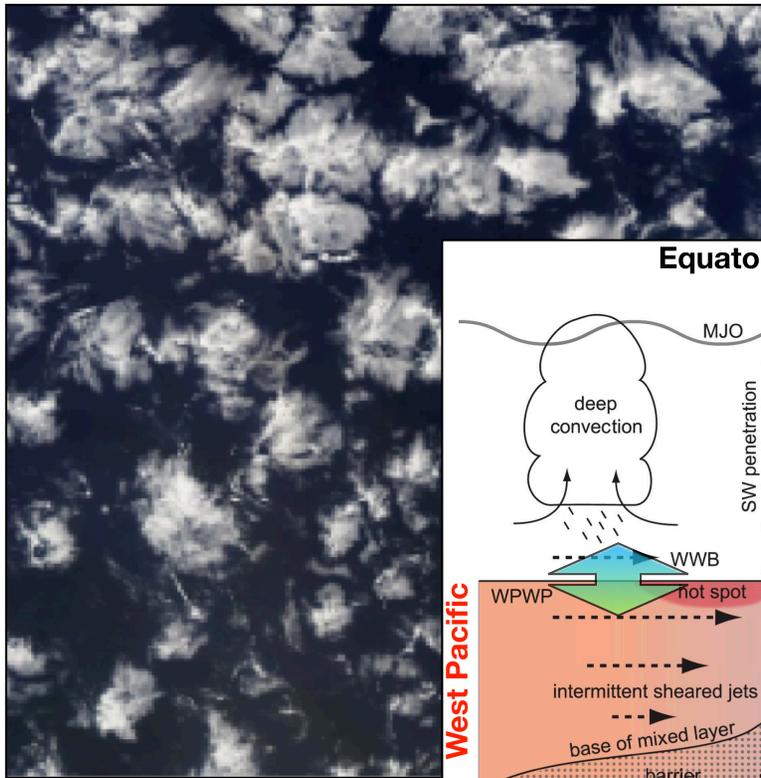
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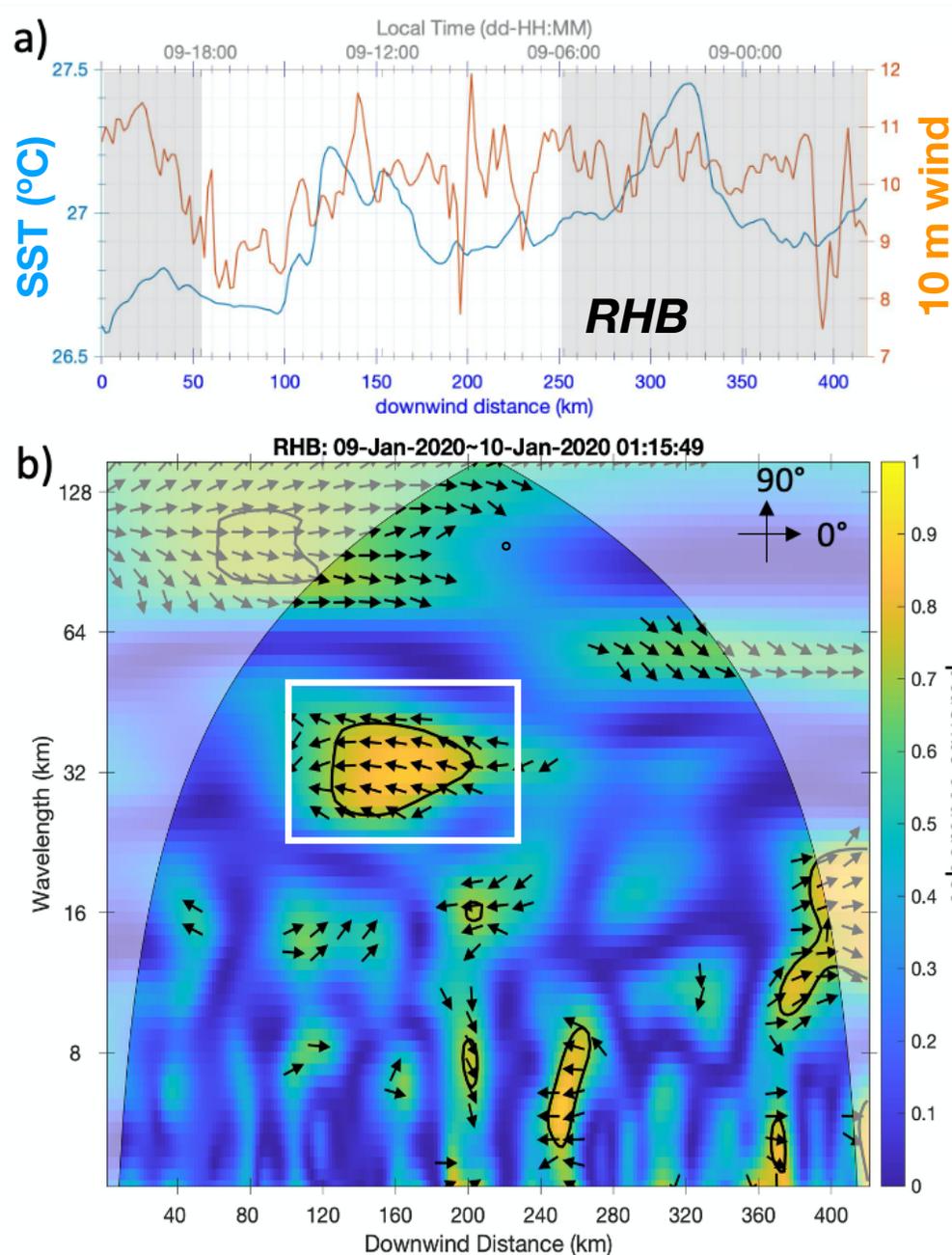
small/fast

large/slow

SST effects during EUREC4A/ATOMIC



Xuanyu Chen (NOAA/CIRES)

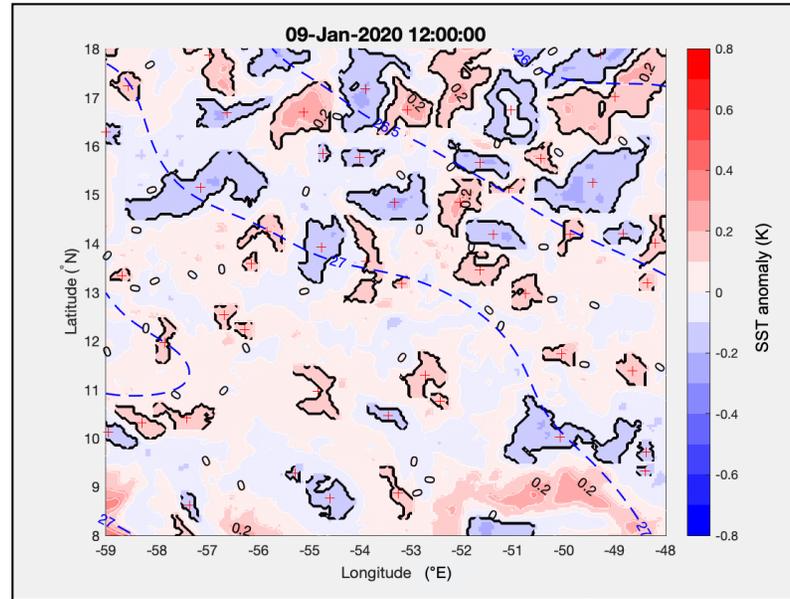


- analyzed SST, winds from RHB, wave gliders
- SST, wind vary coherently on 14- and 28-km scales

SST effects during EUREC4A/ATOMIC



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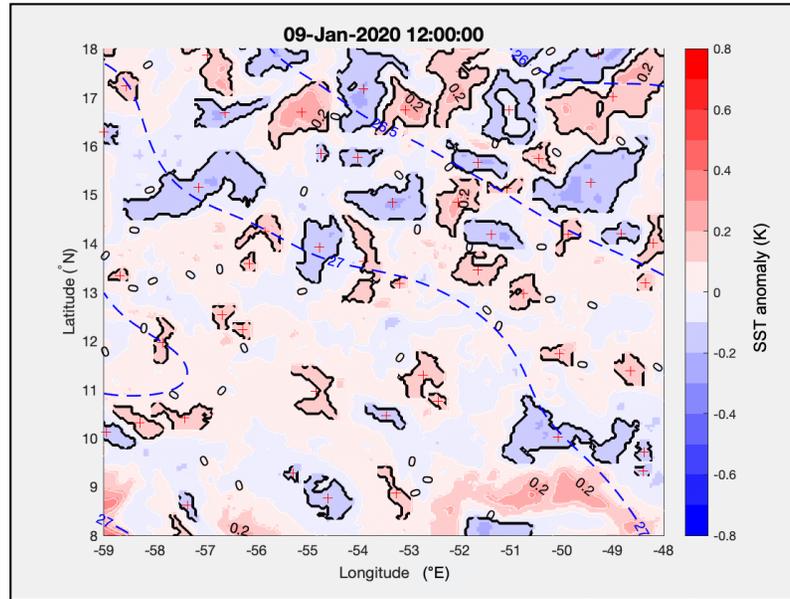


- GHRSSST high-pass filtered SST anomalies ($\sim 10^\circ \times 10^\circ$ lat-lon box)
- $\pm 0.2\text{K}$ warm/cold SST patches 40~140 km dia.

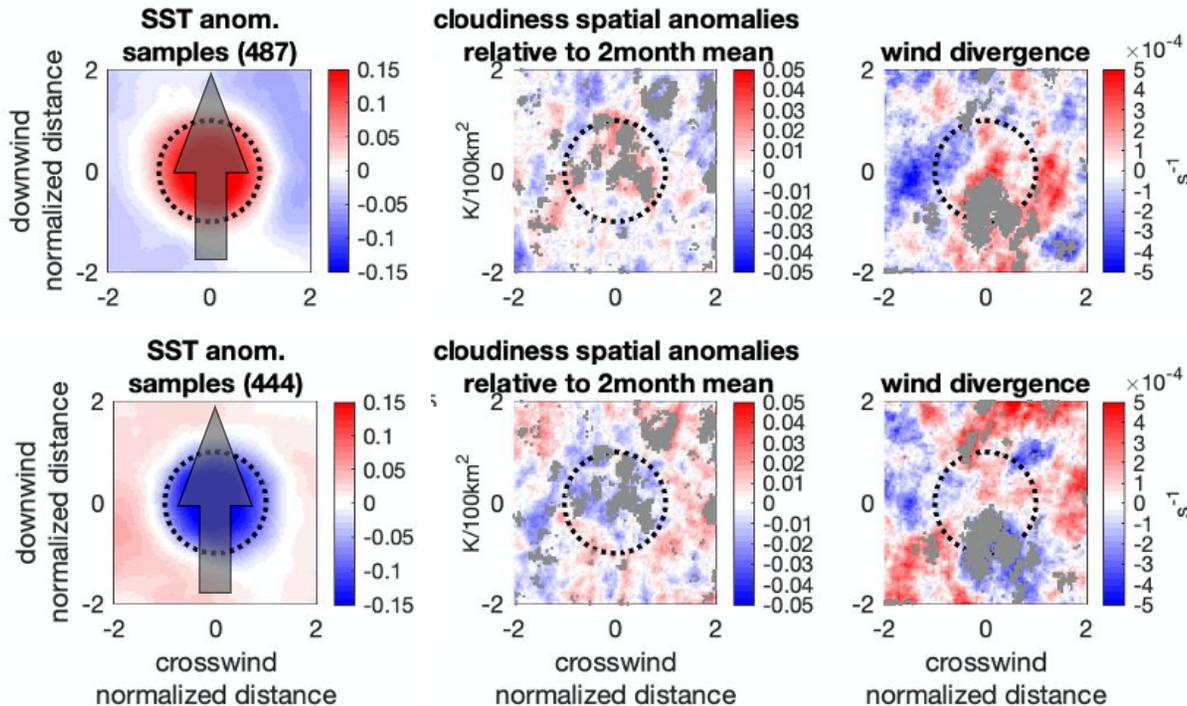
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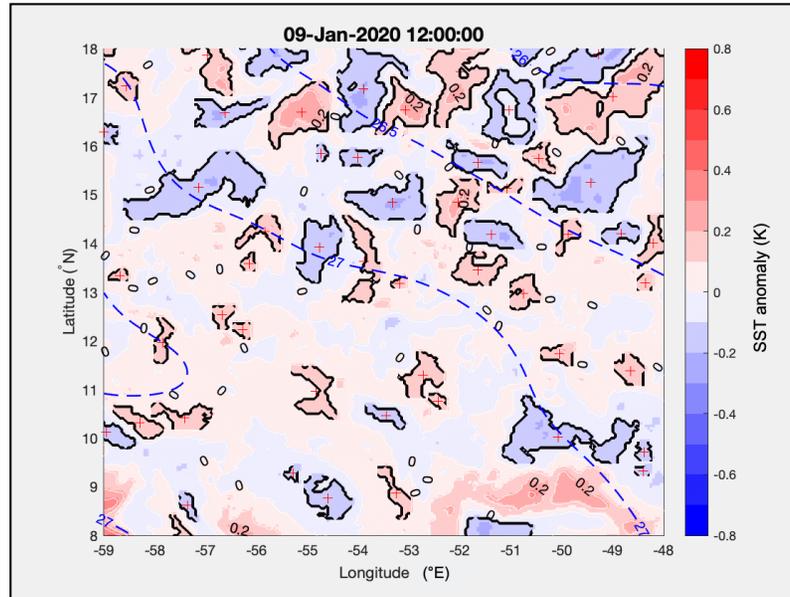


- small, weak SST anomalies regulate surface divergence, cloudiness

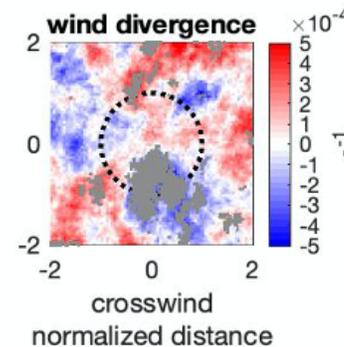
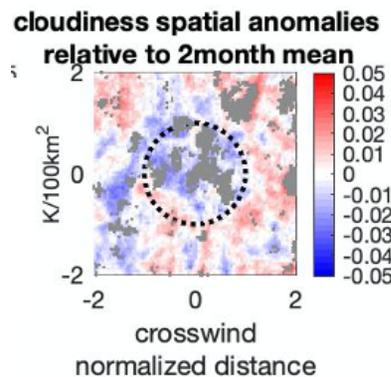
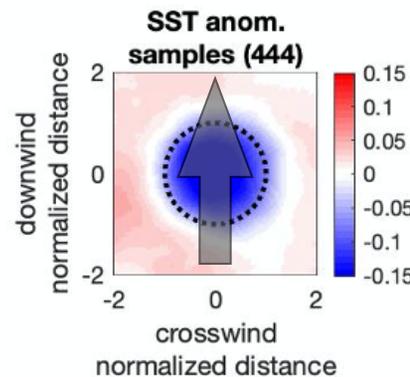
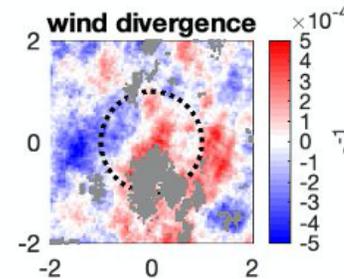
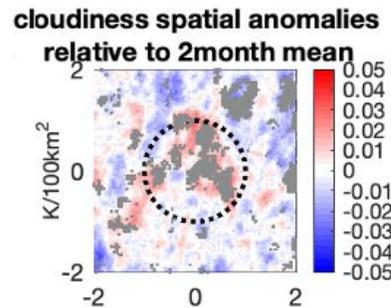
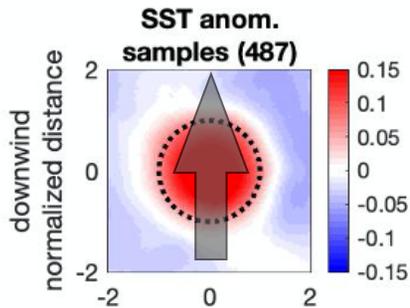
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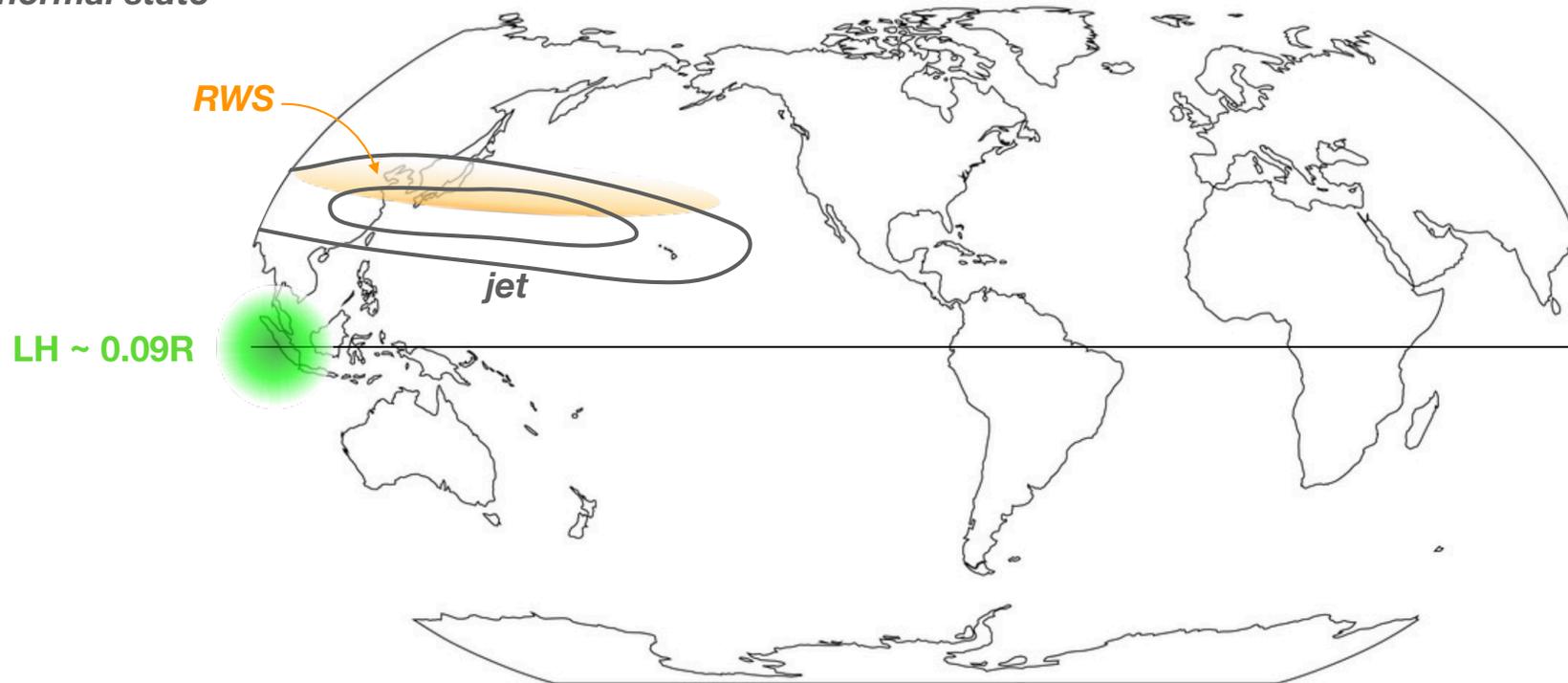


Kyle Shackelford

bulk surface flux effects for weather & climate simulation

Example: MJO teleconnections

normal state

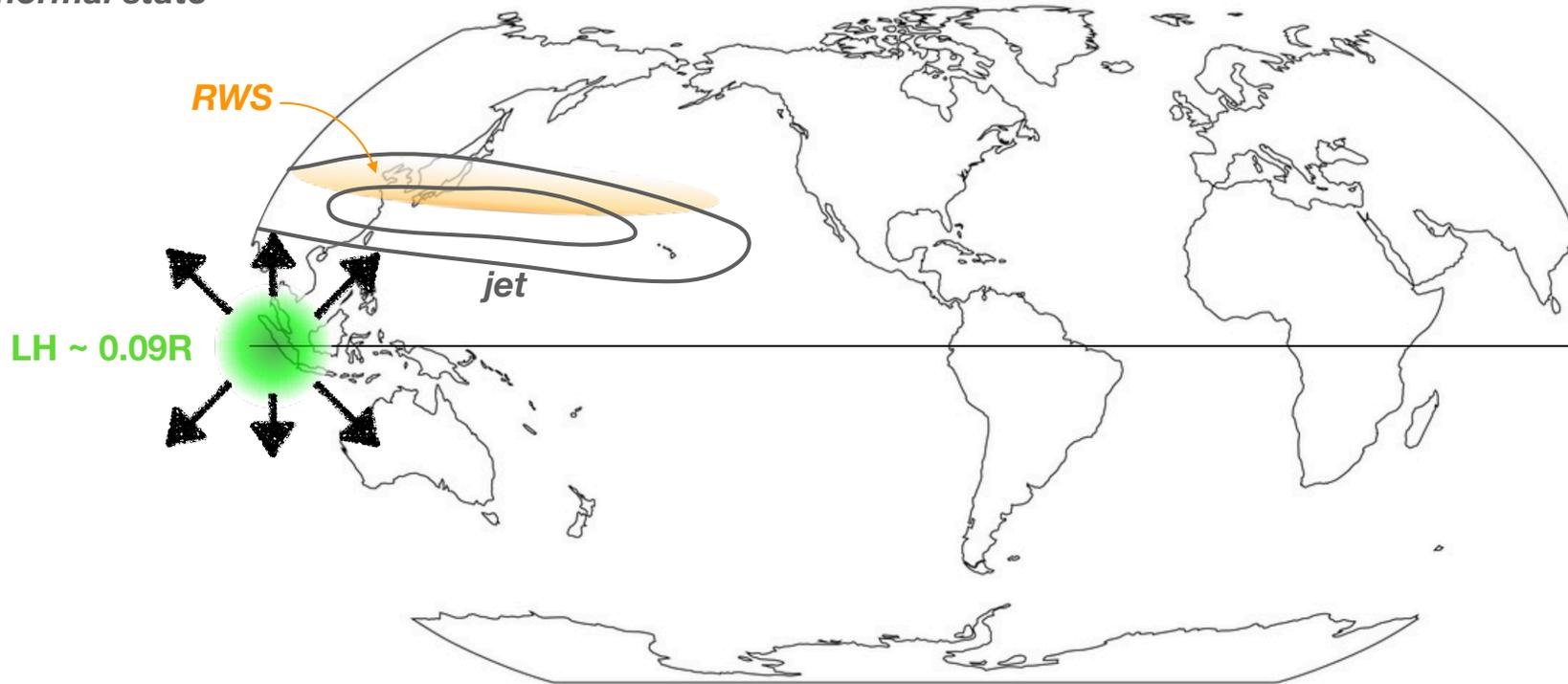


animations adapted from Henderson et al. 2017

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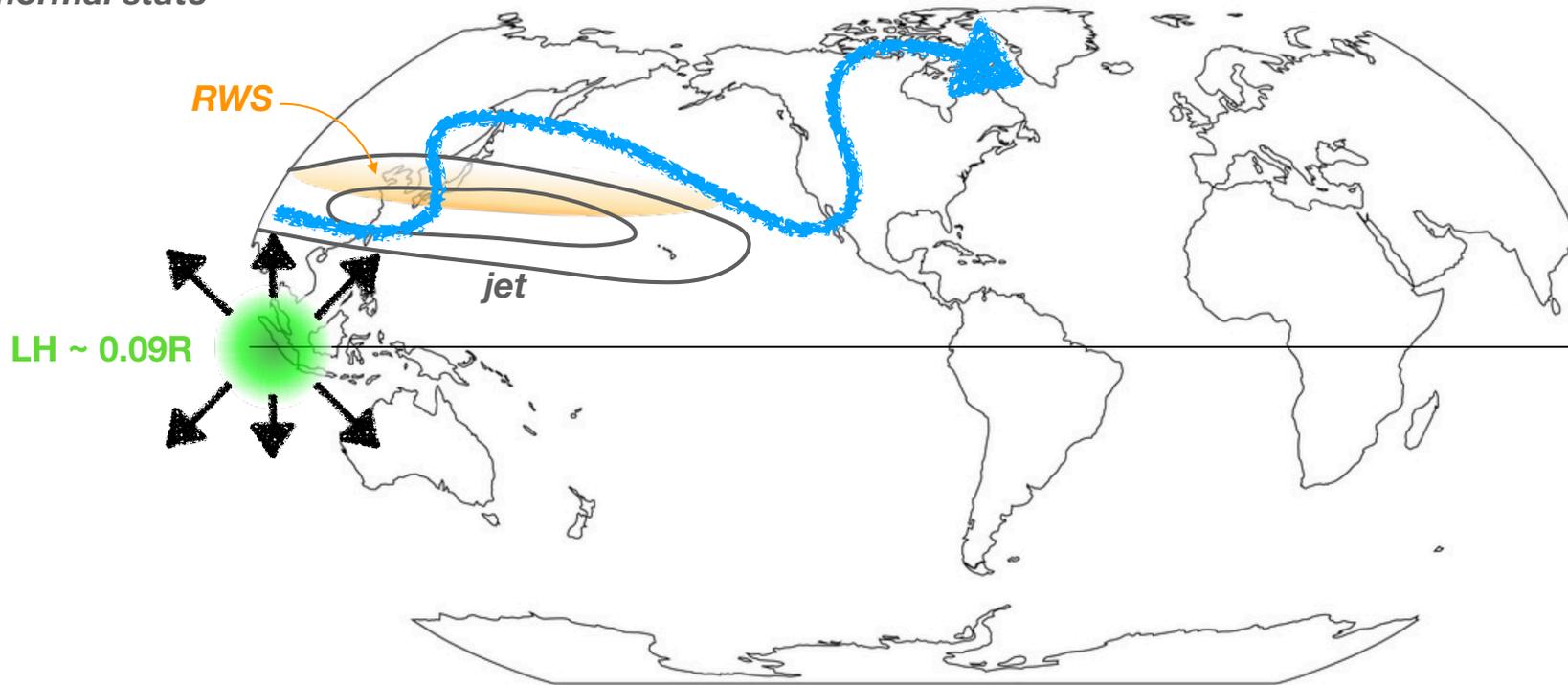


