



# The use of satellite data-based "critical relative humidity" in cloud parameterization and its role in modulating cloud feedback

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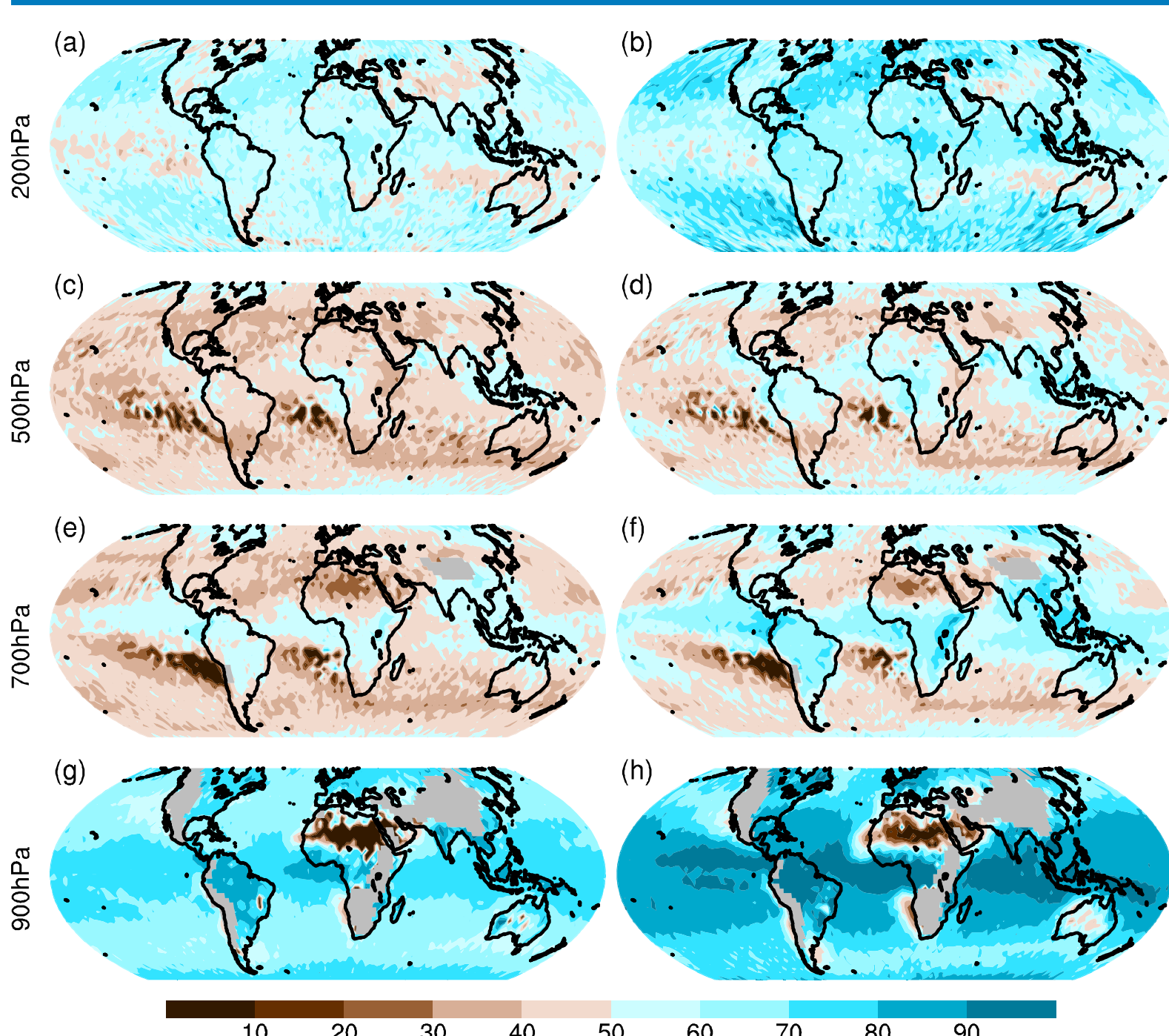
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## 1. Introduction

The critical relative humidity (RH<sub>c</sub>), which approximately measures the subgrid-scale variability of moisture, is important to cloud parameterization. Based on the diagnostics from CloudSat/CALIPSO satellite data, we propose an improved RH<sub>c</sub> formula that incorporates geographic dependence and allows for non-monotonic variations in the vertical direction. With the parameterized RH<sub>c</sub>, a cloud macrophysics scheme is constructed in which fractional cloudiness and subgrid-scale condensation are synergistically solved, with the latter being calculated using two different approaches. Cloud feedback is analyzed under the new scheme.

