Analysis of upper tropospheric cloud properties and water vapor variability in relation to the large-scale atmospheric state

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Introduction

Parameterization of the distribution and variability of water vapor. cloud microphysical properties, and cloud fraction within a global climate model (GCM) grid box remains one of the key uncertainties in improving our prediction of future climate. We analyze the distribution and fractional variance of observed cloud scale vertical motion, cirrus microphysical properties, and upper tropospheric water vapor in 12 unique atmospheric states. Within this framework, we also discuss the potential for relating atmospheric states to cirrus formation

(1) Classification of the Large-Scale **Atmospheric State**

Clustering Algorithm

· Clustering algorithm performed on ~ 3 ½ years of Rapid Update Cycle (RUC) output centered over the ARM SGP site

- Extract at each 3 hour time step: Surface Pressure
- Temperature, Humidity, winds (U, V, and W)
- over a 9 x 9 grid where each grid box is roughly equal to a GCM grid box
- at 8 different pressure levels Cluster using a Competitive Neural Network to assign each time step to an atmospheric state

Validation

- · Time series of the vertical profile of cloud fraction provides an independent validation data set · Developed a statistical test to
- not:
- account for the presence of highly correlated data (Wilks 1997; Efron and Tibshirani 1993)
- State Distinctness
- at a final, validated set of states

Frequency distribution of RH as a

function of pressure level for each state

Humidity

Reference: Marchand et al. (2006), J. Atmos. Sci, 63

(2) Unique Atmospheric States

Leve

Cloud fraction as a function of height derived from MMCR reflectivity profiles



determine if two states are distinct or

- · Bootstrap resampling technique to
- Test for two things:
- State Stability
- An iterative process is used to arrive

(3) Cloud Scale Properties

Cloud Properties

 IWC and particle effective size (D_{ge}) for individual cloud systems are retrieved using a lidar-radar algorithm (Wang and Sassen 2002, JAM, 41)

 Lidar/radar algorithm applied to Raman lidar and MMCR data at SGP 1999-2000

 Atmospheric state assigned to cloud observations based on RUC analysis (see (1))

· Mean and fractional variance (f_v) computed for each cirrus event on a given day



properties and atmospheric state on 16 Feb 2000 at SGP

 Bin MMCR Doppler Velocity (V_D) according to layer height and reflectivity

• Average V_p in each height-Z

• Average V_D=Fall Velocity (V_{fall})

• $V_{meso} = V_D - V_{LS} - V_{fall}$ where V_{LS} is the large-scale vertical velocity.

- Require 100 samples in a bin
- Interpolation performed
- between bins

Summary

are computed for each atmospheric state at the SGP site to understand the variability within states.

required to parameterize their variability. However, mean cloud

Future Work

(5) Application: Cirrus Formation

Heterogeneous Nucleation · 2 cm/s updraft · Lower critical supersaturation

· Small numbers of ice crystals grow

threshold

Homogeneous Nucleation • 50 cm/s updraft

· High critical supersaturation threshold · Large numbers of ice crystals: growth limited; fewer large crystals



(4) Mean and Variability of Cloud and Atmospheric Properties

- Mean and fractional variance (f_x) computed for each atmospheric state
- RHI and V_{meso} are for "in cloud" points



Hogan and Illingworth (2003), J. Atmos. Sci., 60



Fractional Variance



- See Marchand et al. ARM (2004) and Orr and Kropfli (1999)
- dBZe -25 -23 -21 0:0:0: 0.0.0.0.0

Height

Mesoscale Velocity (Vmeso)

bin