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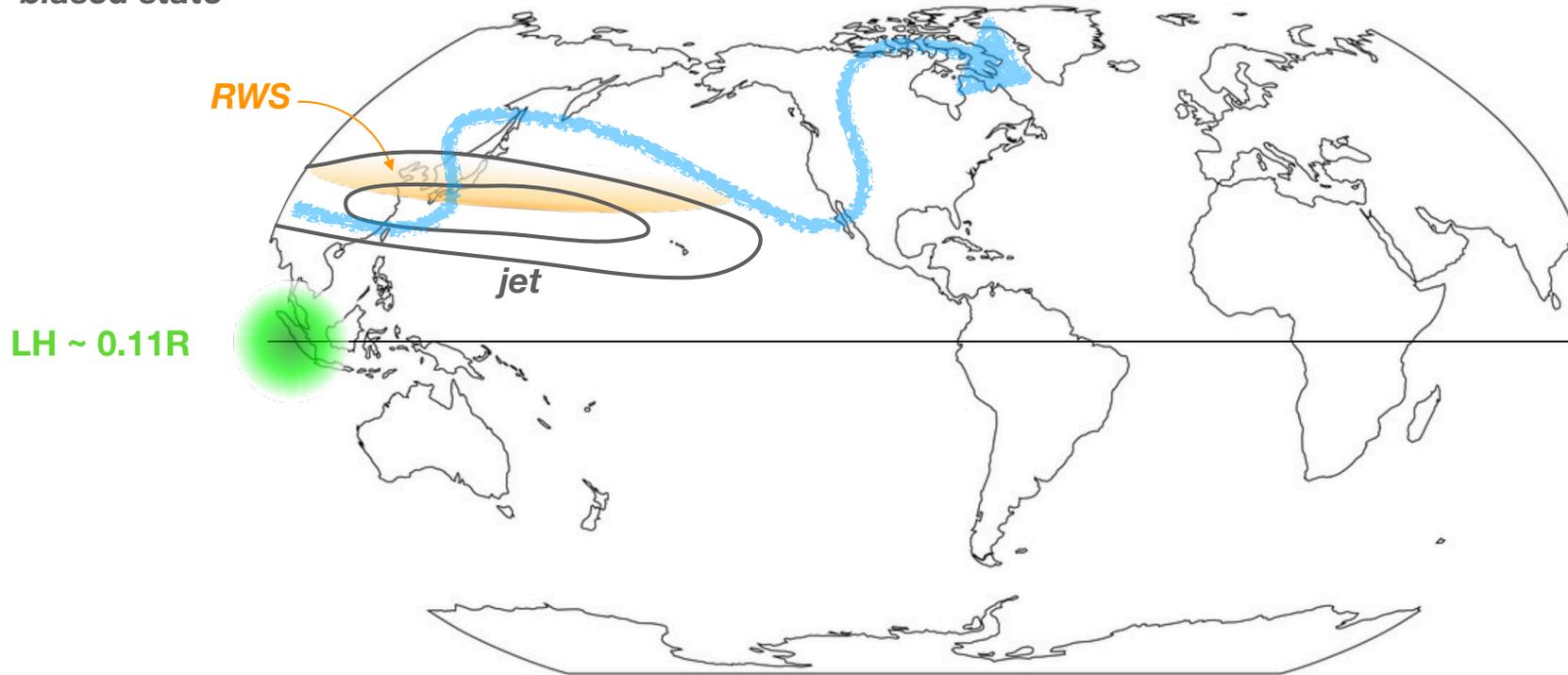


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Example: MJO teleconnections

biased state

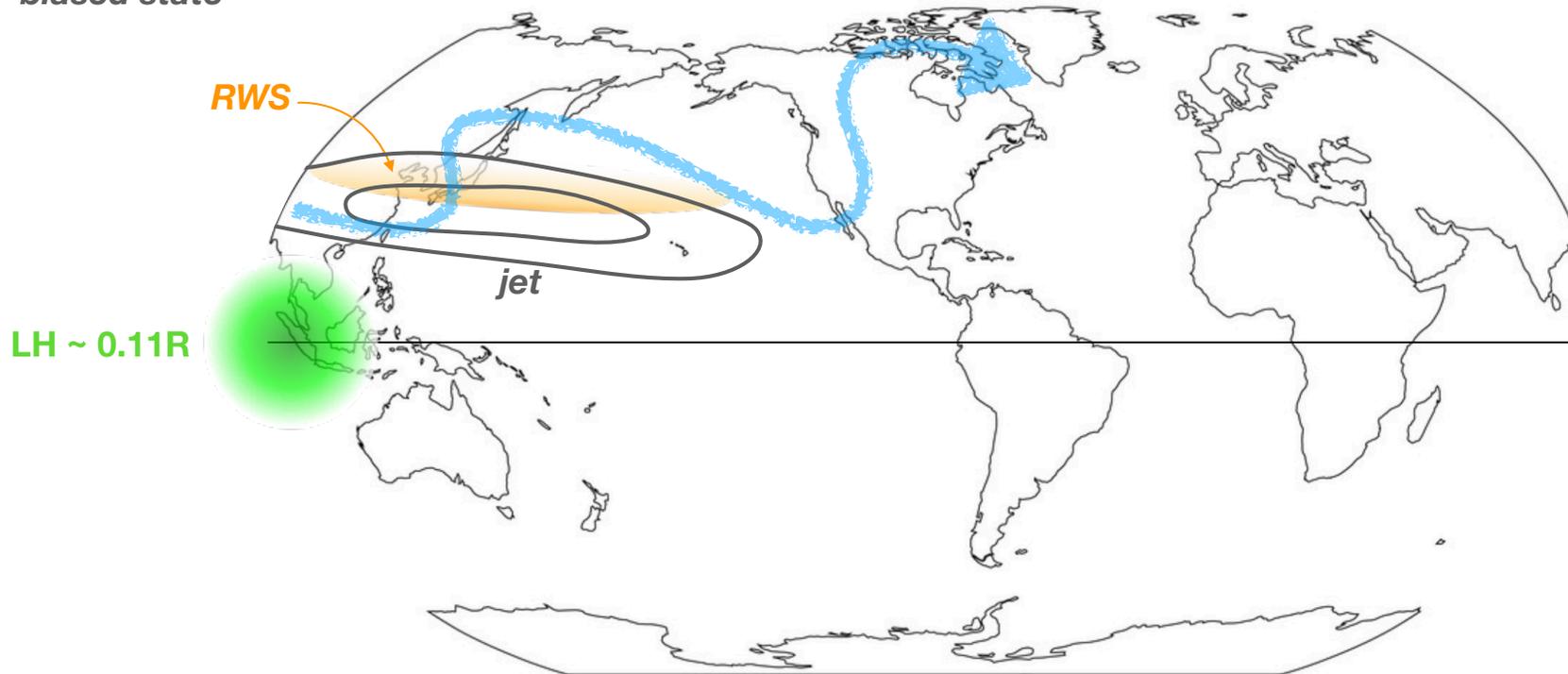


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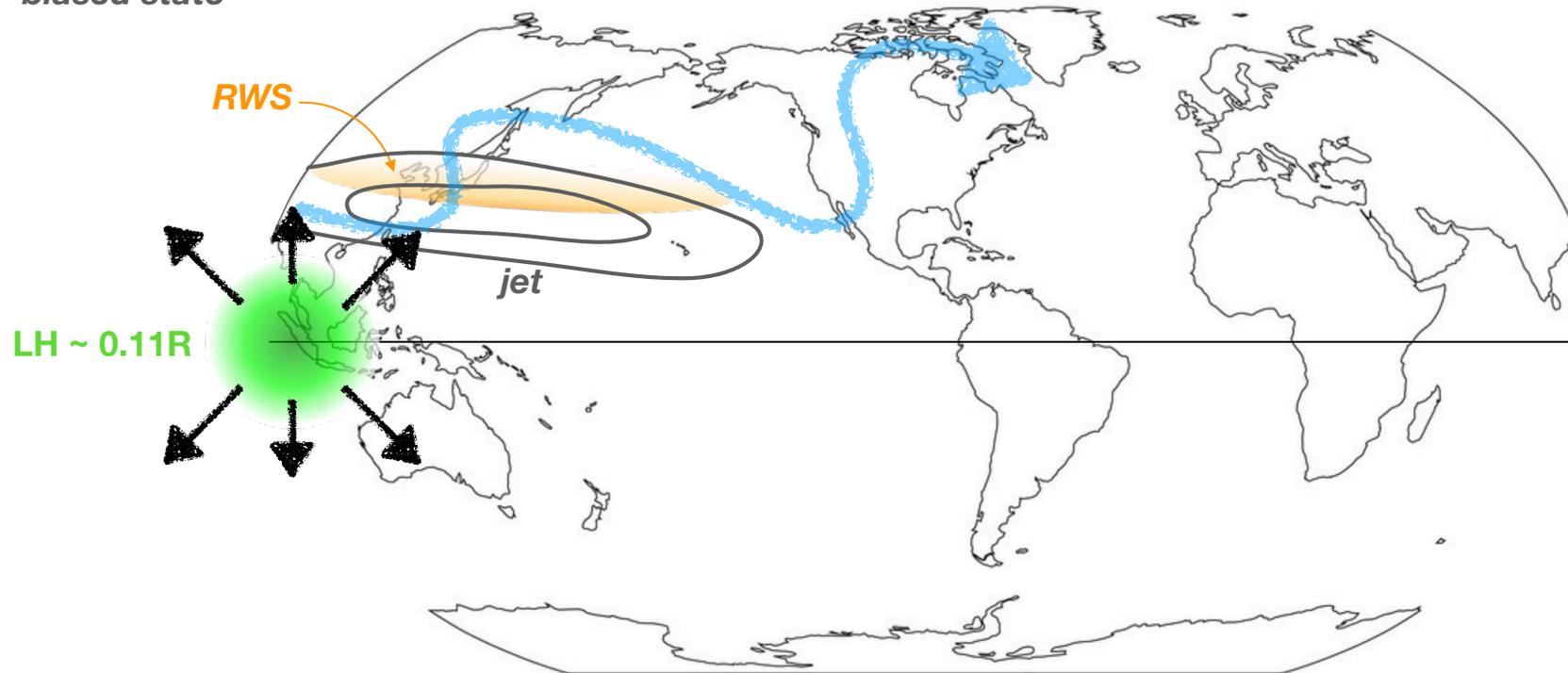


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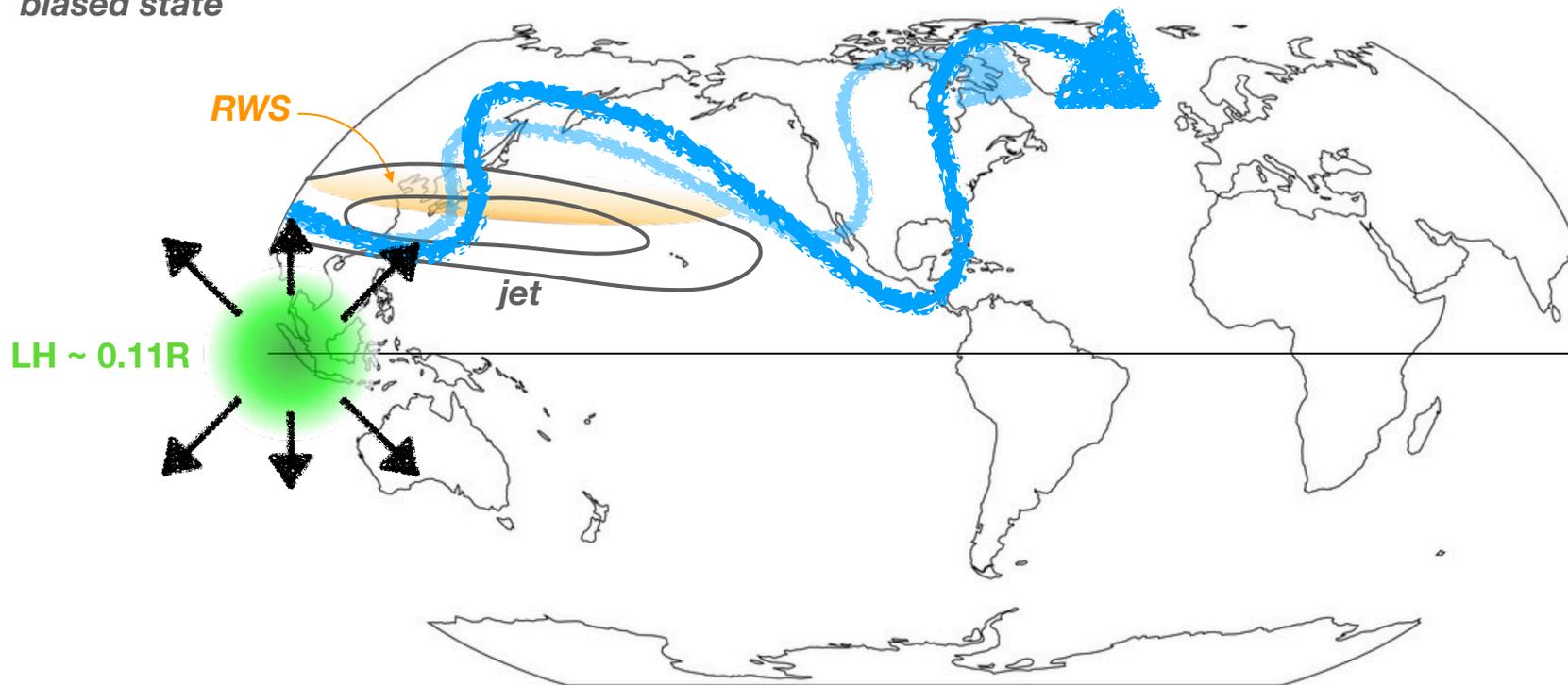


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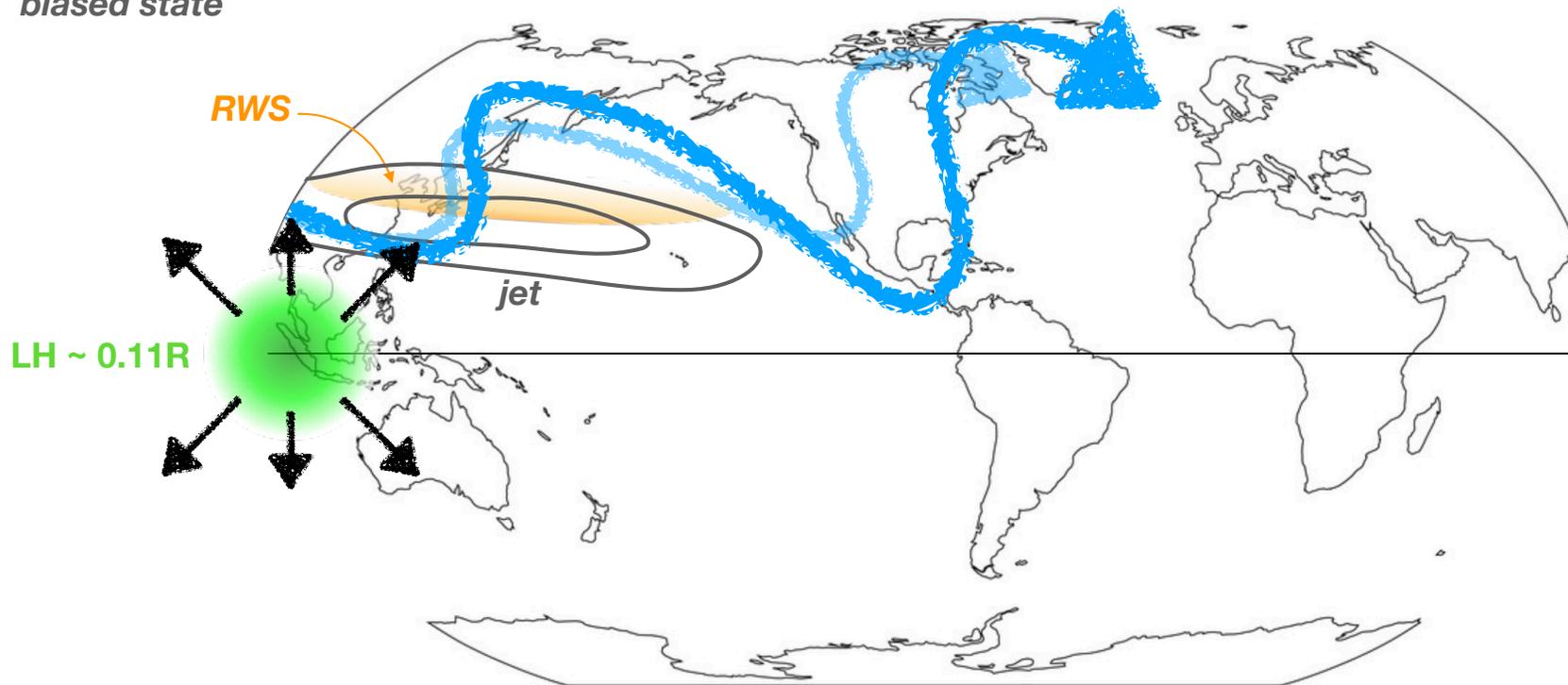


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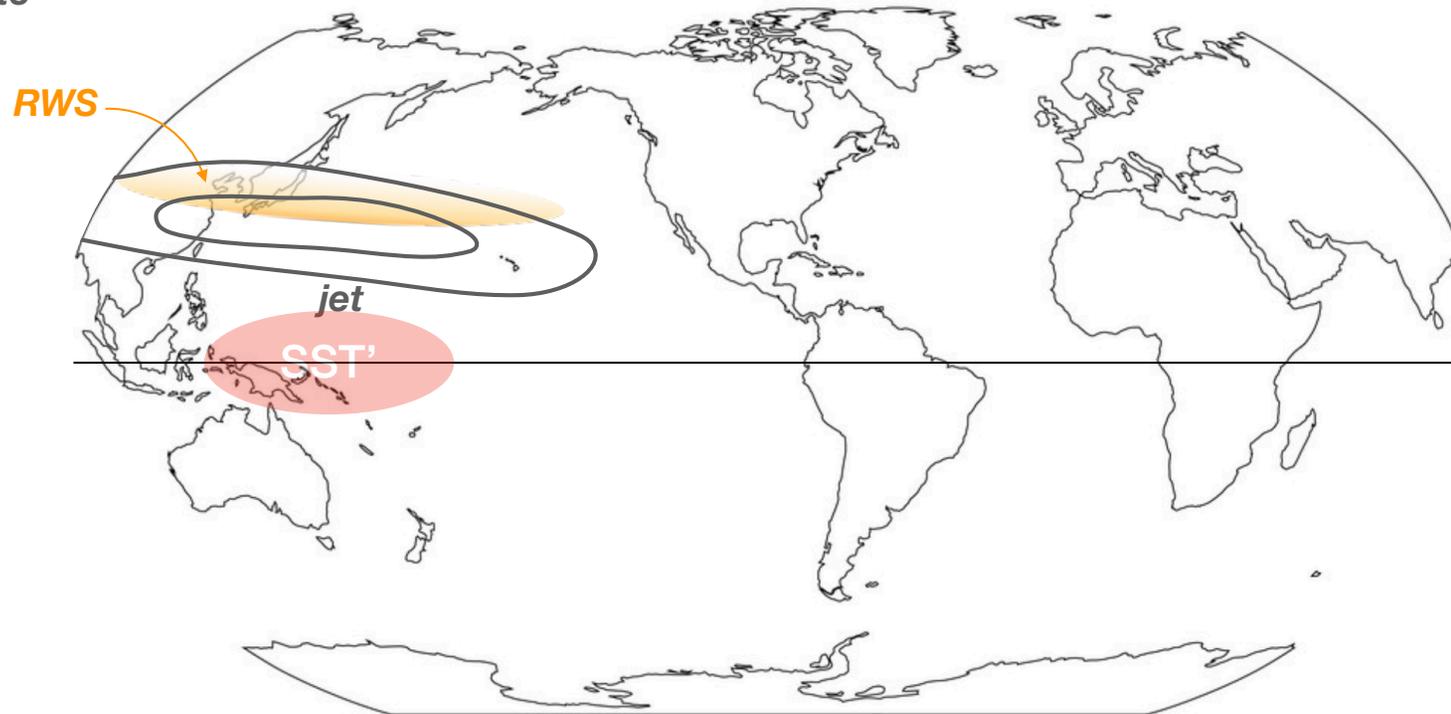
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- MJO amplitude biases -> jet stream biases