## 2. Diagnosis of RH<sub>c</sub>



$$RH_c = 1 - \frac{1 - RH}{(1 - C)^2}$$

RH: grid-mean relative humidity  
C: cloud fraction

- Lower values in subtropics
- higher ones in inner tropics

**Fig. 1.** Geographical distribution of CloudSat/CALIPSO diagnostic RH<sub>c</sub> (%) at selected pressure levels using the temporal average (left) and least-squares method (right).

## 3. Parameterization of RH<sub>c</sub>

In each region, RH<sub>c</sub> has the general form:

$$RH_c = \beta_1 + \beta_2 \times \exp \left[ 1 - \left( \frac{P}{P_s} \right)^{\beta_3} \right] + \beta_4 \times \exp \left[ 1 - \left( \frac{P_s - P}{P_s} \right)^{\beta_5} \right]$$

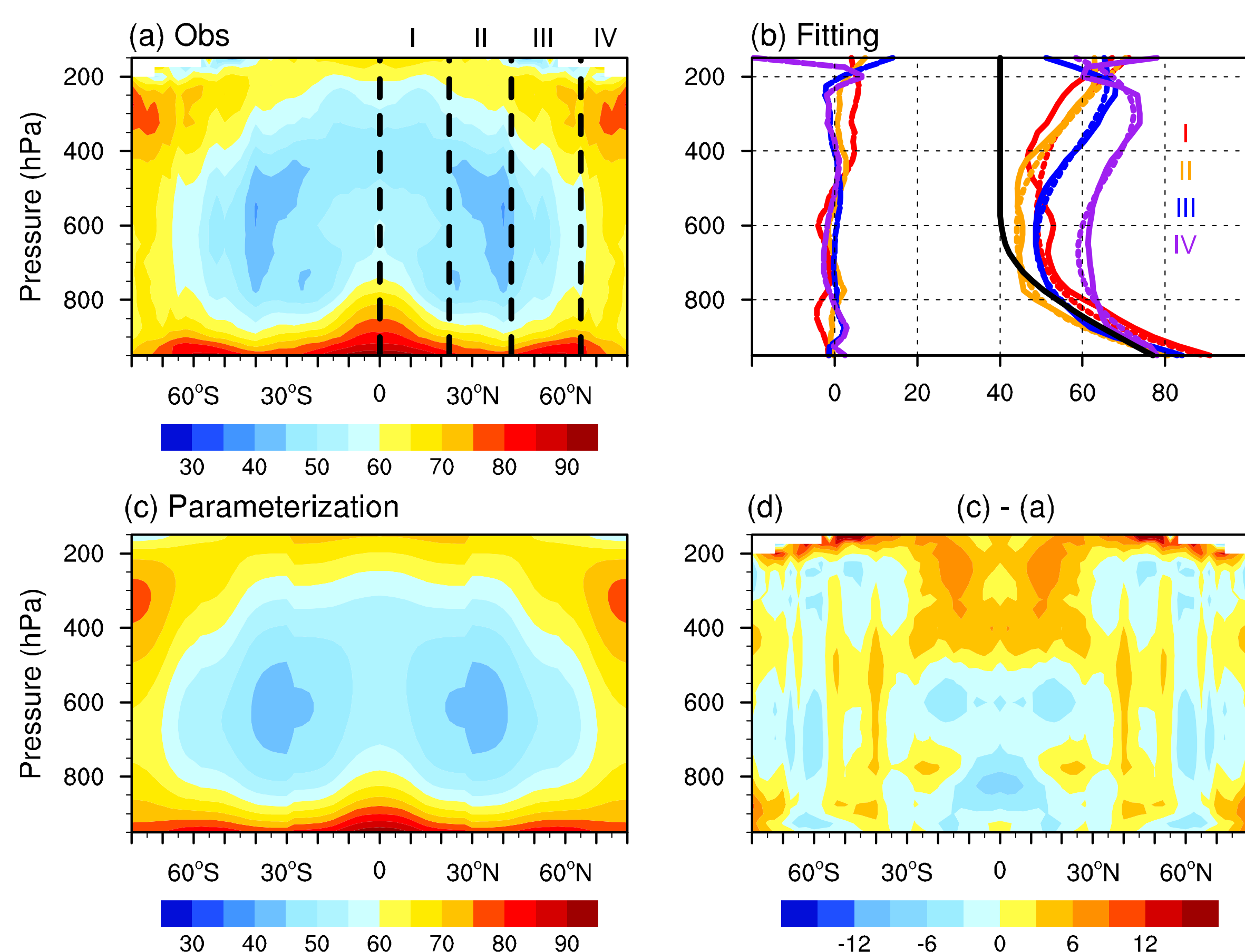
The coefficients β<sub>j</sub> are determined by minimizing the cost function, as follows:

$$g = \sqrt{\frac{1}{N} \sum \left( \beta_1 + \beta_2 \times \exp \left[ 1 - \left( \frac{P}{P_s} \right)^{\beta_3} \right] + \beta_4 \times \exp \left[ 1 - \left( \frac{P_s - P}{P_s} \right)^{\beta_5} \right] - RH_c^{obs} \right)^2}$$

The separated RH<sub>c</sub> is then combined into a single latitude-dependent formula as:

$$RH_c(\phi) = \frac{1 - \alpha_{idx(\phi)} - 1}{2} RH_c(idx(\phi) - 1) + \frac{1 + \alpha_{idx(\phi)} - 1}{2} RH_c(idx(\phi))$$

where φ stands for latitude, and "idx" is the index of the region corresponding to latitude φ. α<sub>idx(φ)</sub> are weighing coefficients satisfying α<sub>idx(φ)</sub> = tanh((φ - φ<sub>0</sub>)/D<sub>0</sub>), where φ<sub>0</sub> and D<sub>0</sub> are tunable parameters



**Fig. 2.** (a) Observed RH<sub>c</sub> (%) after symmetrization in the Northern Hemisphere and Southern Hemisphere. (b) Observed (dashed) and parameterized (solid) RH<sub>c</sub> profiles in four representative regions, with the differences marked by long-dashed lines. The parameterized RH<sub>c</sub> of Quaas (2012) is also superimposed in the figure (solid black). (c, d) Latitude–pressure cross-section of (c) the new parameterized RH<sub>c</sub> and (d) its deviation against the observation.

### Reference:

X Wang, et al. The use of satellite data-based "critical relative humidity" in cloud parameterization and its role in modulating cloud feedback, *Journal of Advances in Modeling Earth Systems*, 2022, 14(10): e2022MS003213

## 4. Calculation of subgrid condensation

### ✓ Prognostic method ( \_prog )

This method retains the property of cloud condensate as a prognostic variable in models. The prognostic equation for relative humidity  $U$  is expressed in terms of total water  $q_t$ , liquid water temperature  $T_l$ , and liquid water  $q_l$ :

$$\frac{\partial U}{\partial t} = \alpha \frac{\partial q_t}{\partial t} - \beta \frac{\partial T_l}{\partial t} - \gamma \frac{\partial q_l}{\partial t}$$

Noting  $\frac{\partial U}{\partial t} = 0$  in the cloudy portion, together with the use of  $\frac{\partial q_l}{\partial t} = C \frac{\partial \hat{q}_l}{\partial t} + c_m \hat{q}_l \frac{\partial C}{\partial t}$ , equation (8) is further expanded in the following form:

$$\left( 1 + c \frac{L}{C_p} \frac{\partial q_s}{\partial T} \right) \frac{\partial \hat{q}_l}{\partial t} + c_m \hat{q}_l \frac{L}{C_p} \frac{\partial q_s}{\partial T} \frac{\partial C}{\partial t} = \frac{\partial \hat{q}_l}{\partial t} - \frac{\partial q_s}{\partial T} \frac{\partial T_l}{\partial t}$$