bulk surface flux effects for weather & climate simulation

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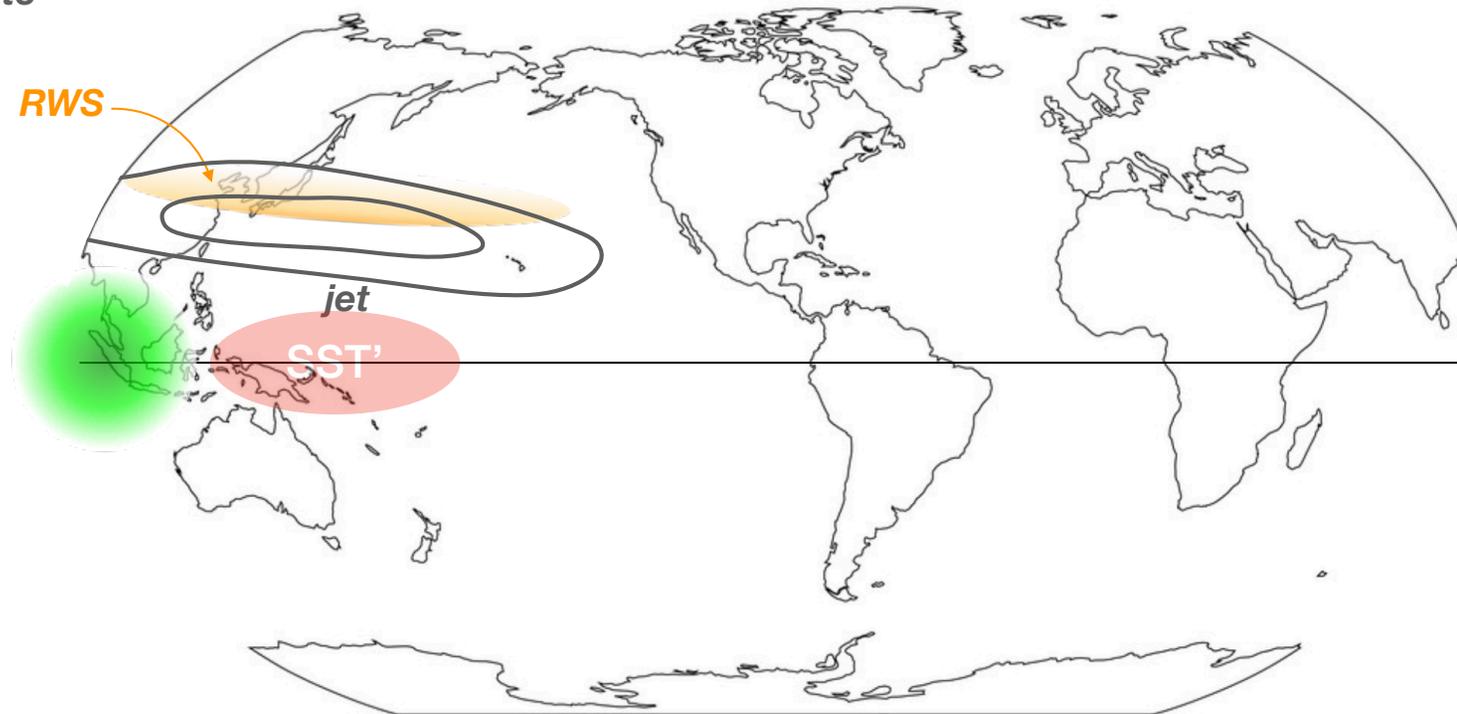
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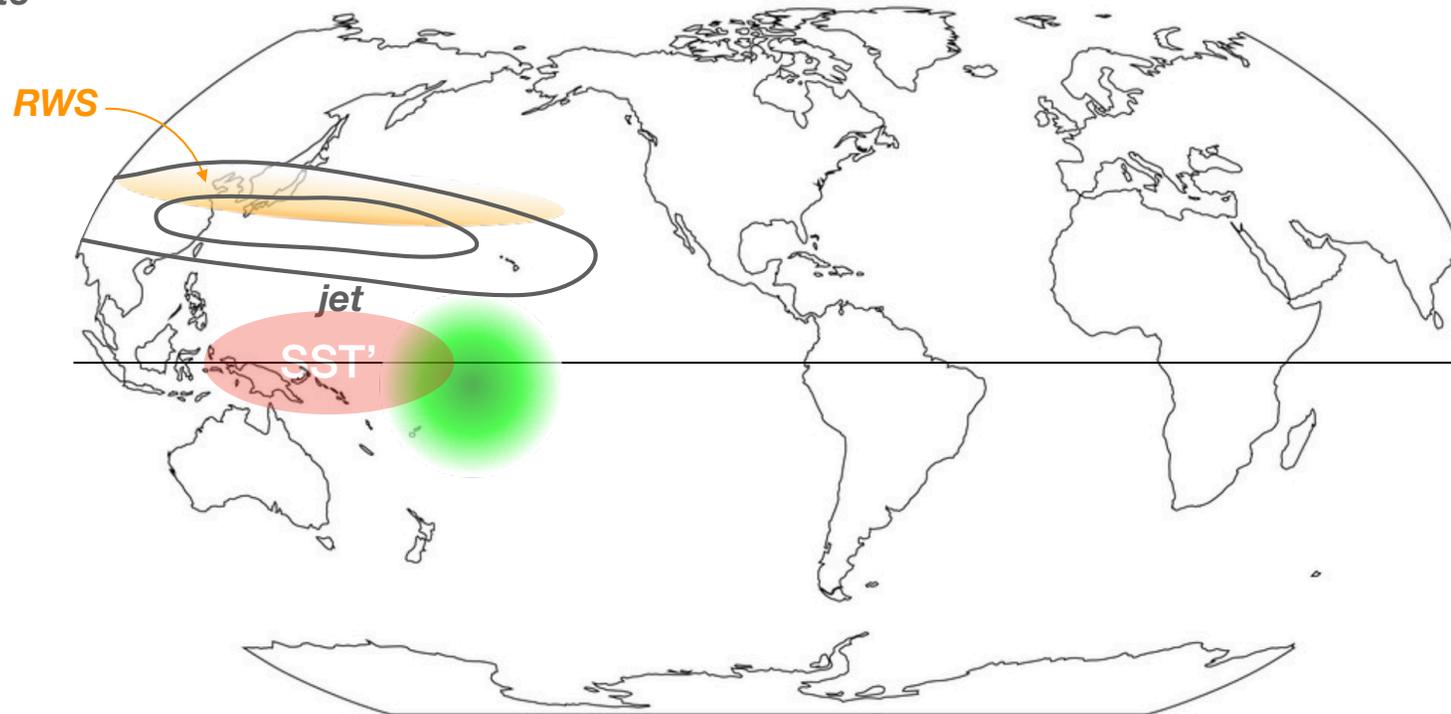
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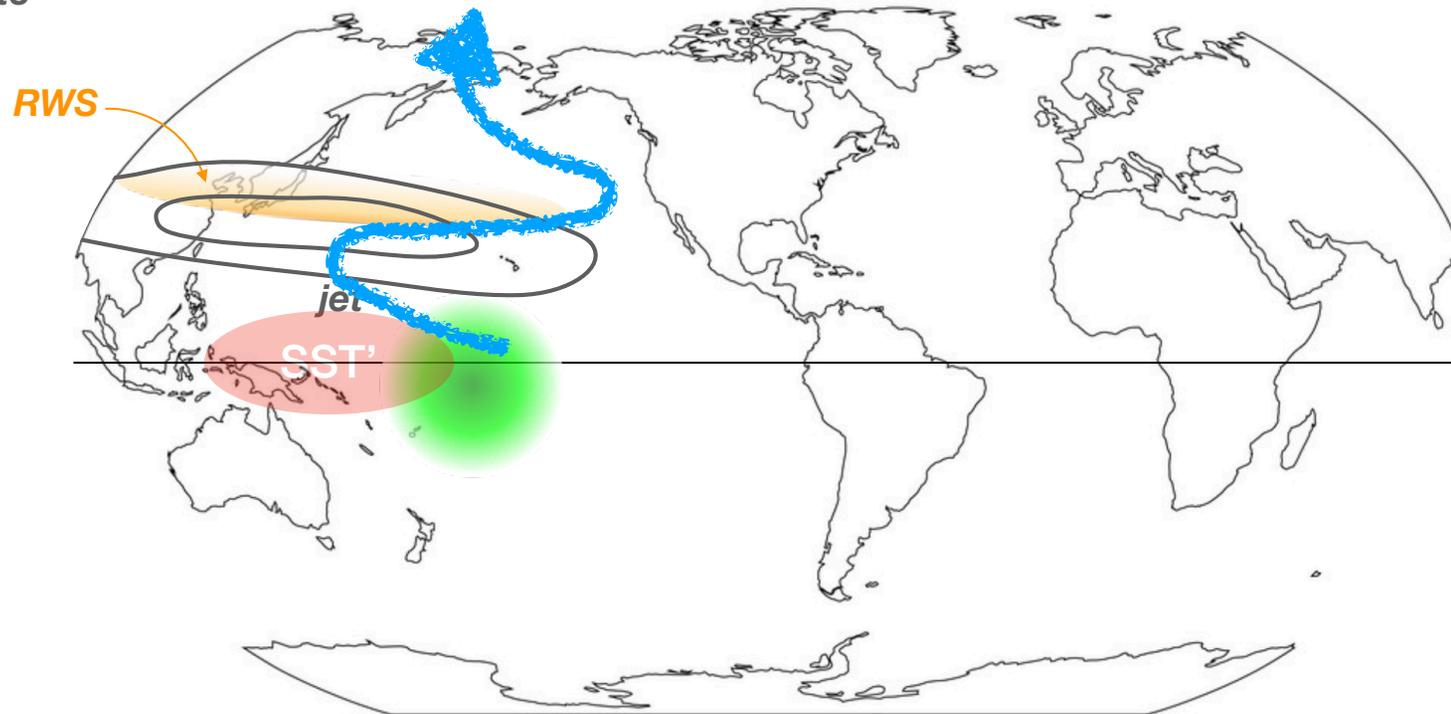
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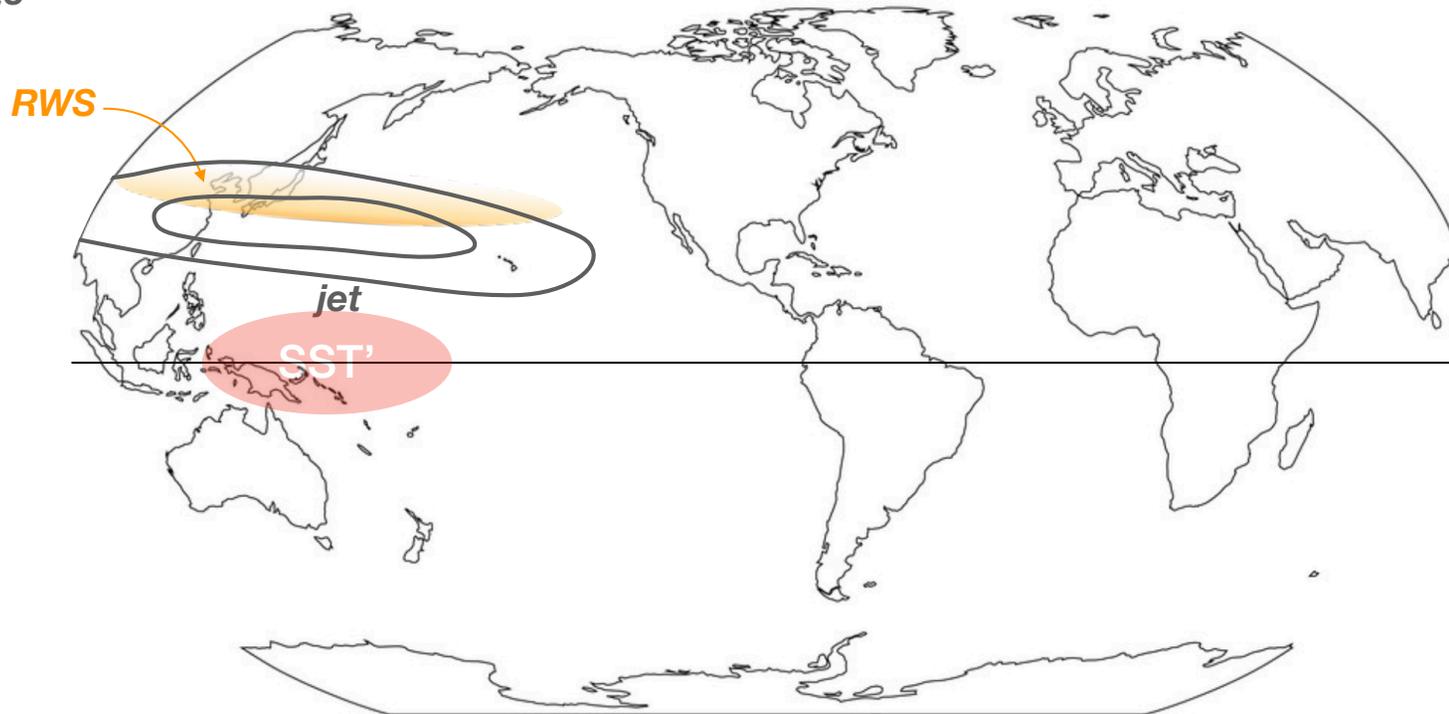
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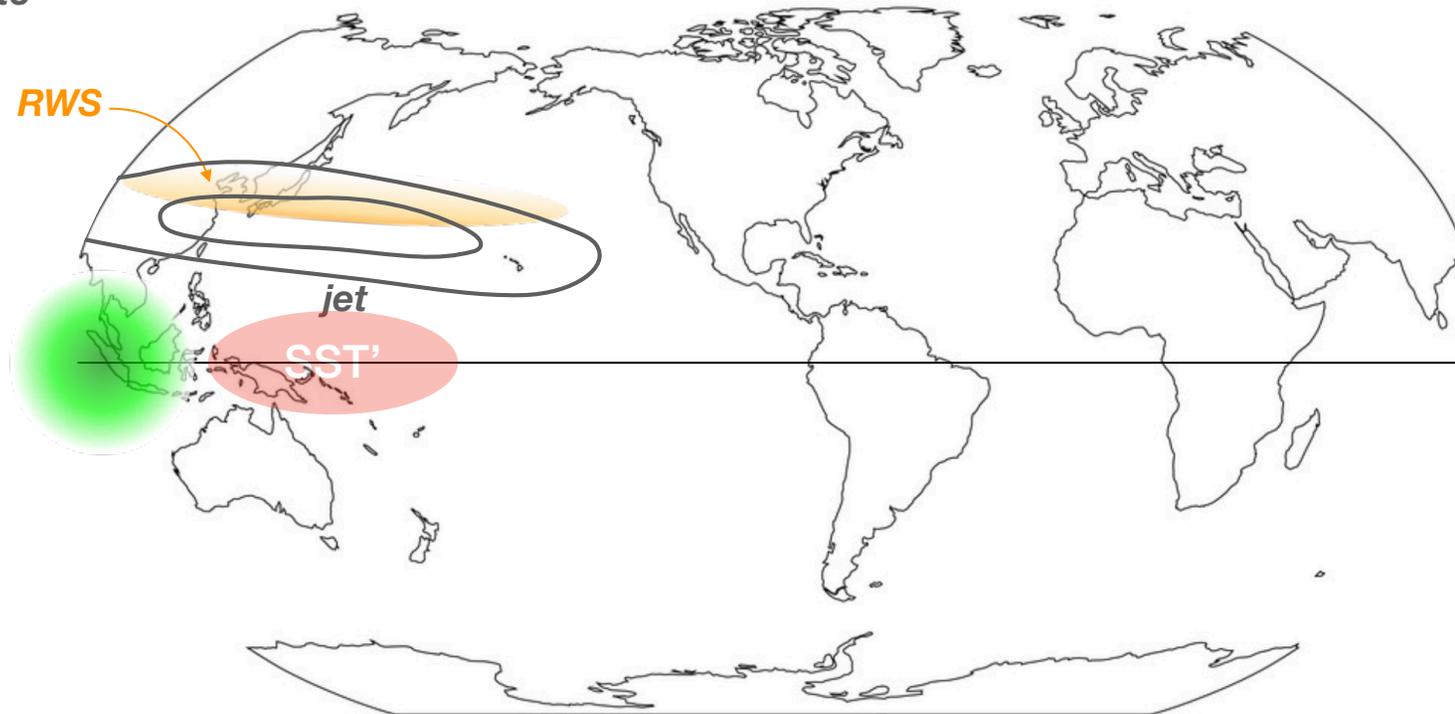
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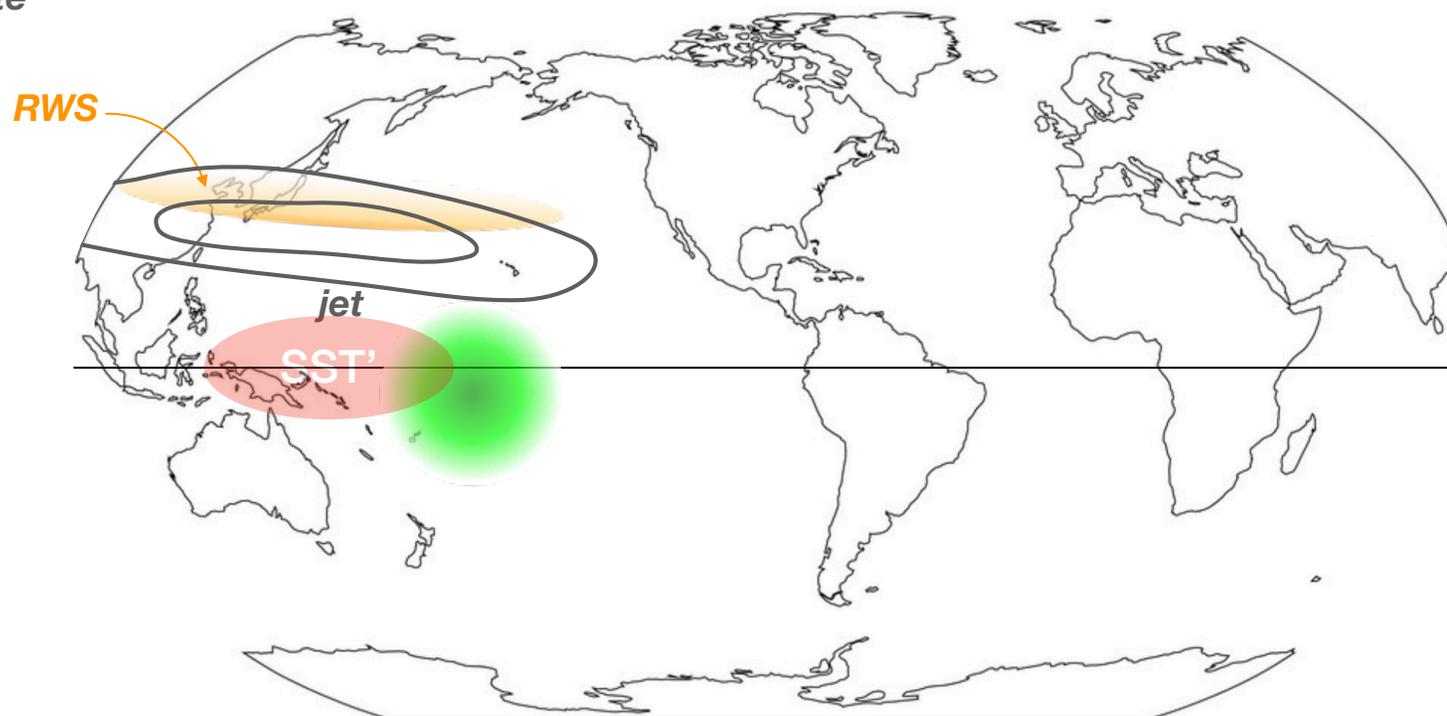
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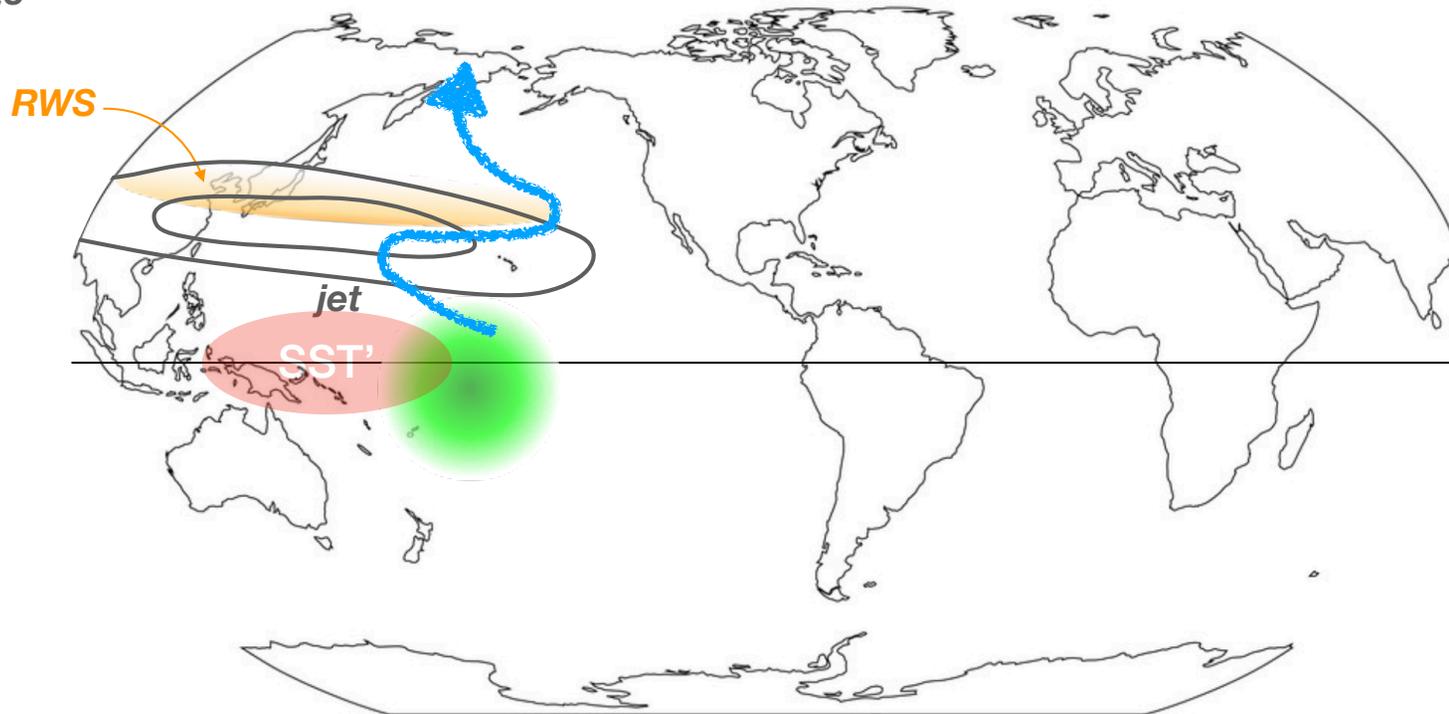
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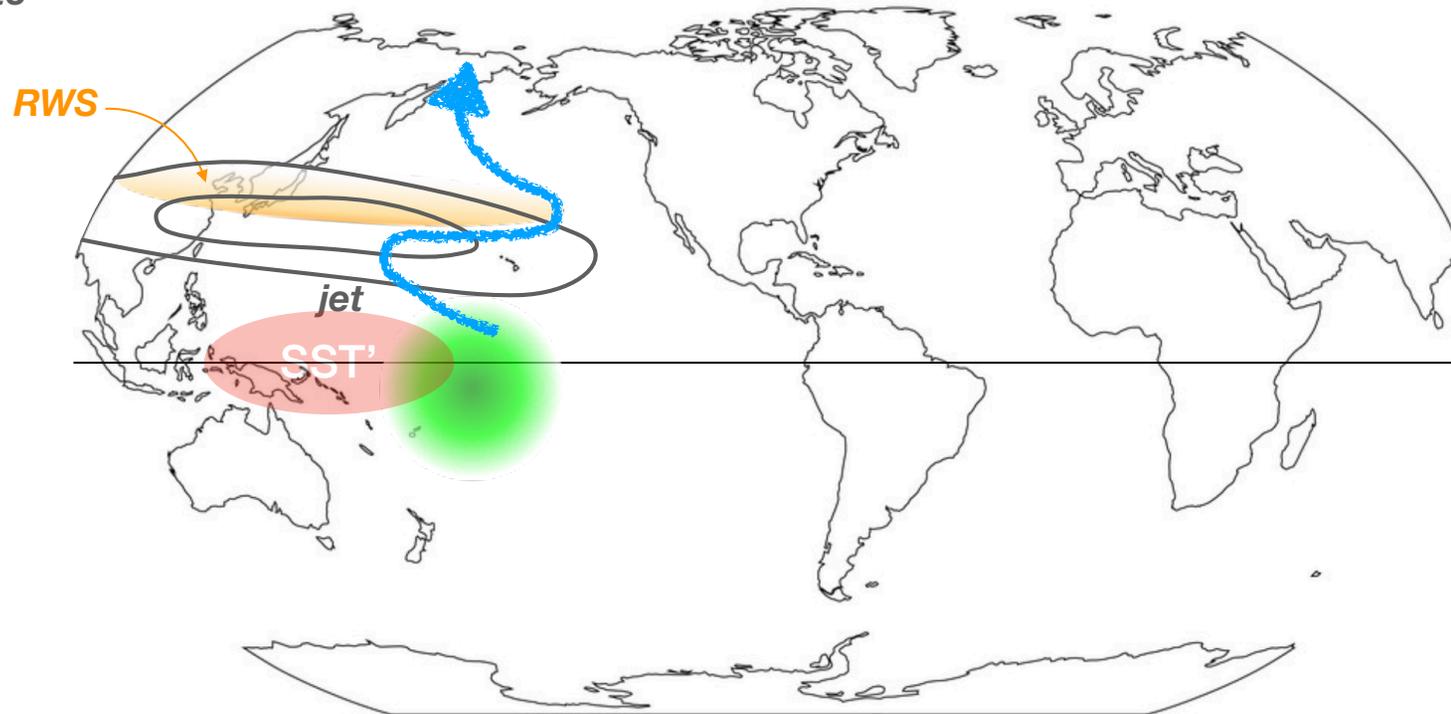
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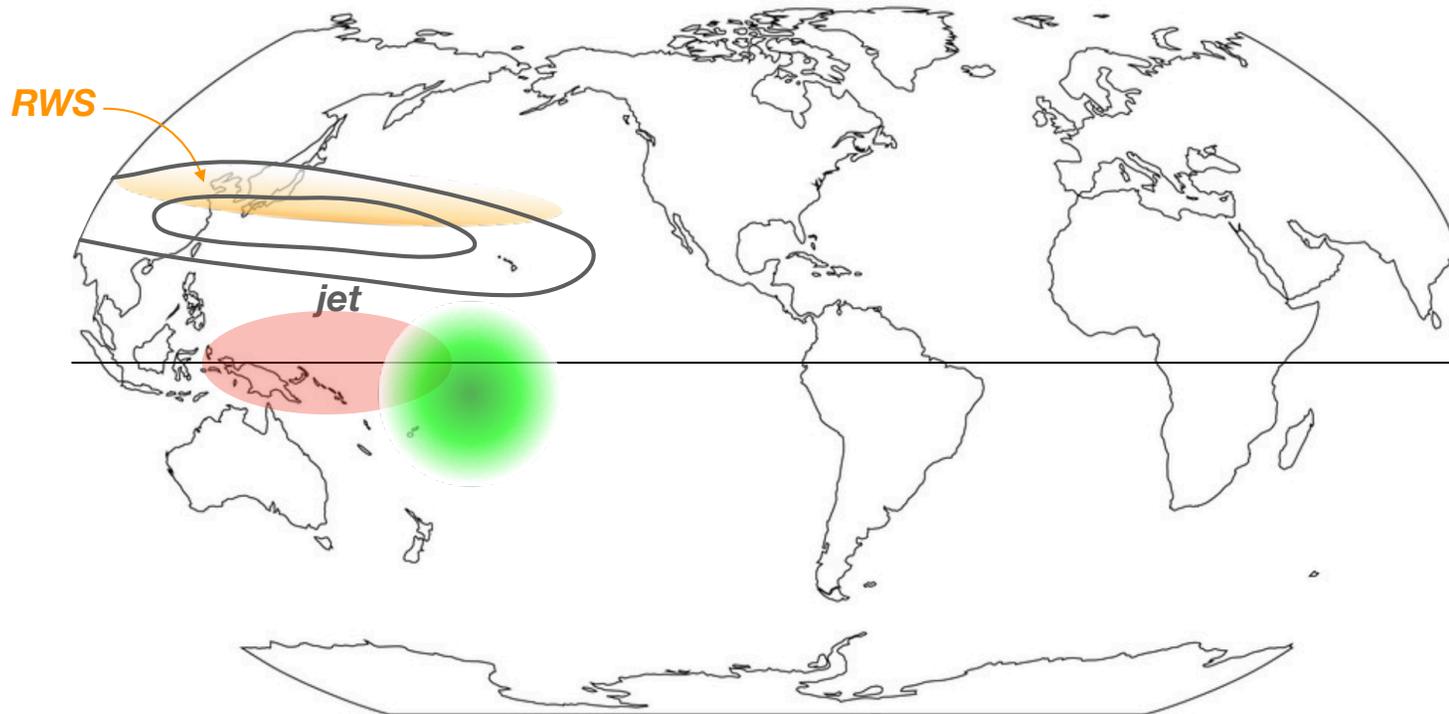