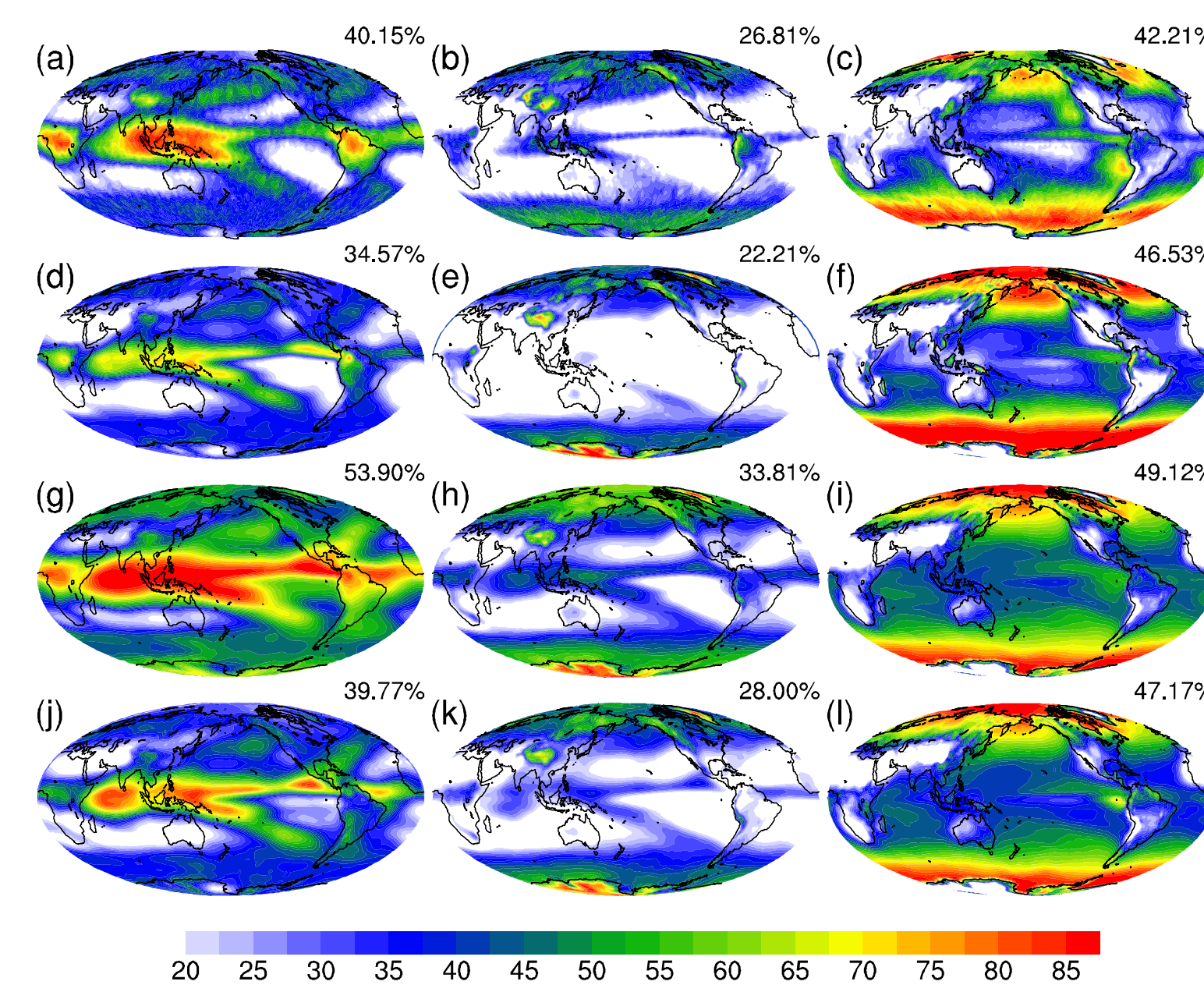
### ✓ Diagnostic method based on PDF ( \_pdf )

With a uniform PDF assumed for total water, one can also obtain a diagnostic formula for cloud condensate,

$$q_l = \int_{q_s}^{q_t + \Delta q} (q - q_s) \frac{1}{2\Delta q} dq$$

which is further expanded as:  $q_l = \frac{(q_t + \Delta q)^2 - q_s^2}{4\Delta q} - \frac{1}{2\Delta q} (q_t + \Delta q - q_s) \cdot q_s$