animations adapted from Henderson et al. 2017

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bulk surface flux effects for weather & climate simulation

Example: MJO-ENSO interactions

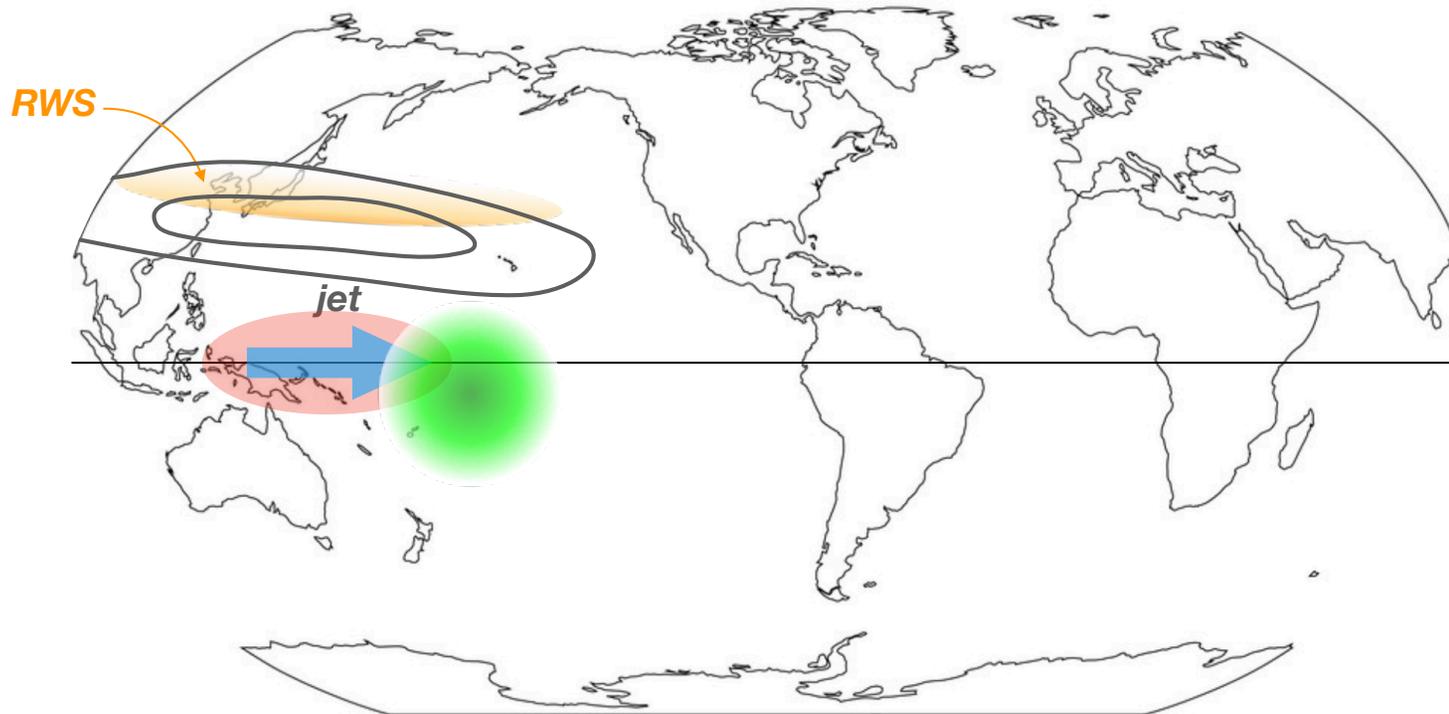


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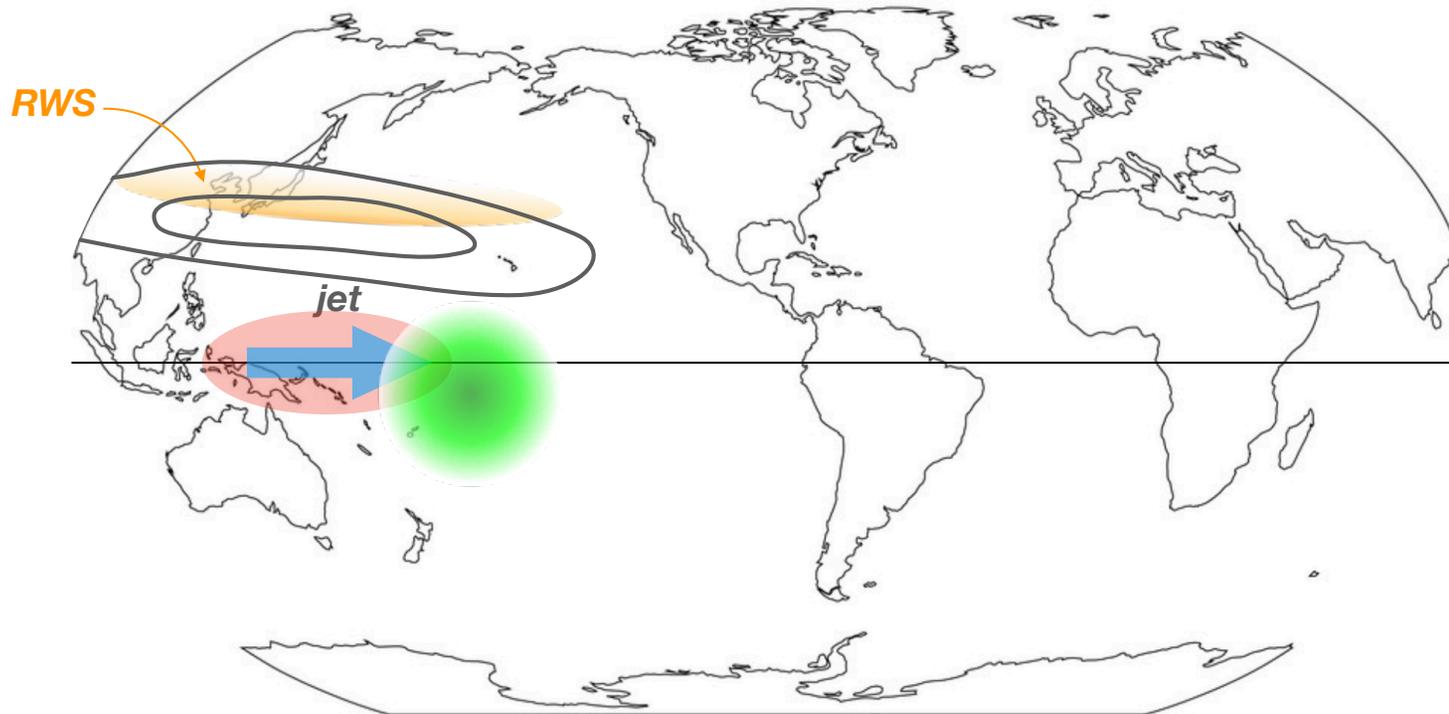


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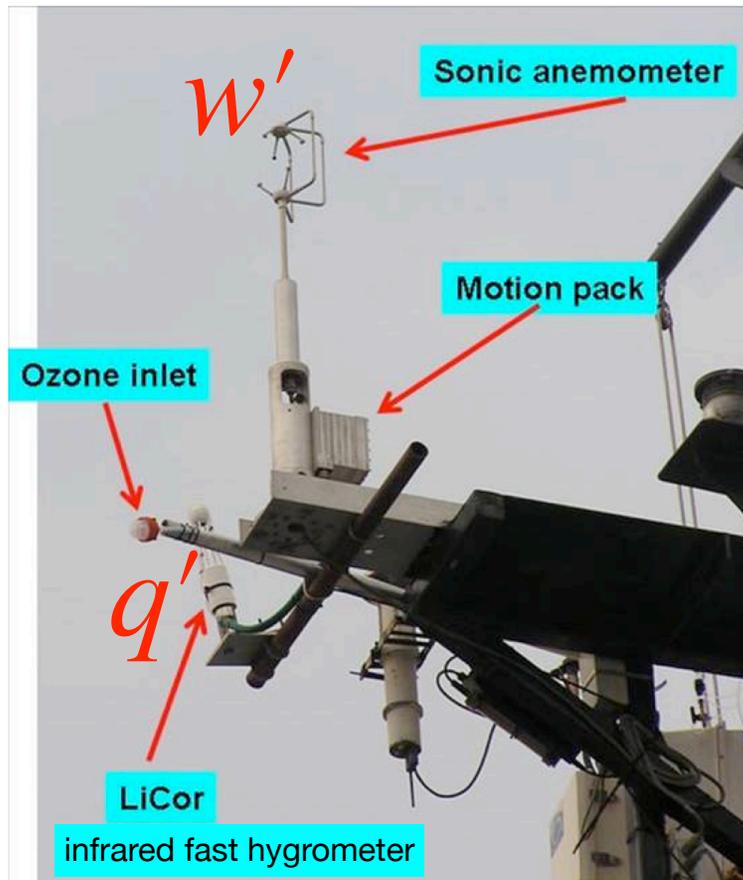
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- MJO amplitude biases -> jet stream biases
- MJO propagation biases -> jet stream biases
- MJO propagation biases -> ocean Kelvin wave, ENSO biases

surface fluxes: measurement versus estimation

direct covariance measurement

$$LH = L_e \overline{w'q'}$$



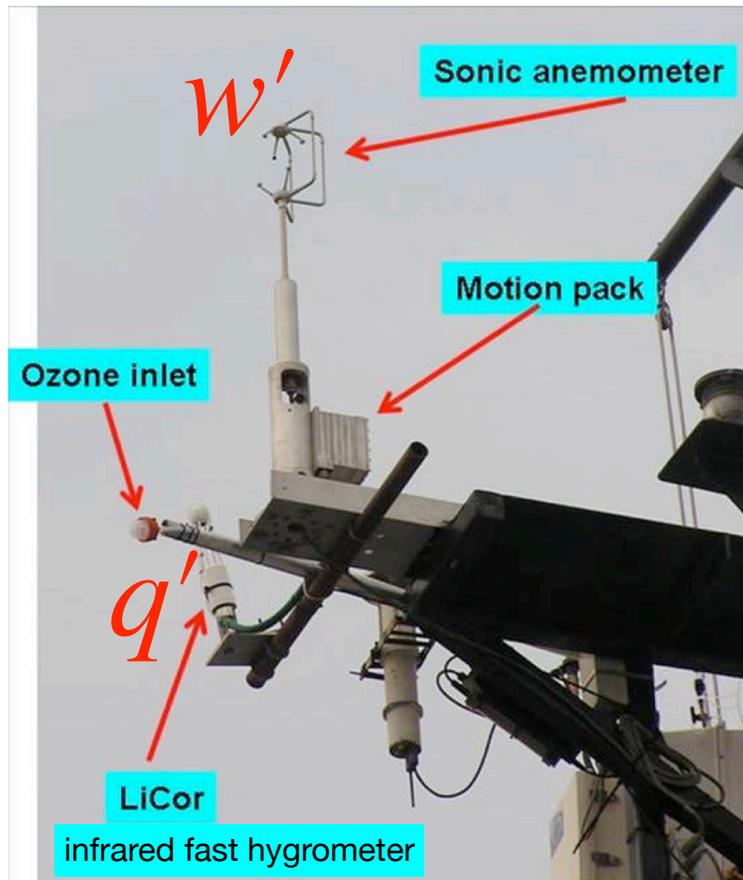
surface fluxes: measurement versus estimation

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bulk estimation

$$LH = L_v C_e |V| (q_{SST}^* - q_{2m})$$

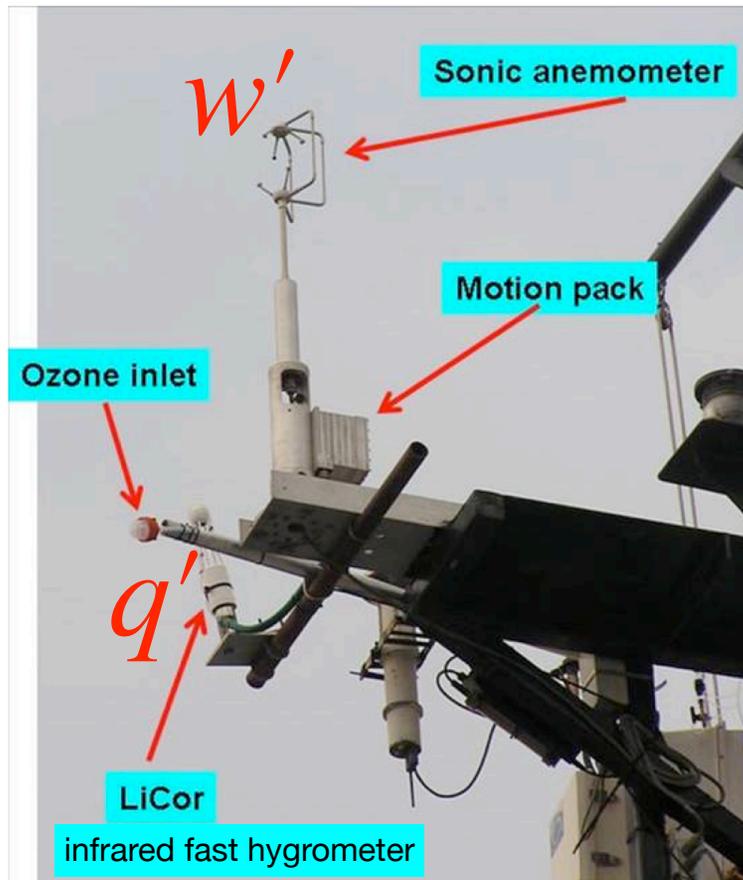


Weller et al. 2008

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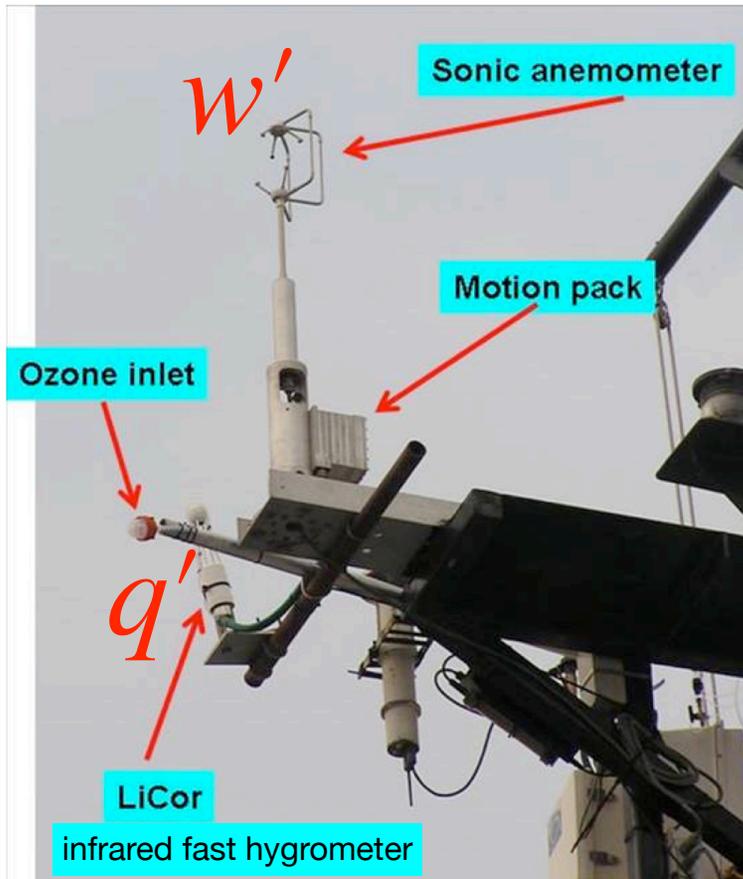
parameterization

bulk inputs

surface fluxes: measurement versus estimation

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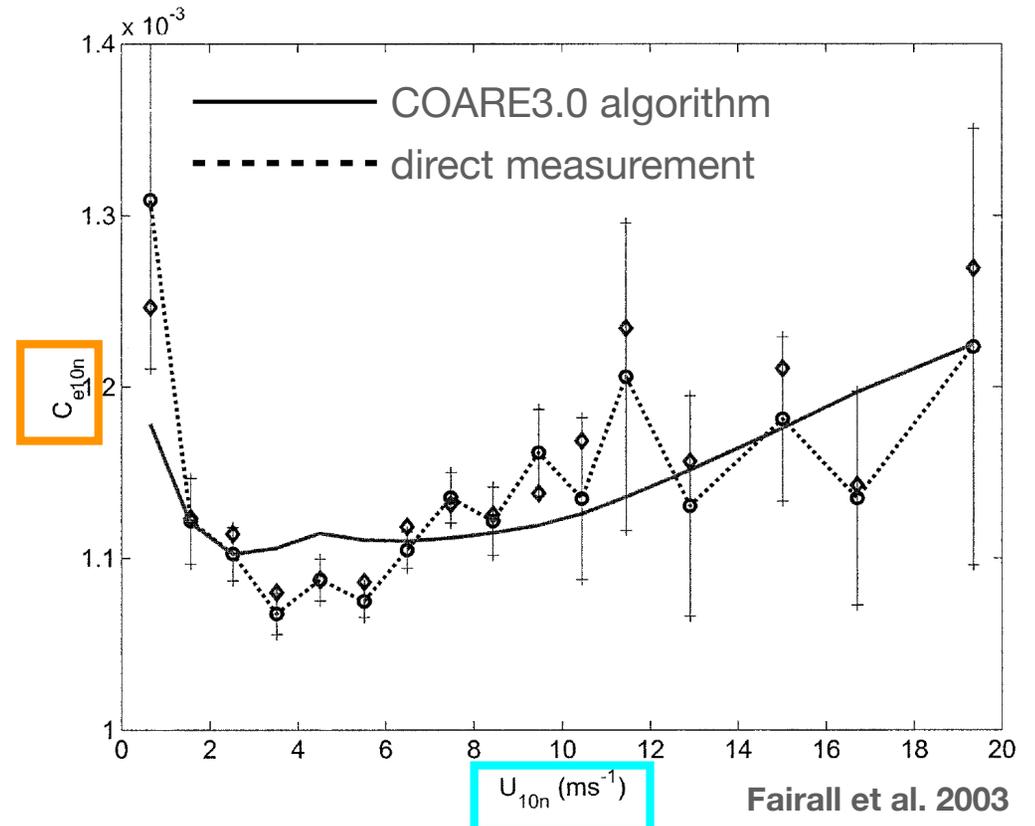


Weller et al. 2008

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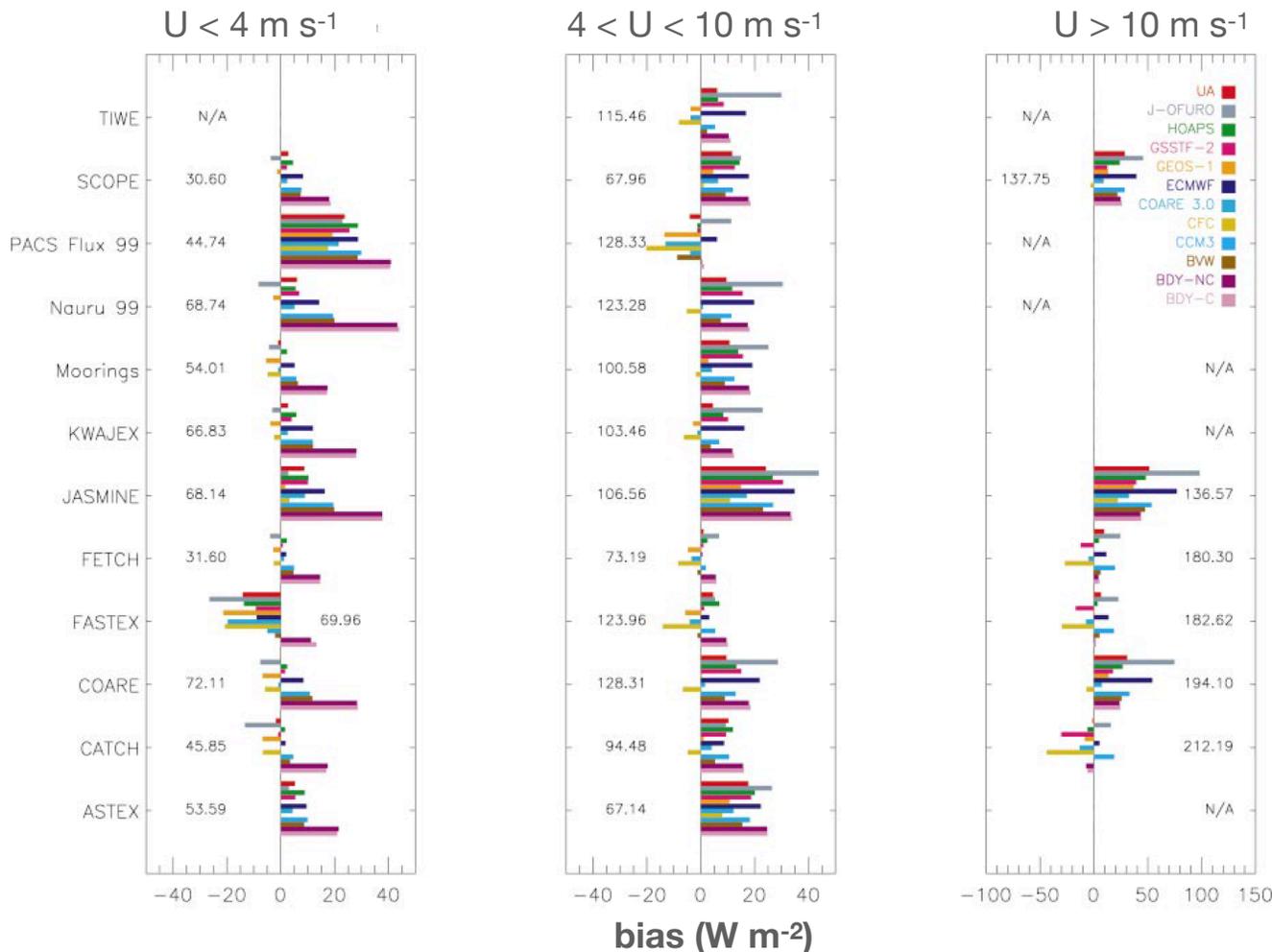
parameterization bulk inputs



Fairall et al. 2003

surface flux biases among bulk algorithms

- climate and forecast models compute surface fluxes using a variety of bulk surface flux algorithms
- most algorithms overestimate surface fluxes



10-20% biases
on average

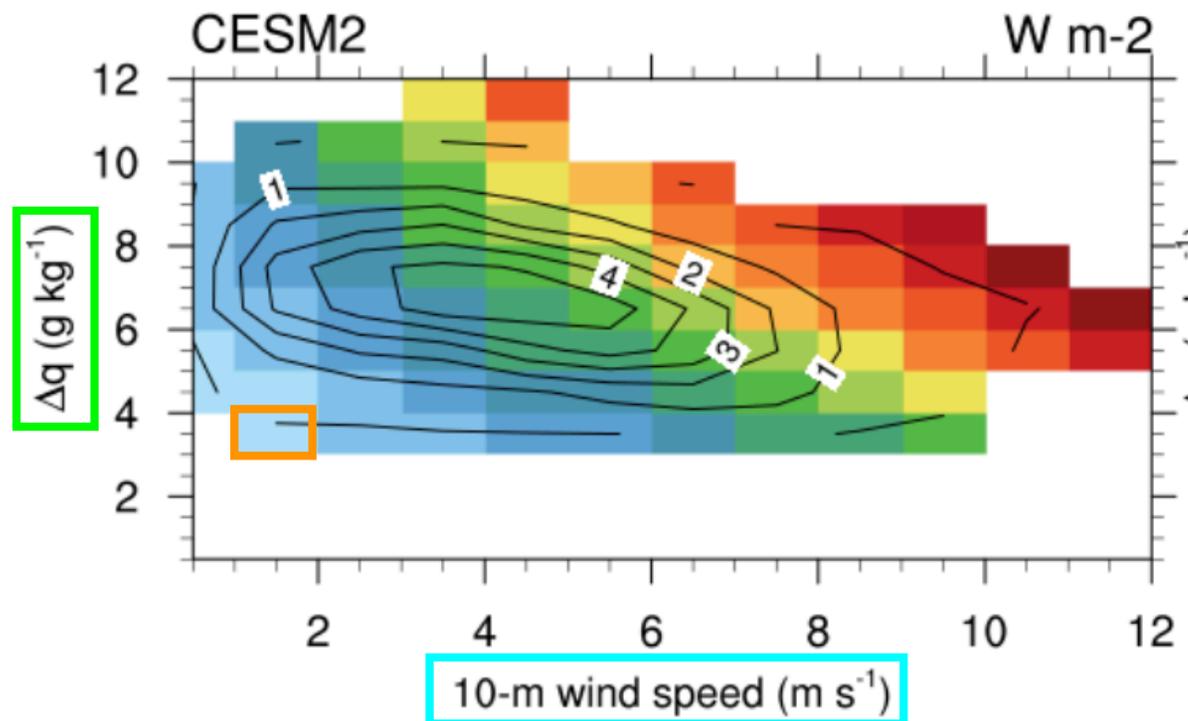
COARE algorithm is
among the “least
problematic”

Brunke et al. (2003)

surface fluxes: how to diagnose their biases in models?

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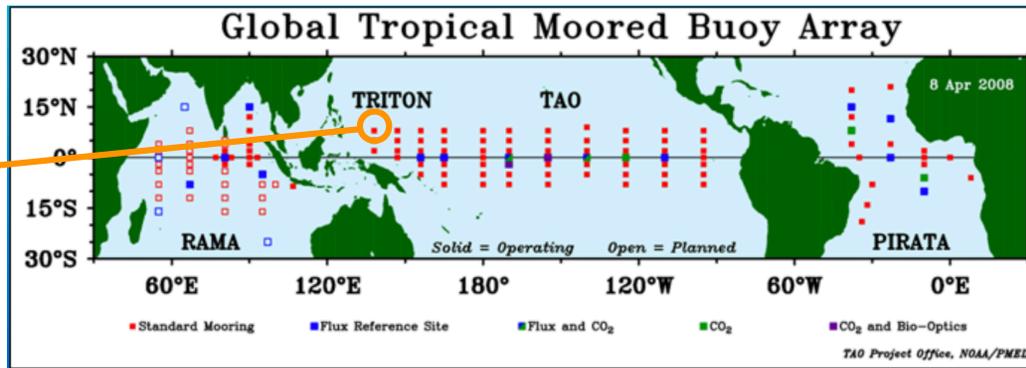
$$LH = L_v C_e |V| \Delta q$$



surface fluxes: how to diagnose their biases in models?

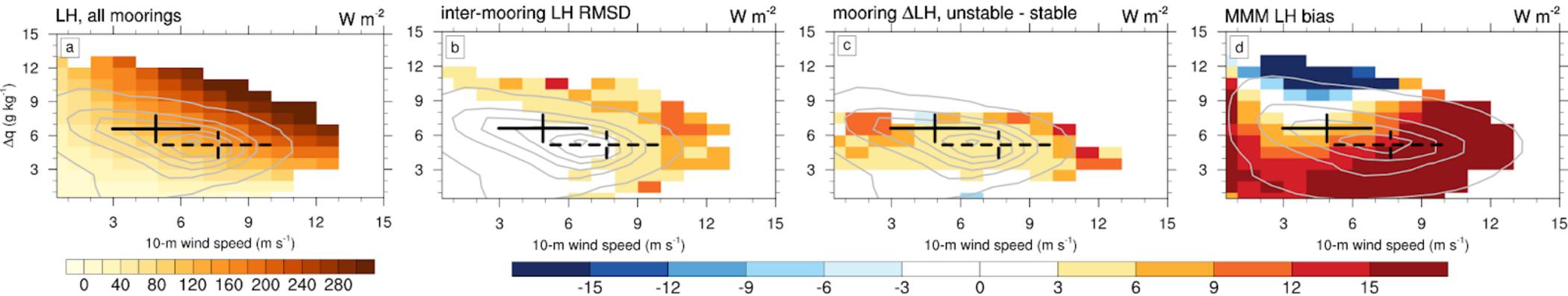


McPhaden (2008)

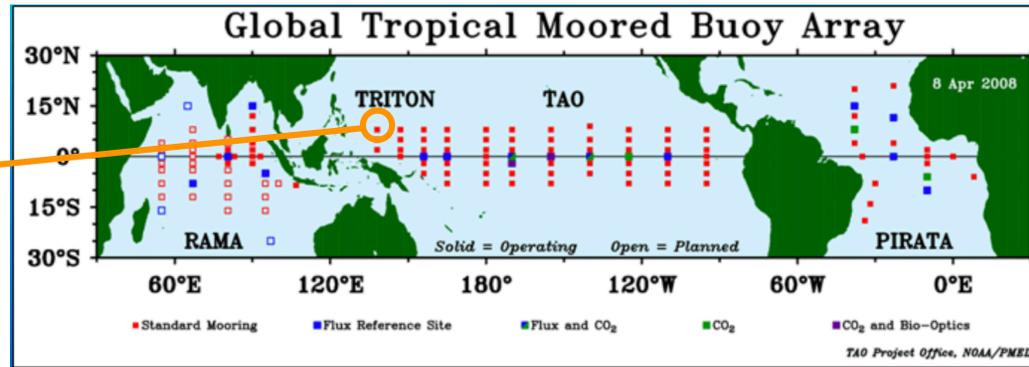


NOAA

surface fluxes: how to diagnose their biases in models?

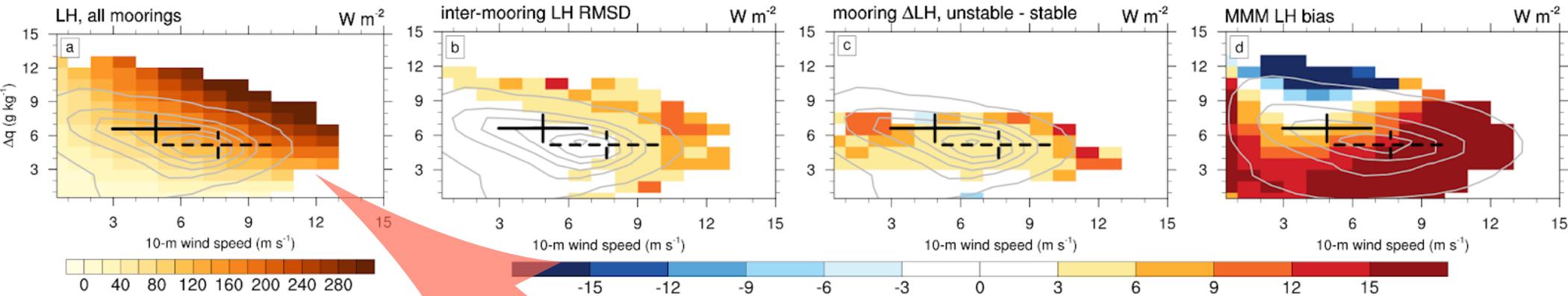


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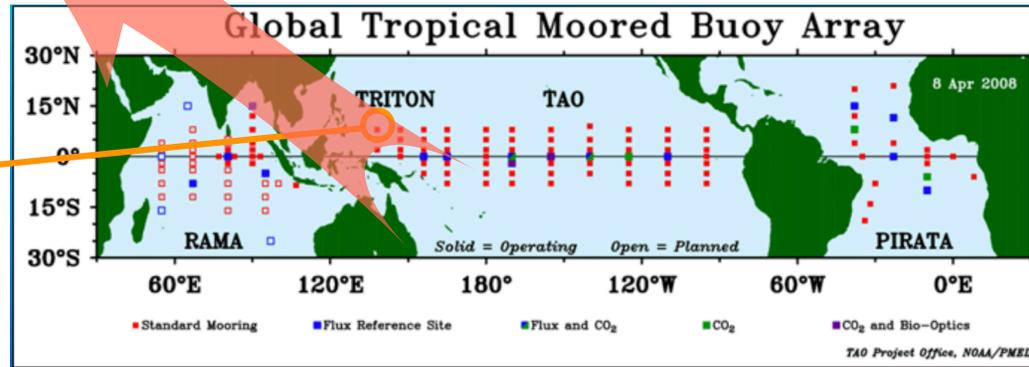


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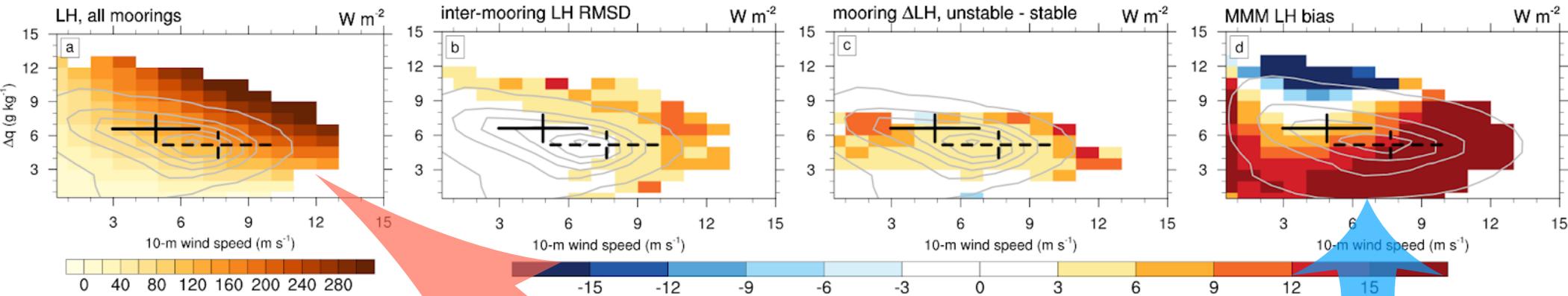


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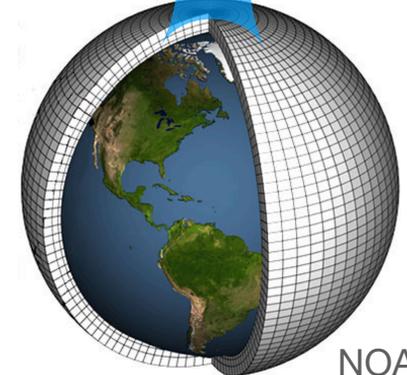
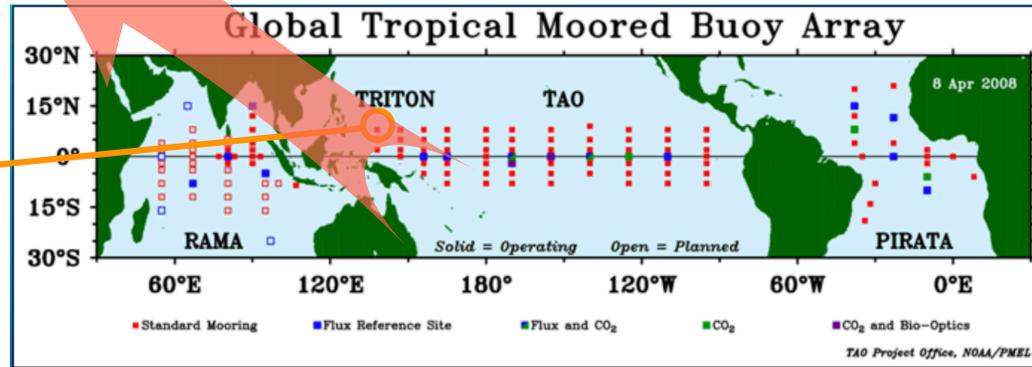


NOAA

surface fluxes: how to diagnose their biases in models?

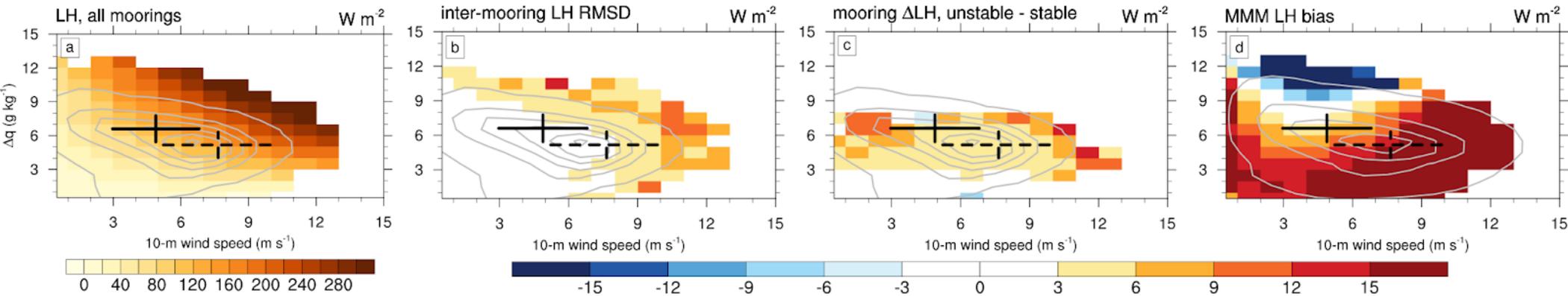


McPhaden (2008)

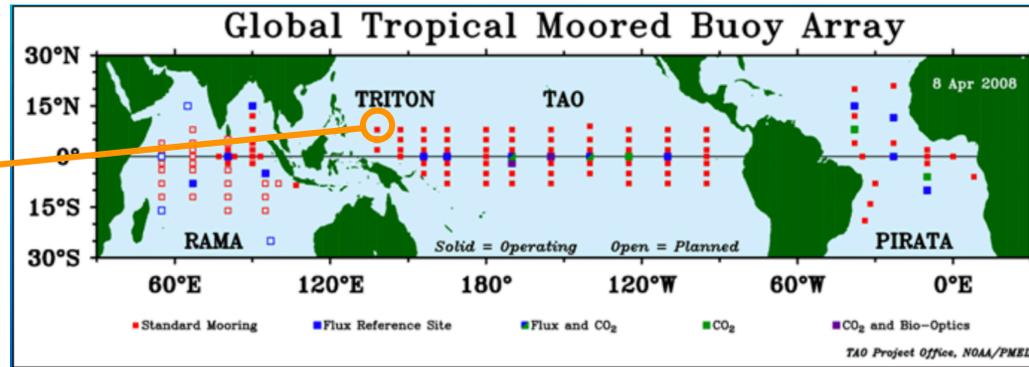


NOAA

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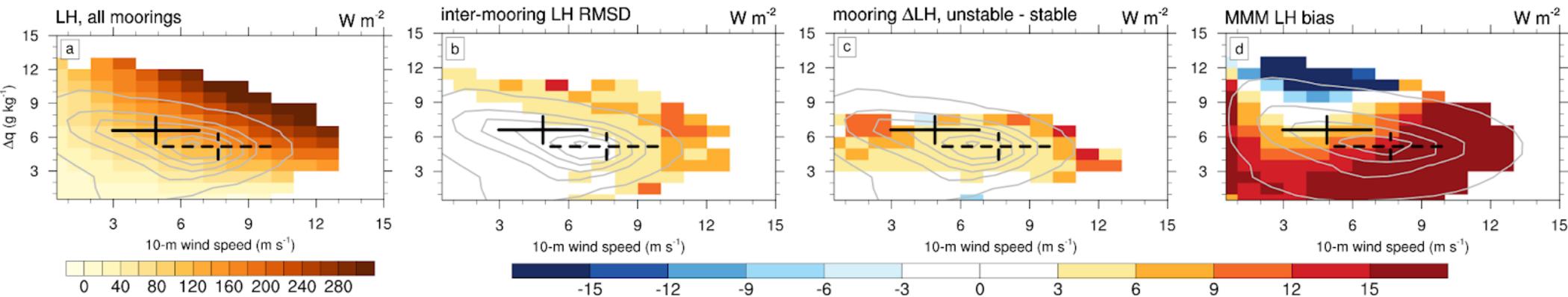


McPhaden (2008)

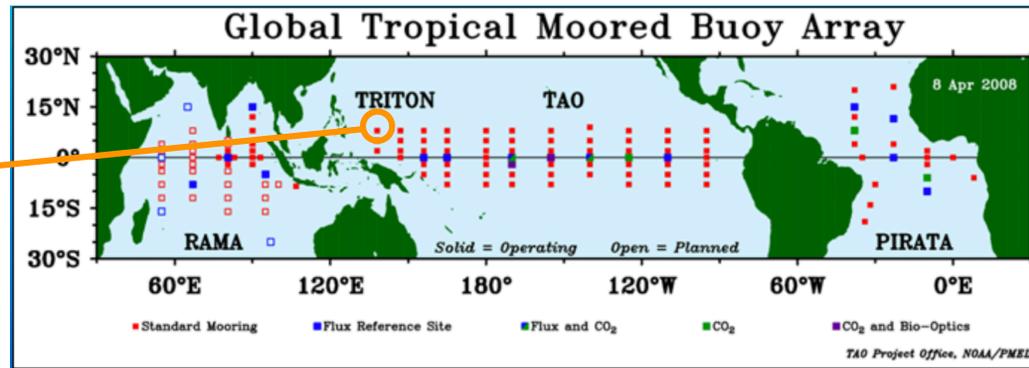


NOAA

surface fluxes: how to diagnose their biases in models?