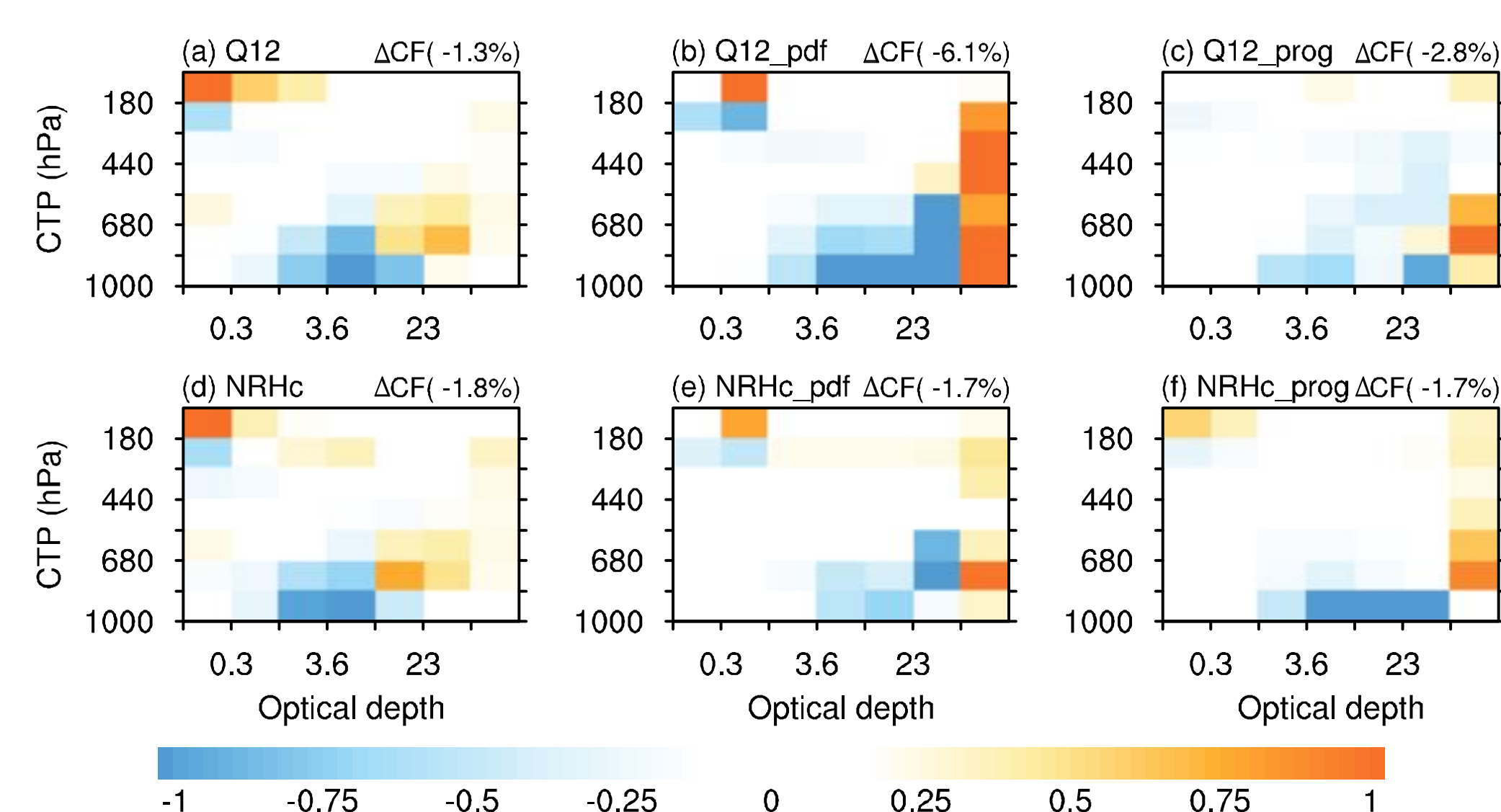
## 5. Performance on AMIP simulation



- XR96 and Q12 underestimates clouds
- NRHc scheme performs better

**Fig. 3.** Geographical distribution of high (left column), mid (middle column) and low (right column) cloud cover (%) from (a–c) CloudSat/CALIPSO and simulations using the cloud scheme of (d–f) XR96, (g–i) Q12, and (j–l) NRH<sub>c</sub>. The global mean value is shown in the top right of each figure.