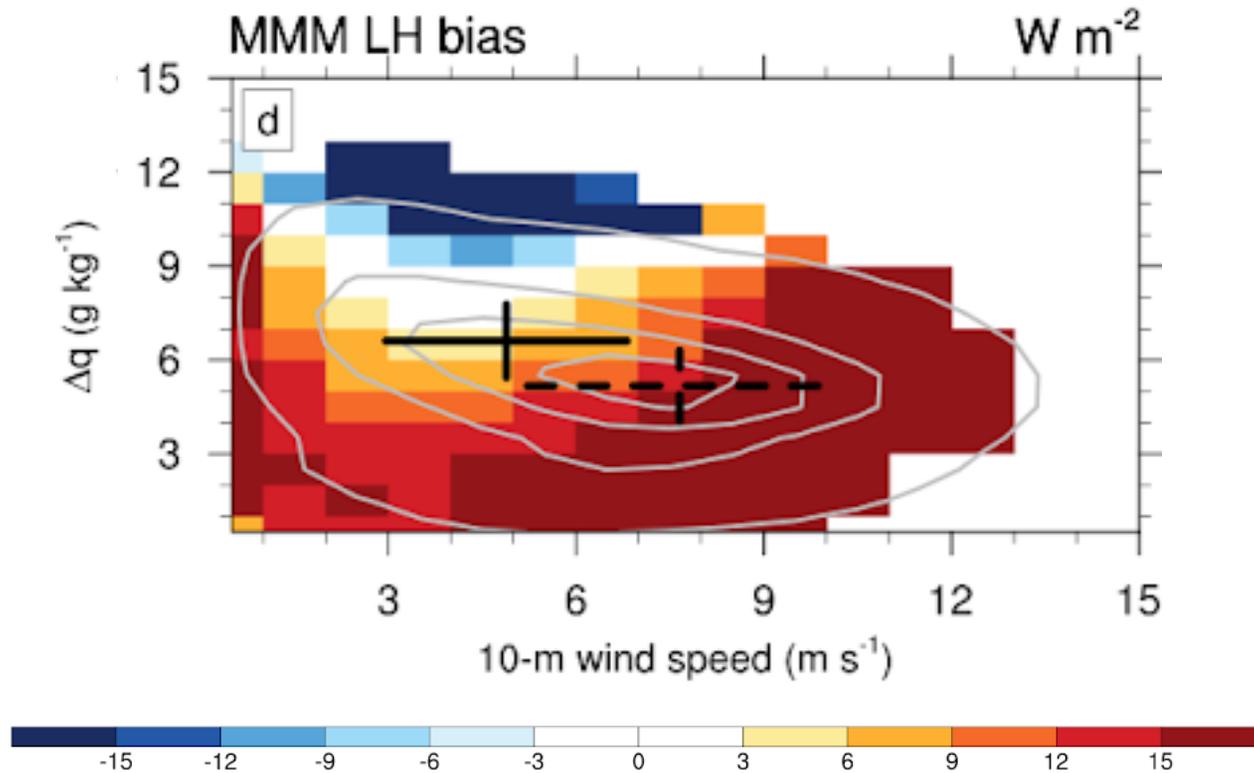
McPhaden (2008)



NOAA

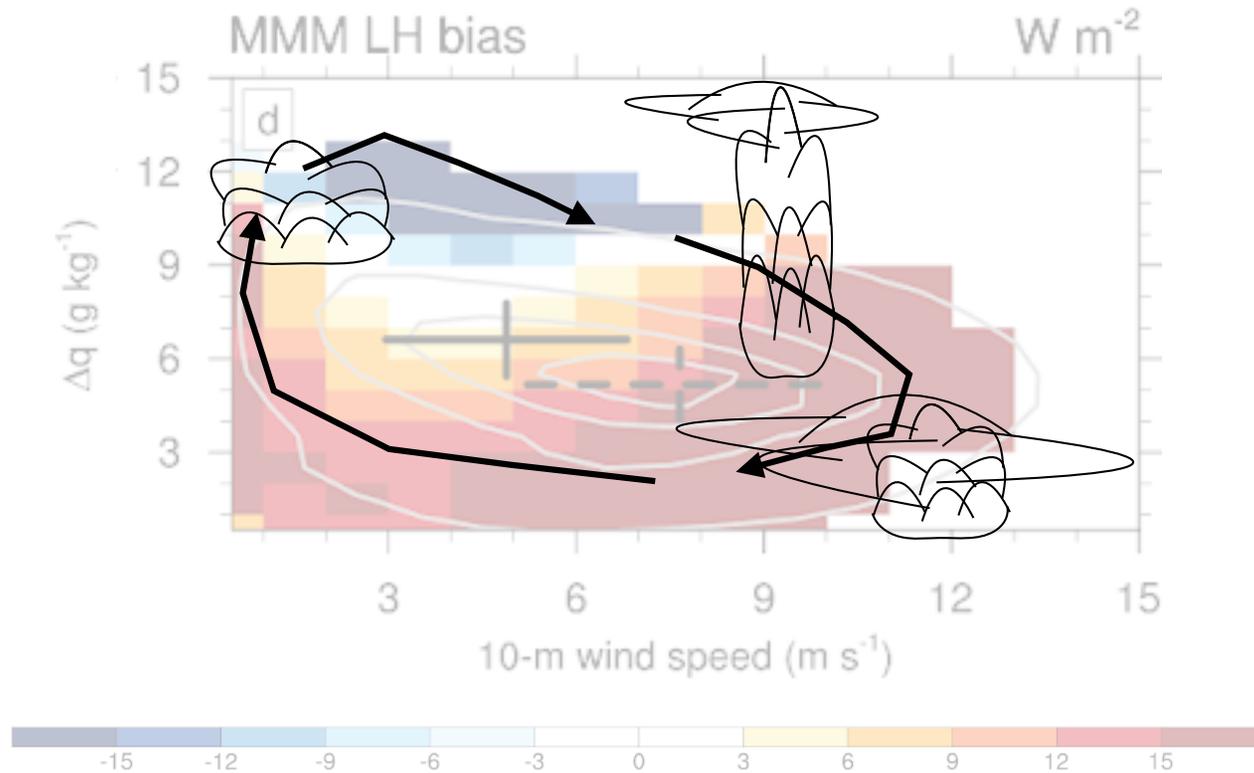
- the “flux matrix” diagnostic illustrates where model fluxes are most biased as a function of wind speed and humidity disequilibrium.

surface fluxes diagnostic implications



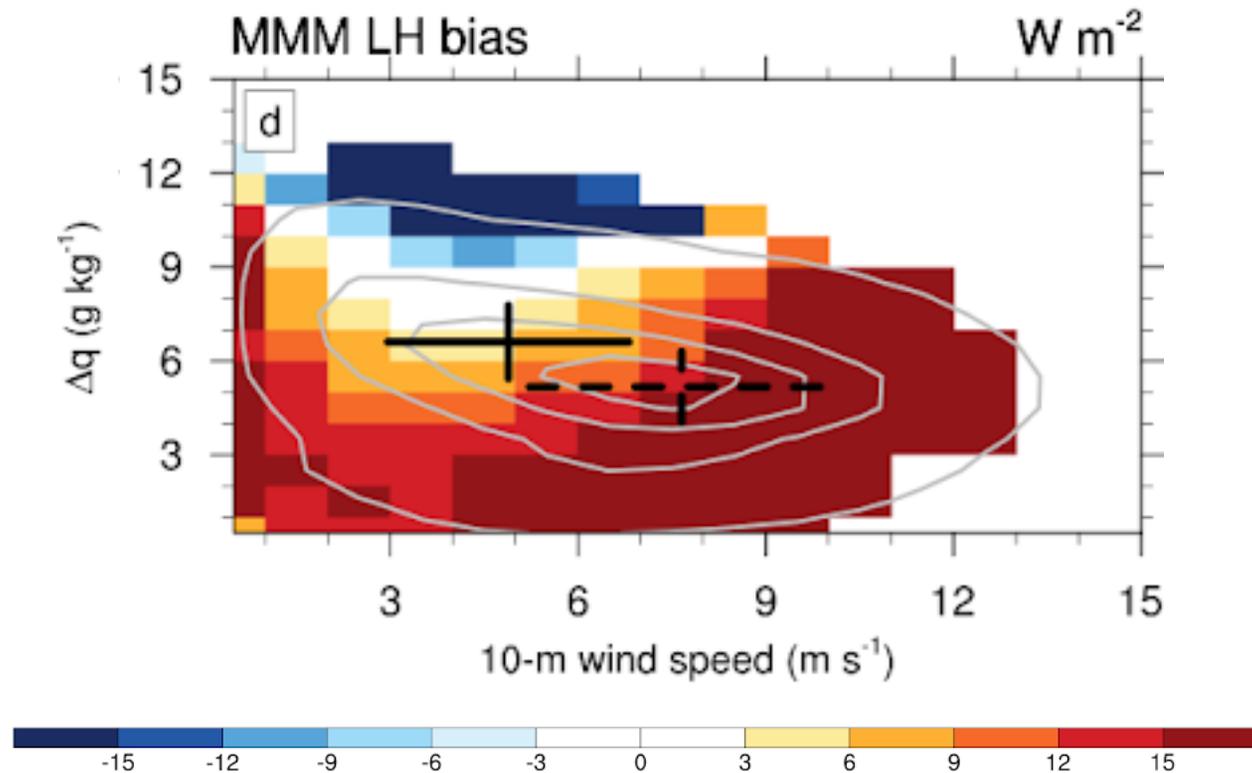
- flux biases are not uniform throughout the convective lifecycle!
 - fluxes too large for mature convection
 - fluxes too small for shallow and transitioning convection

surface fluxes diagnostic implications



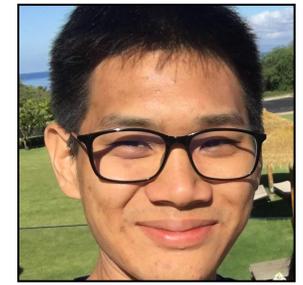
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how to assess bulk algorithm impacts

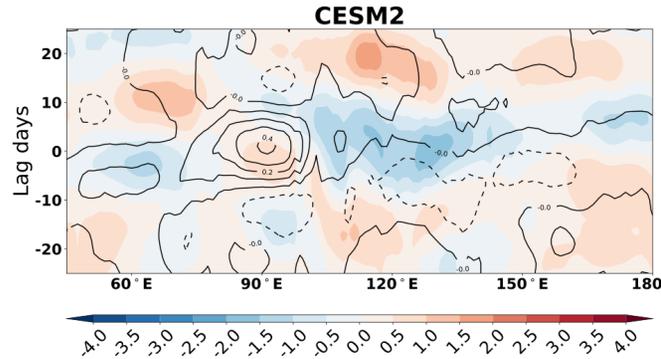
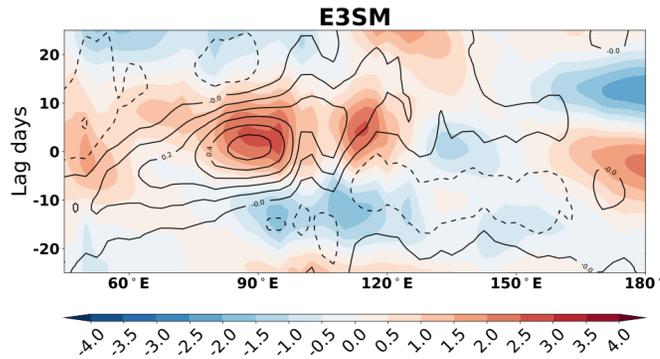


- “offline” correction
 - use flux matrix to generate a “corrected” flux time series
 - no feedbacks to ocean or atmosphere
- “inline correction”
 - replace a model’s default flux algorithm with the COARE flux algorithm

flux corrections for the MJO

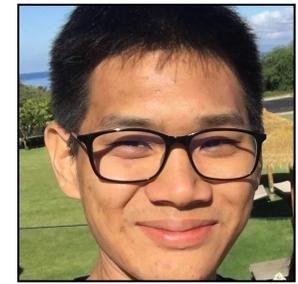


Chia-Wei Hsu (NOAA/PSL)



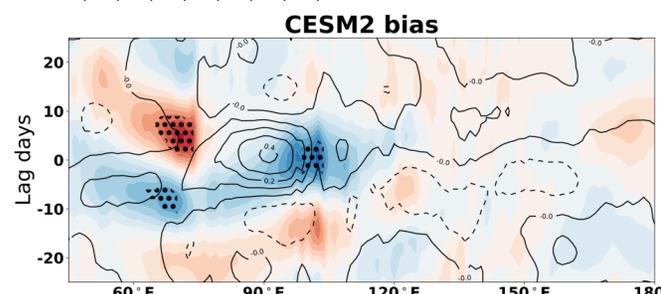
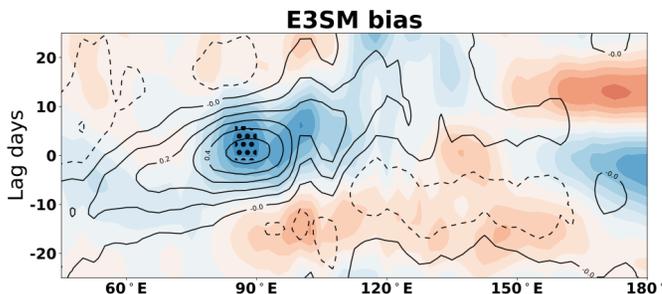
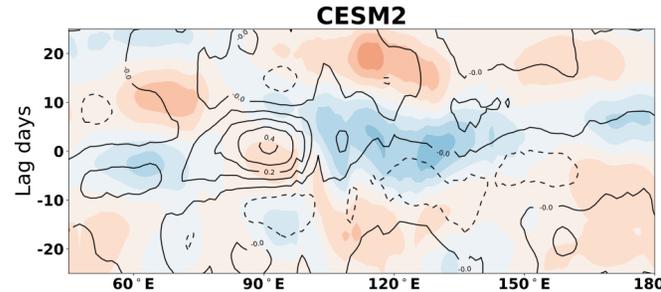
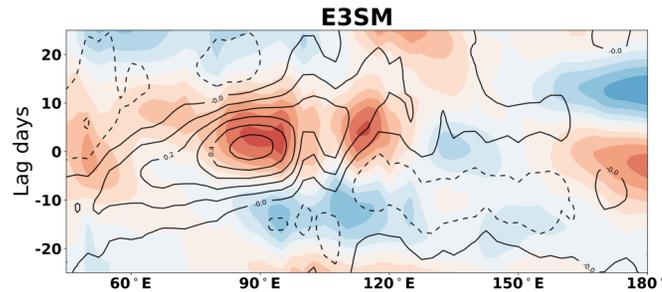
MJO rainfall (contour) and
surface latent heat flux (shading)

flux corrections for the MJO



Chia-Wei Hsu (NOAA/PSL)

MJO rainfall (contour) and surface latent heat flux (shading)



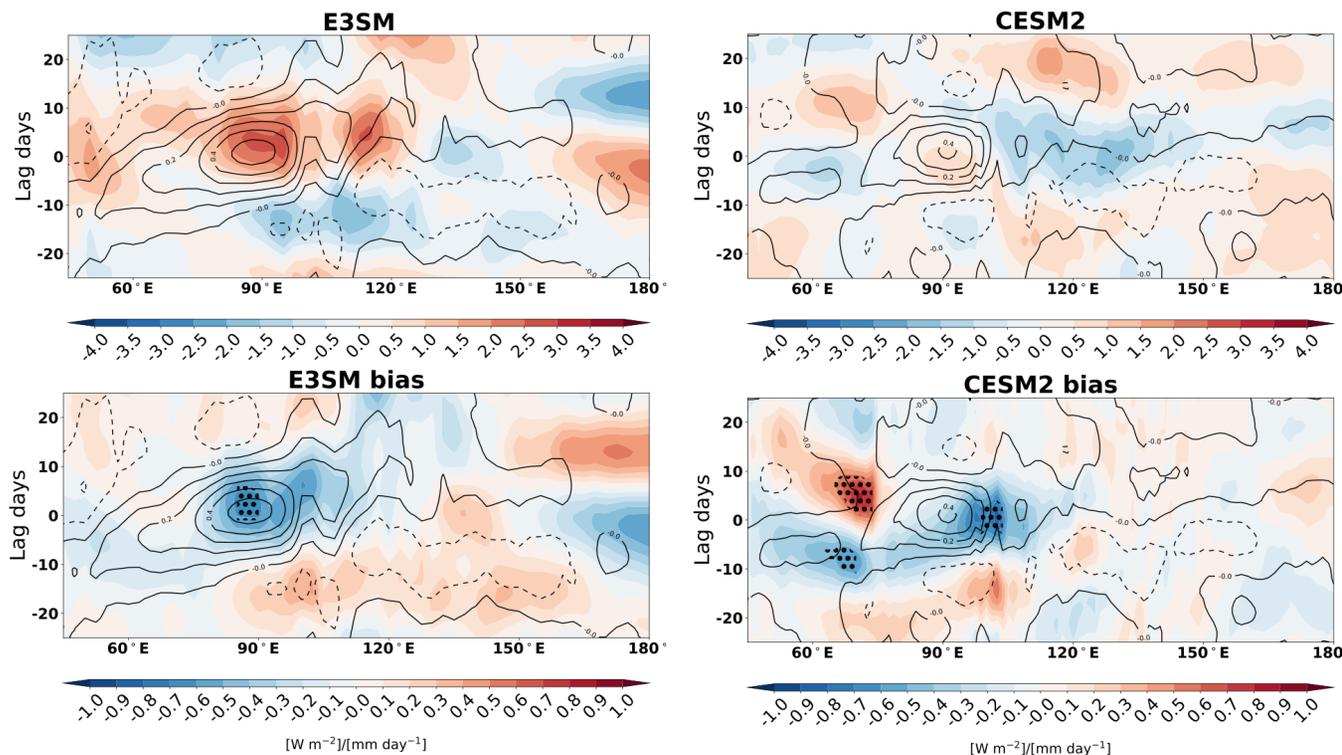
offline correction: default flux algorithm is overly supportive of MJO convection; too weakly supportive of MJO propagation

flux corrections for the MJO



Chia-Wei Hsu (NOAA/PSL)

MJO rainfall (contour) and surface latent heat flux (shading)



offline correction: default flux algorithm is overly supportive of MJO convection; too weakly supportive of MJO propagation

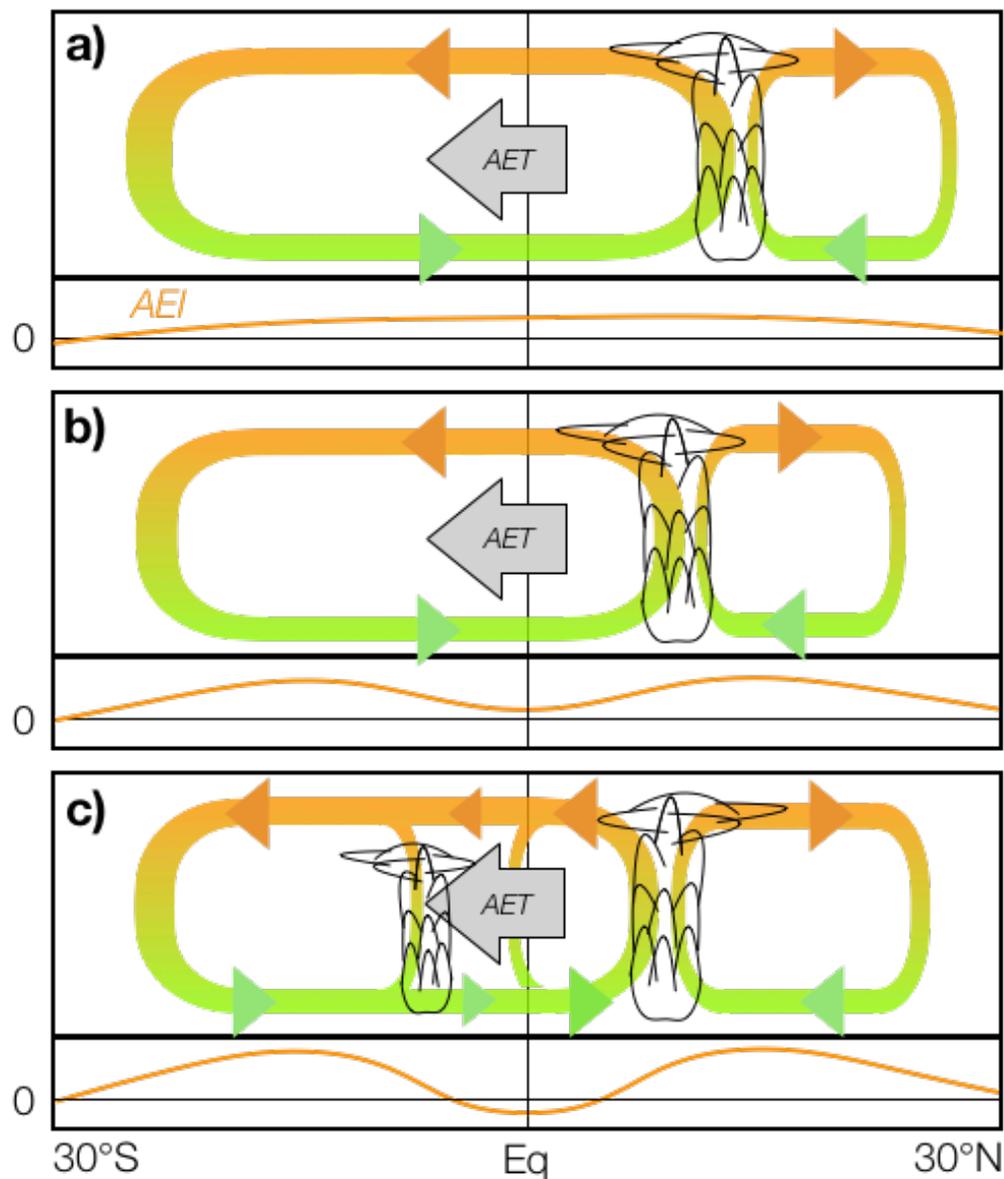
Model/OBS	CTRL	COARE	COARE-CTRL
E3SM_climo	1.01	1.08	0.07
CESM2_amip	0.95	1.10	0.15
TRMM ^a	2.3		

inline correction: COARE fluxes improve MJO eastward propagation

bulk surface flux effects for weather & climate simulation

Example: the ITCZ

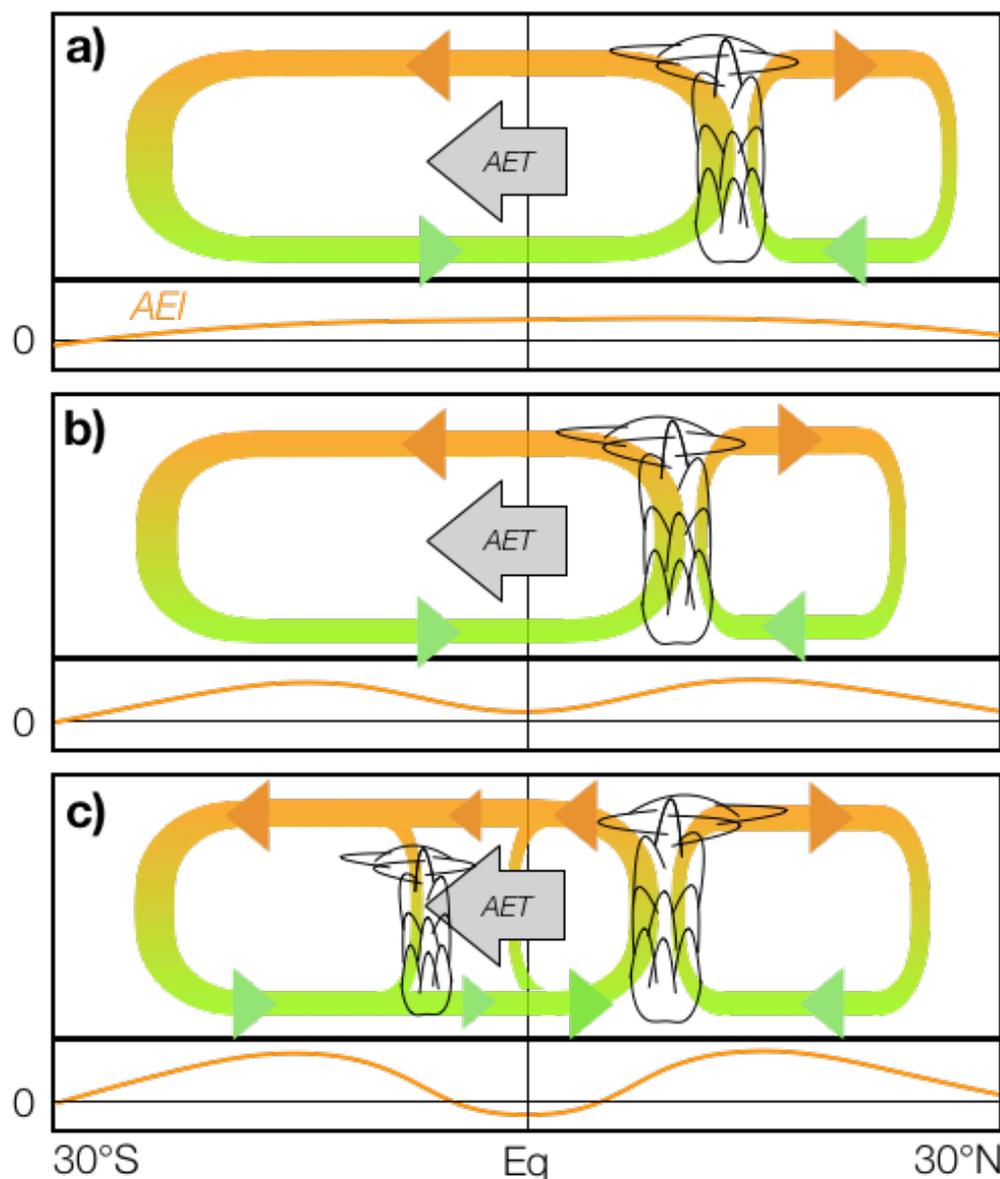
- ITCZ position set by
 - **AET** (Atmospheric Energy Transport) zonal mean across Equator
 - **AEI** (Atmospheric Energy Input) meridional structure
- $AEI = \langle SW \rangle + \langle LW \rangle - OHC$
 - the residual of net shortwave heating, net longwave cooling, and ocean heat uptake (**OHU**)
- **OHU** depends in part on **surface fluxes**



bulk surface flux effects for weather & climate simulation

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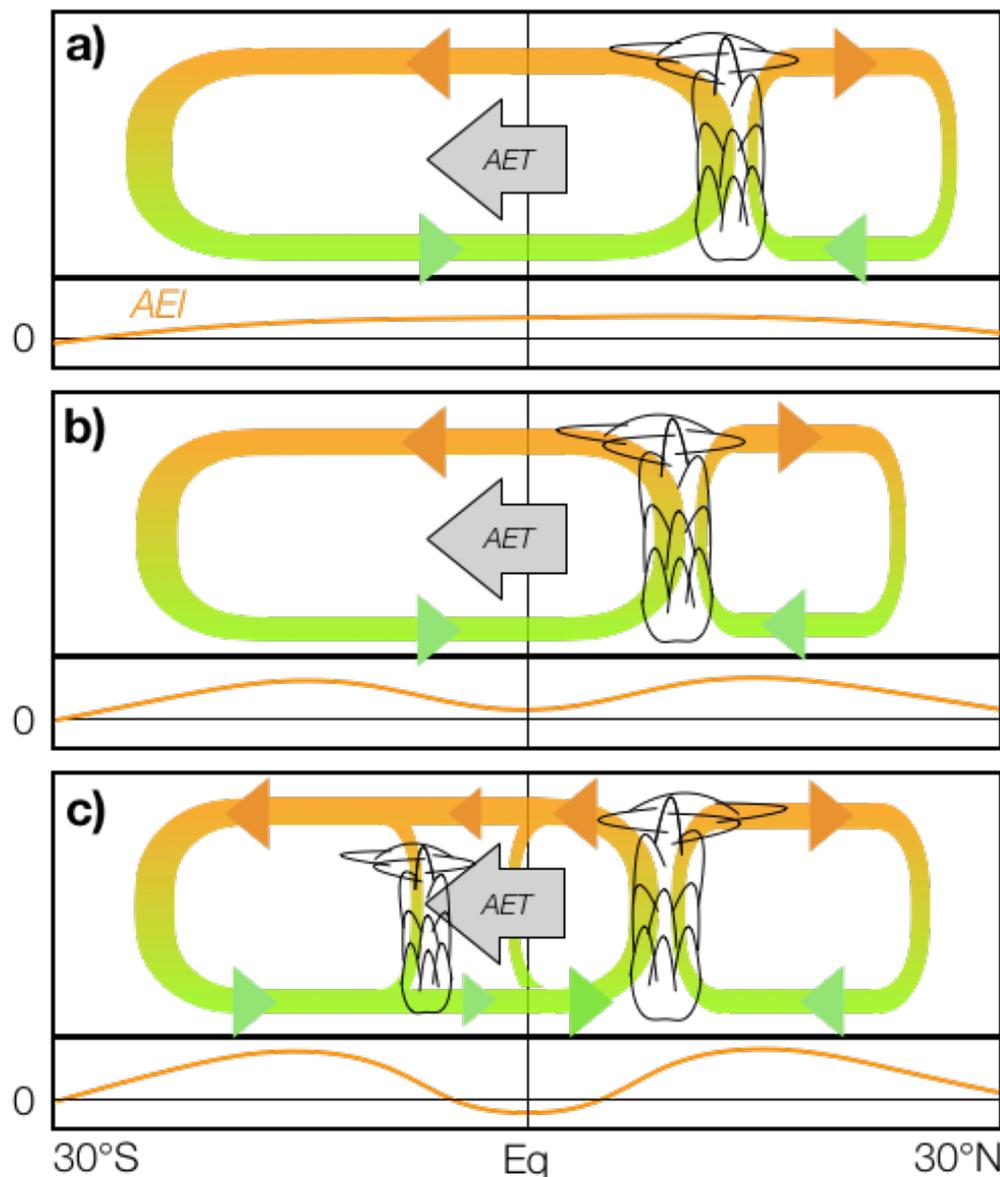
schematic interpretation of Bischoff and Schneider (2014, 2016)

bulk surface flux effects for weather & climate simulation

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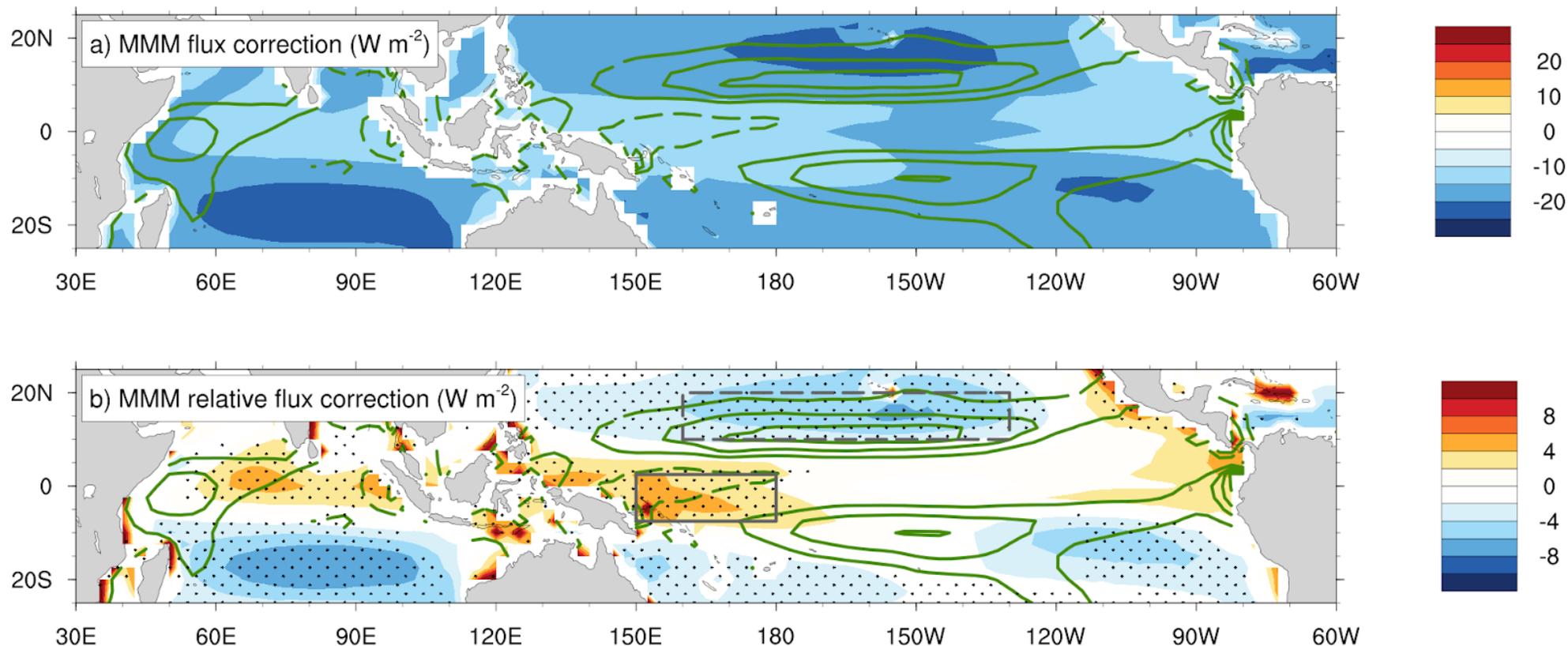
How might biases rooted in surface flux algorithms contribute to model ITCZ biases?



schematic interpretation of Bischoff and Schneider (2014, 2016)

bulk surface flux effects for weather & climate simulation

Example: the ITCZ

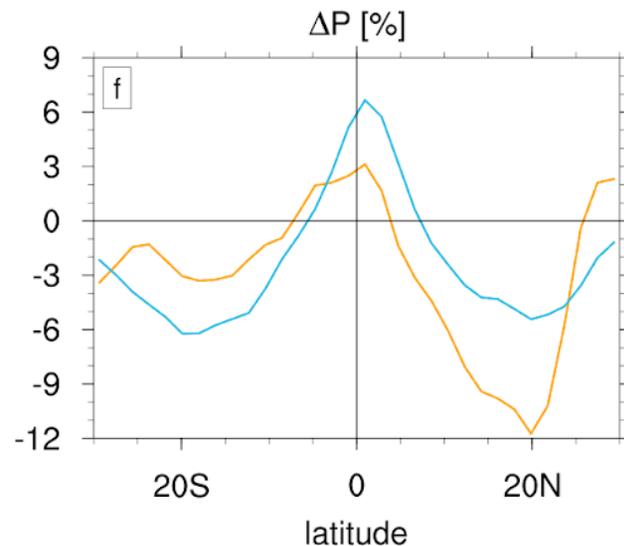
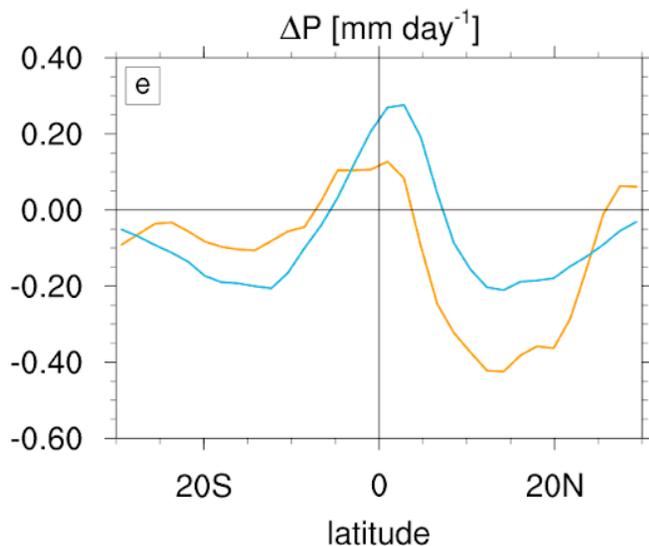
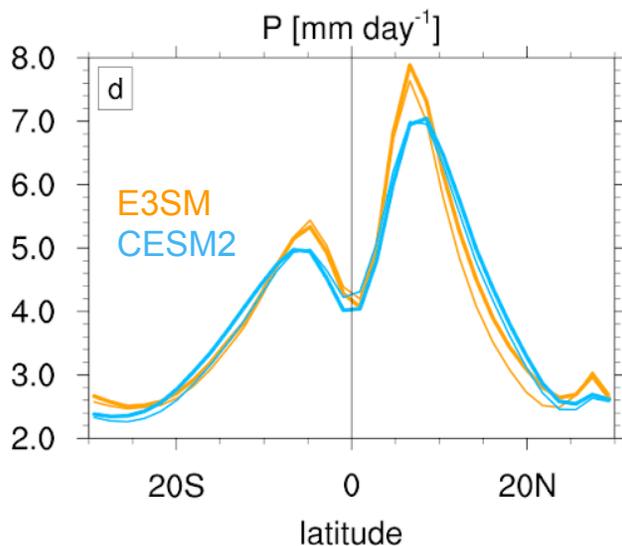


DeMott et al. (2022; in prep)

- “offline” assessment indicates that surface flux algorithm biases may contribute to the double ITCZ bias in climate models
- what about “inline” surface flux corrections?

bulk surface flux effects for weather & climate simulation

Example: the ITCZ

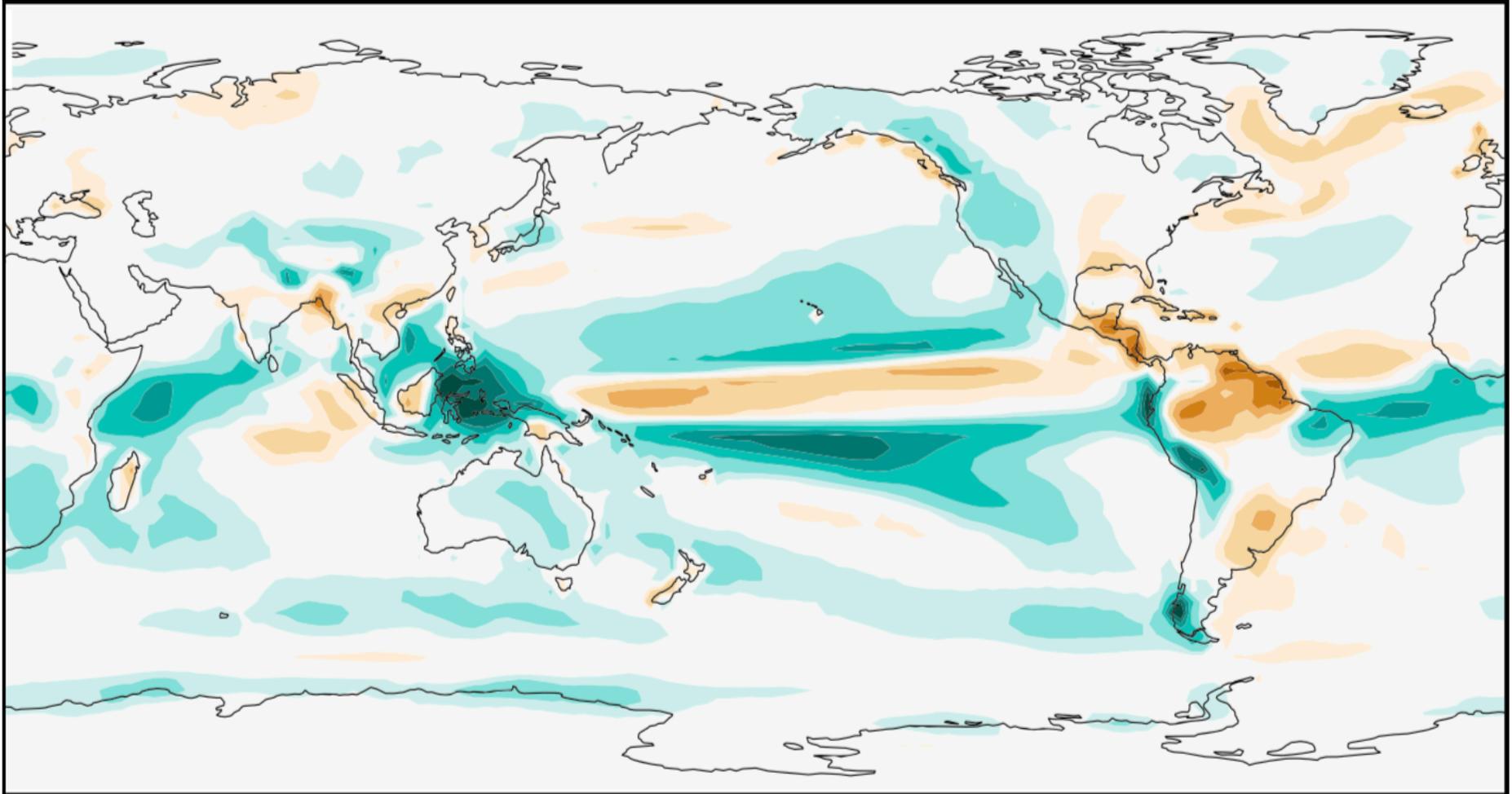


DeMott et al. (2022; in prep)

- “inline” correction in atmosphere-only simulations of two models reduce the double ITCZ bias

E3SMv1 coupled simulations

Example: the ITCZ

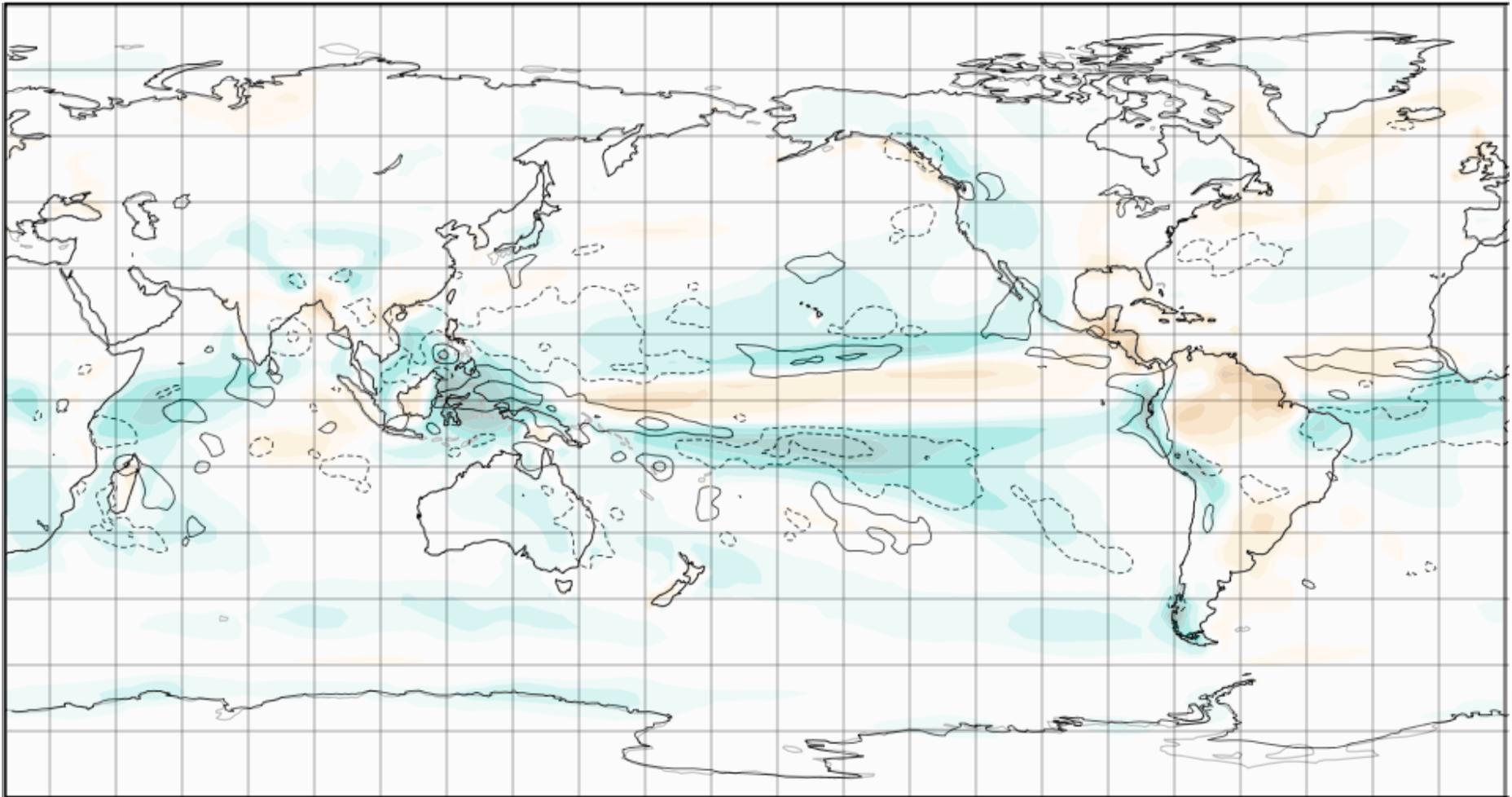


shading = bias

Golaz et al. 2019

E3SMv1 coupled simulations

Example: the ITCZ



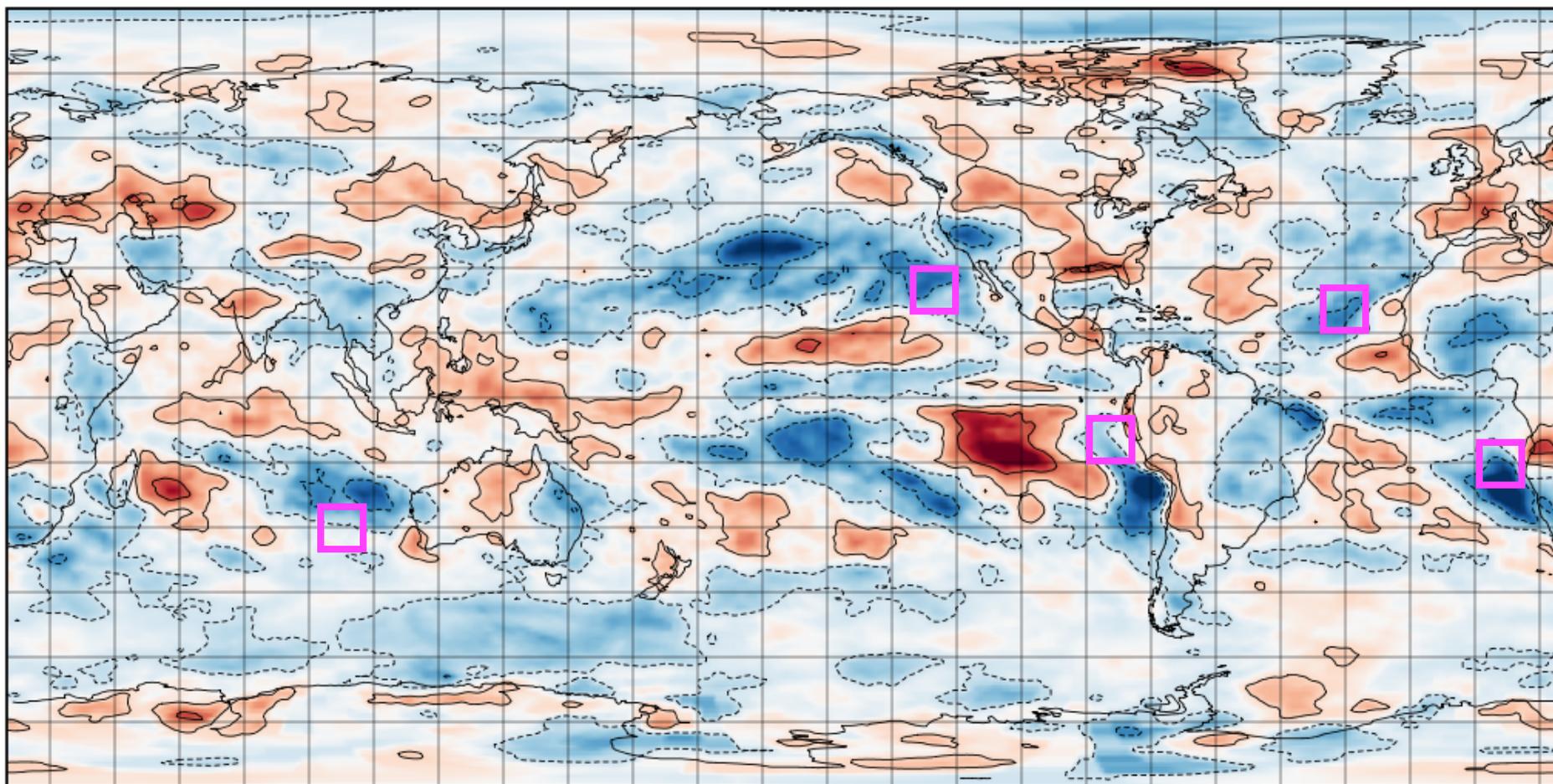
shading = bias

contour = c35-ctl

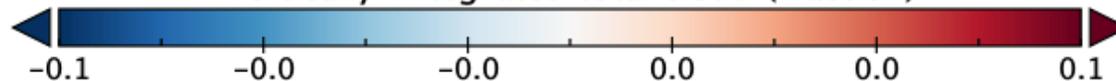
Golaz et al. 2019

E3SMv1 coupled simulations

Example: shallow clouds c35-ctl: Vertically-integrated total cloud



Vertically-integrated total cloud (fraction)



Data Min = -0.1, Max = 0.1, Mean = -0.0

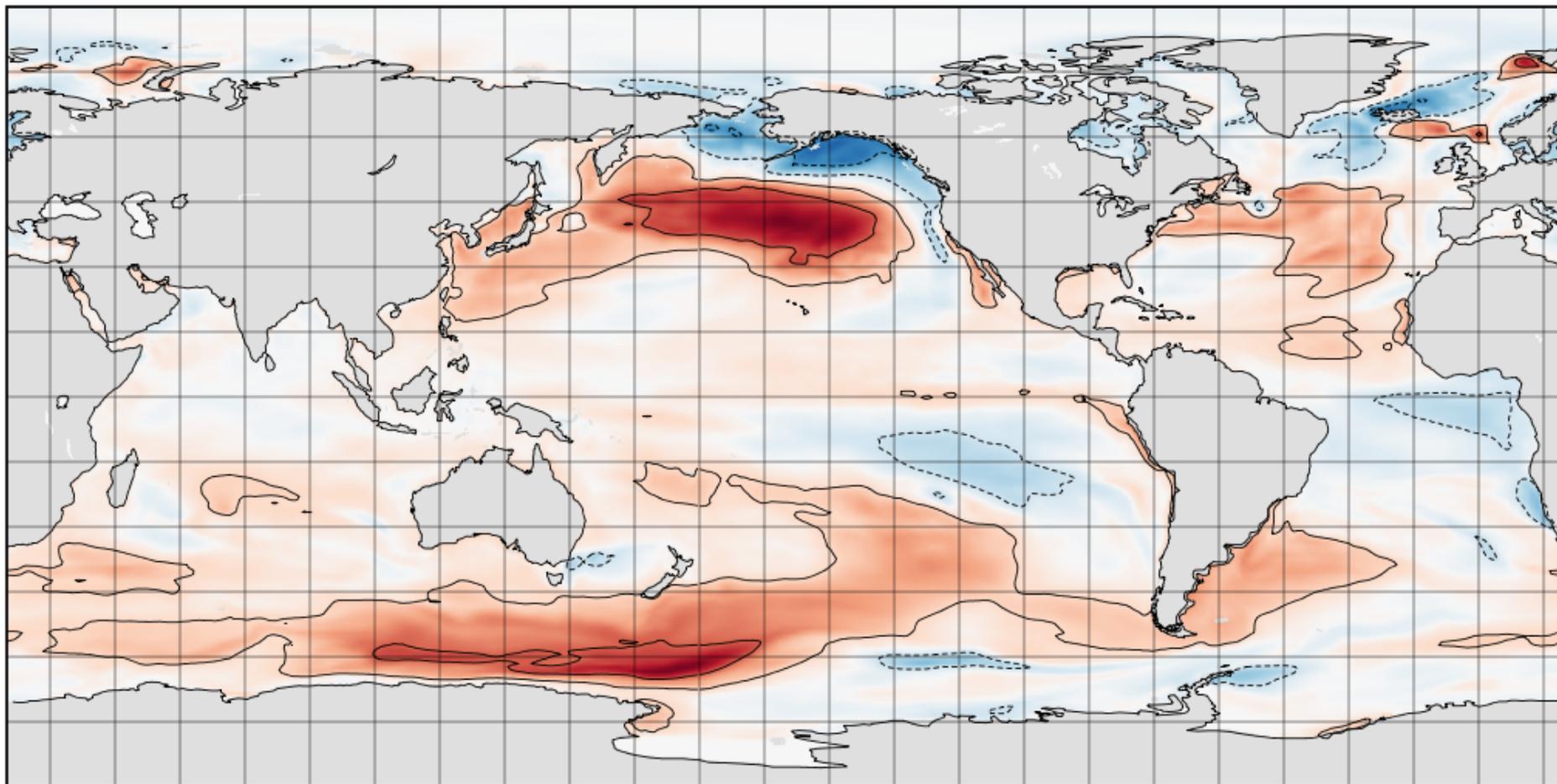
summary

- ocean processes modulate SST
- surface fluxes communicate SST to the atmosphere
- bulk flux algorithms in most climate models overestimate surface fluxes
- surface flux biases are not uniform across convective lifecycles
 - erroneous ocean feedbacks to convective development, teleconnections
- offline and inline corrections to bulk fluxes indicate
 - improved MJO
 - reduced double ITCZ bias
 - more analysis is needed!