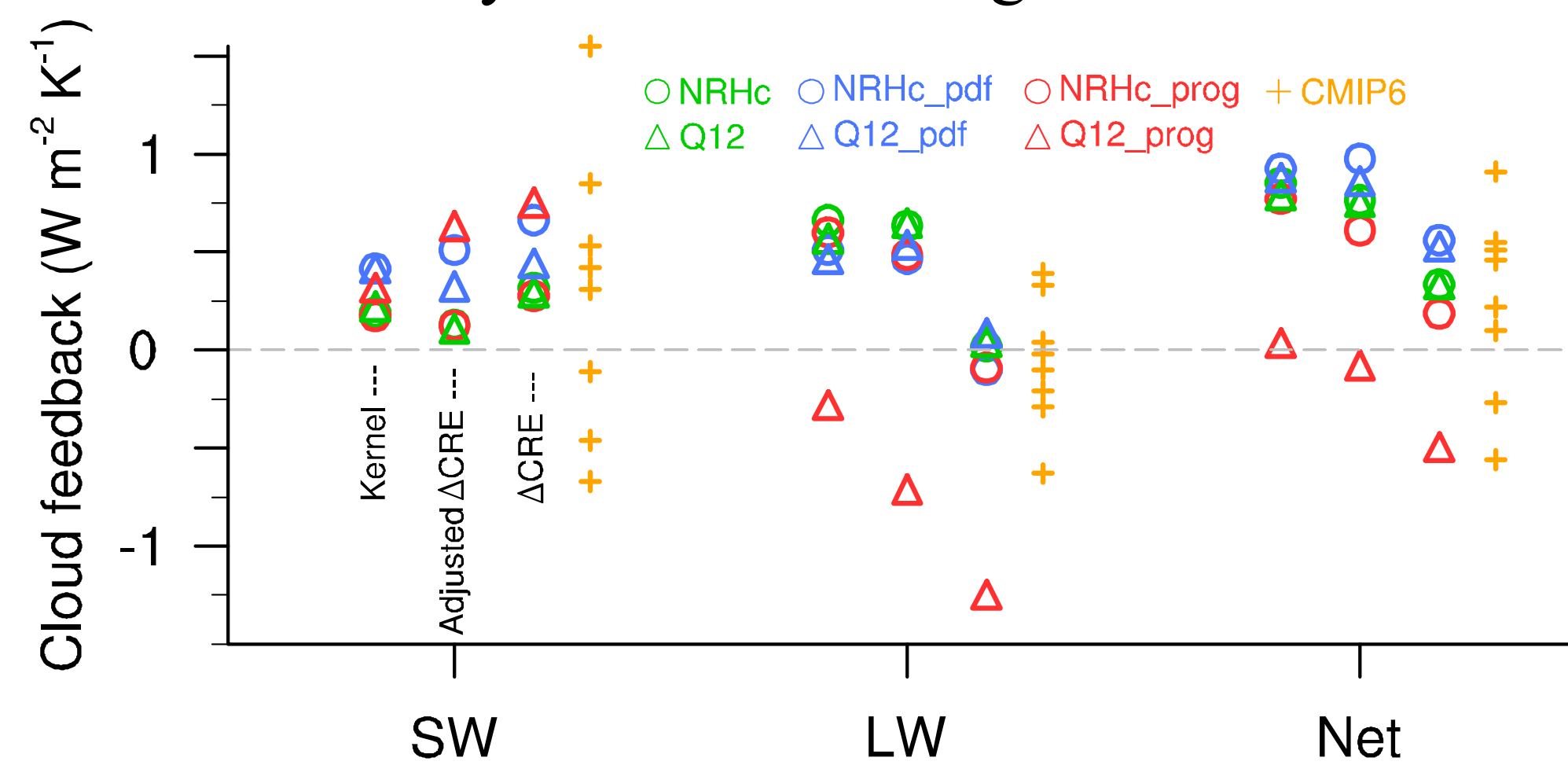
## 6. Cloud feedback analysis



- All simulations show a reduction in globally-averaged cloudiness
- The sensitivity to RH<sub>c</sub> is smaller in simulations applying RH<sub>c</sub> in cloud cover alone

**Fig. 4.** Globally averaged changes in cloud fraction in CTP-τ histograms between the +4 K and control experiments in simulations with different RH<sub>c</sub> configurations. The sum of each matrix is shown in the top right of each panel.

- Large discrepancies of cloud feedbacks are found in a single model
- The way to calculate subgrid condensation and the choice of RH<sub>c</sub> are both important



**Fig. 5.** Global- and annual-mean cloud feedbacks estimated by three different methods in simulations with different RH<sub>c</sub> configurations. The multimodel cloud feedbacks estimated by ΔCRE from the CMIP6 APE experiments are also overlaid, with each model represented by a plus sign.

## 7. Conclusion

- We propose an improved RH<sub>c</sub> formula that incorporates geographic dependence and allows for non-monotonic variations in the vertical.
- Varying RH<sub>c</sub> and techniques in calculating subgrid condensation replicates the range of uncertainty found in CMIP6 cloud feedback estimates.
- Varying RH<sub>c</sub> and its implementation in models leads to the diversity of cloud feedback being mainly due to optically thick clouds