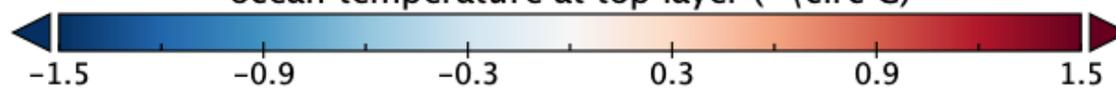
extra slides

E3SMv1 coupled simulations

c35-ctl: ocean temperature at top layer



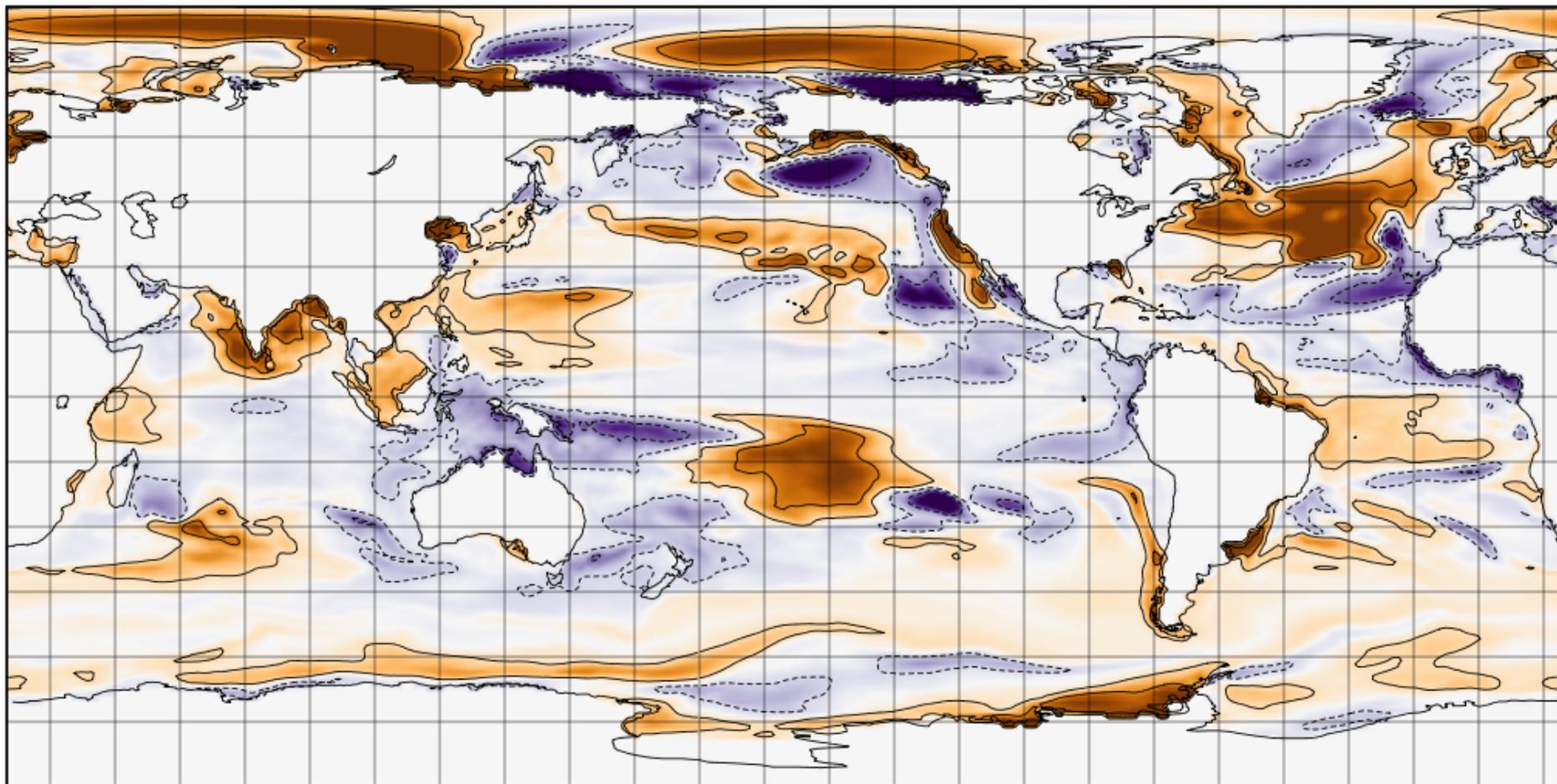
ocean temperature at top layer ($^{\circ}\text{C}$)



Data Min = -1.2, Max = 1.5, Mean = 0.1

E3SMv1 coupled simulations

c35-ctl: salinity at top layer



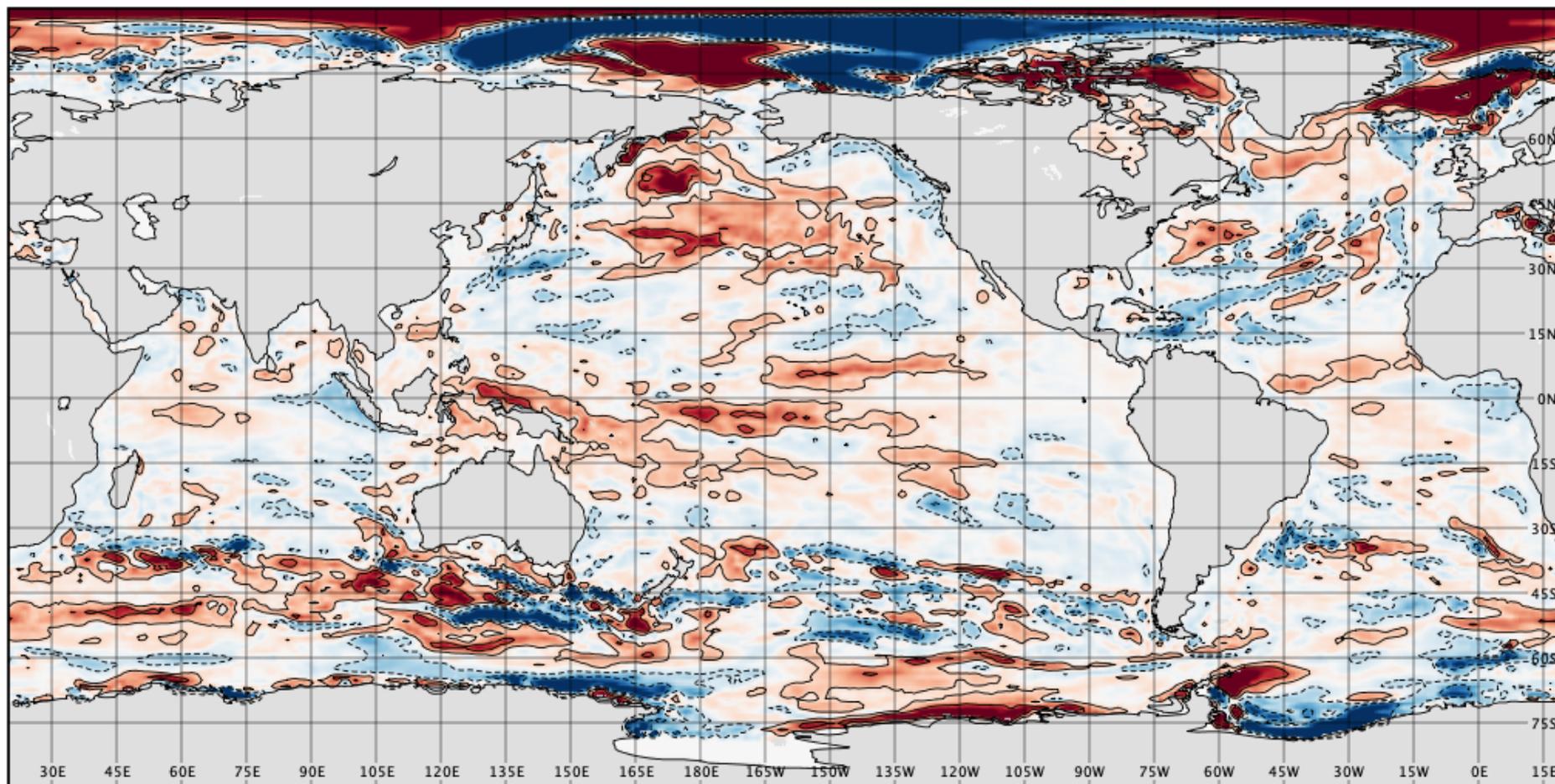
salinity at top layer (PSU)



Data Min = -1.5, Max = 1.5, Mean = 0.0

E3SMv1 coupled simulations

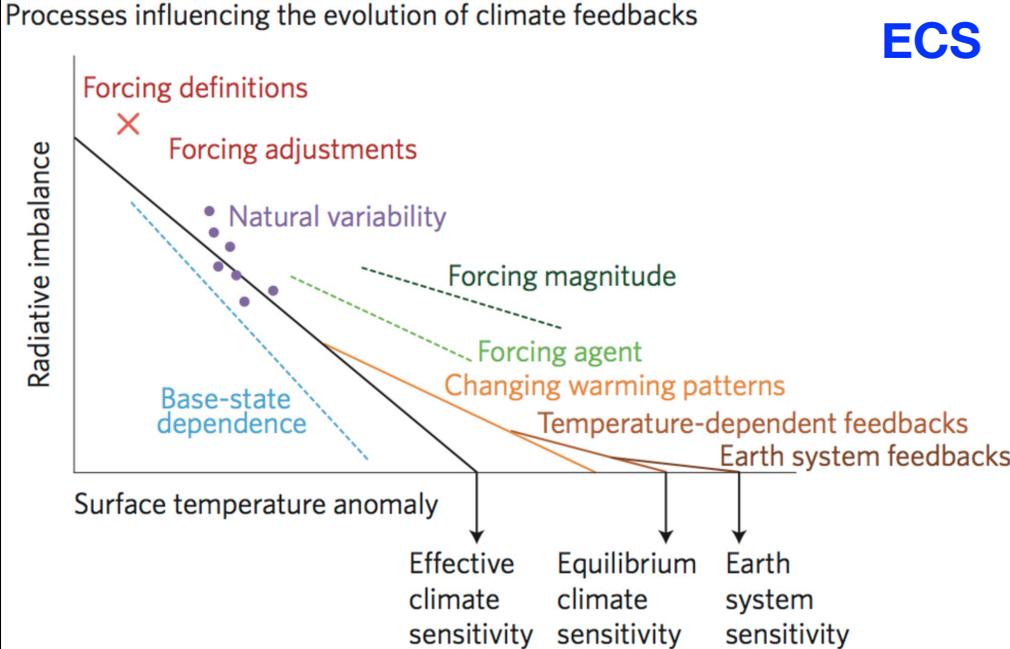
c35-ctl: barrier layer thickness (threshold method)



barrier layer thickness (threshold method) (m)

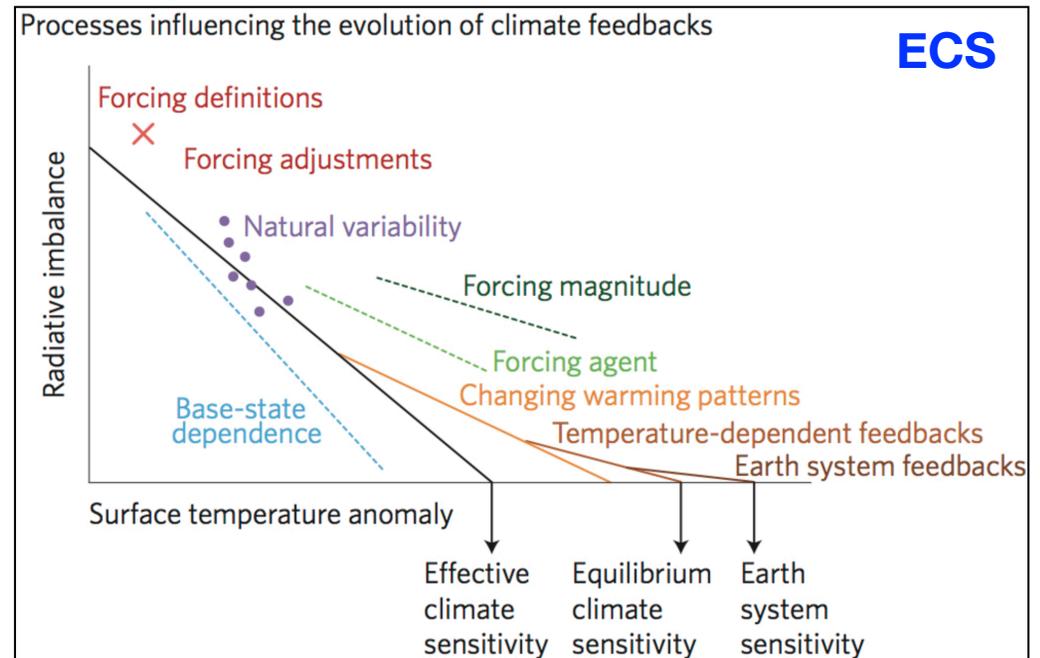


Data Min = -25.8, Max = 30.2, Mean = 0.1



Knutti et al. (2017)

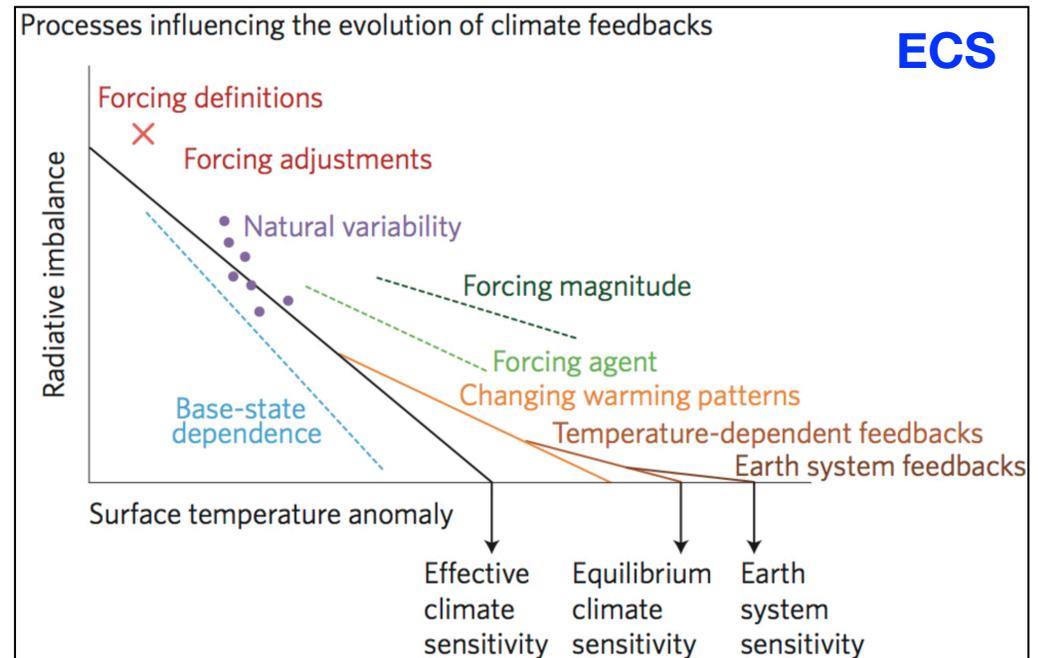
- how does the inline correction affect MJO-ENSO coupling?
- how does the inline correction affect ECS?



Knutti et al. (2017)

ongoing work

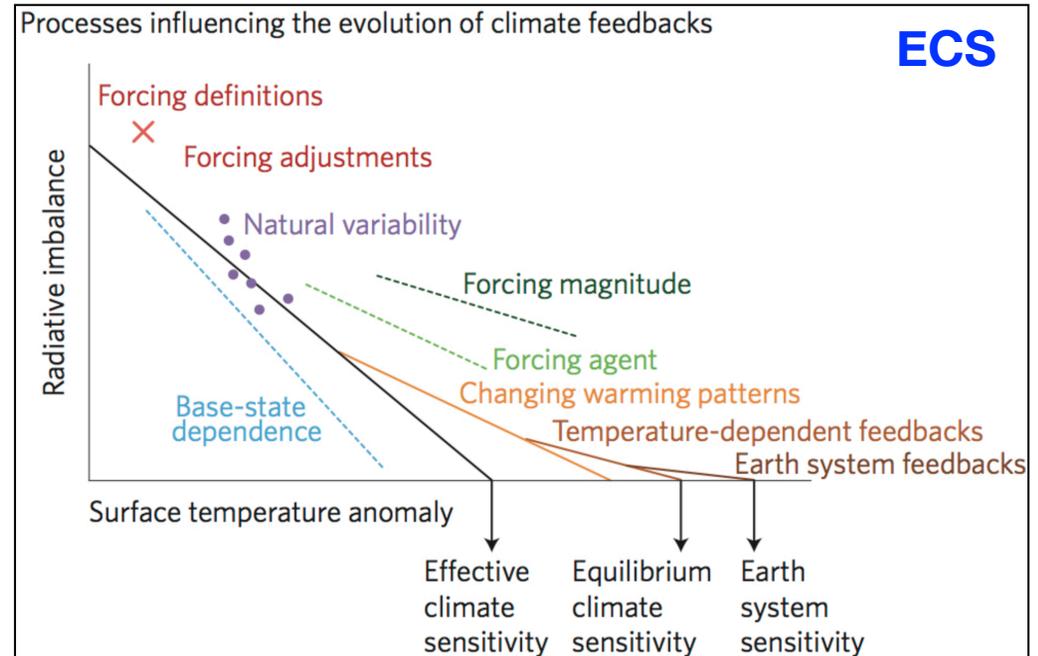
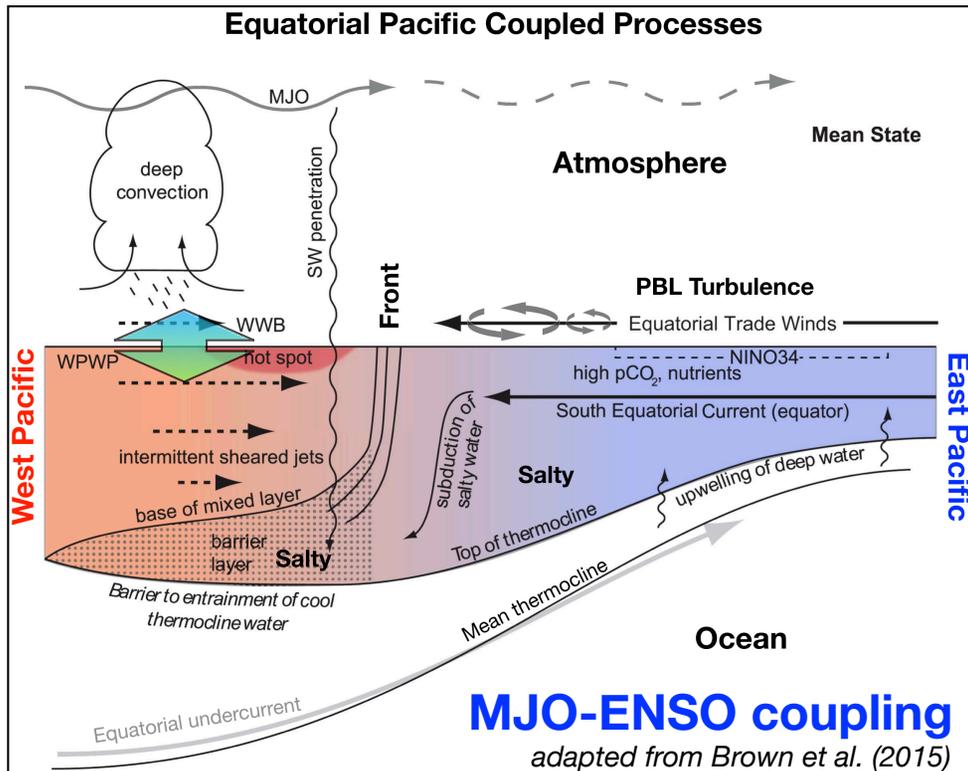
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Knutti et al. (2017)

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Knutti et al. (2017)

surface fluxes: what can affect their parameterization?

- absolute or relative wind speed
- wind gustiness
- stability of the boundary layer
- surface roughness
- surface salinity

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“many aspects of the algorithms are empirical, relying on constants and functional forms estimated from (a relatively small number of) ship- and buoy-based observational campaigns”
—Reeves Eyre et al. (2021)

pathways for future progress

- Needs:
 - Improving CONUS precipitation forecasts on S2S+ timescales relies (in part) on improving the representation of tropical O-A coupling
 - Coupled forecast models need to be initialized with atmospheric and oceanic states that are well-constrained by observations
 - An increasing recognition that the OML and AML must be thought of and observed as a single entity: the Air Sea Transition Zone (ASTZ).
- Needed observations
 - Important, but hard-to-observe ASTZ processes require targeted field programs.
 - Improving understanding and model representation of coupled cross-scale interactions requires sustained, detailed measurements of the ASTZ at multiple locations.