Motivation

The realism of GCM cloudy radiation and precipitation estimates is currently limited by a poor representation of clouds. Specifically, both the horizontal subgrid variability of cloud properties and its vertical correlation with other cloudy layers (cloud overlap) needs improvement.

# Analysis of Cloud Resolving Model Simulations



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High and low layers from a CRM simulation (Goddard Cumulus Ensemble model, 128x128 km<sup>2</sup>). A frontally-lifted large-scale cloud overlies a cold dry layer penetrated by warm, moist convective bands from a wetted surface.



Step III: Calculate radiation (Independent Column Approximation) for original CRM clouds and copula regenerated synthetic clouds. Copula versions were built for maximum overlap (MAX), random overlap (RAN), maximum random overlap (MRO, Geleyn & Hollingsworth), and Räisänen-like generalized overlap (total water version). Two exact overlaps, XOH (in-cloud Homogenized) and XOR (in-cloud Randomized), are shown for comparison. The Gaussian copula (GCOP) performs very well.







Step I: For each layer, the total saturation ratio  $(S=q_T/q_S)$  data are ranked to the range (0, 1) by two methods: (a) by fitting the frequency distribution of S with an analytic PDF [the Generalized Extreme value (GEV) distribution above]. Notice the pronounced skewness; (b) by empirically ranking the S values via their sorted index [the so-called Empirical Distribution Function (EDF)].

Inter-layer Correlations



Step II: the inter-layer S correlations are modeled with a Gaussian copula (a type of joint distribution in rank-space). The blue, cyan, orange and brown lines contain 20, 40, 60 & 80% of the joint S probability. The thick lines are for the CRM data and the thin lines are for synthetic data regenerated from the Gaussian copula. The agreement is good. The clear, light pink, and dark pink areas are clear in both layers, cloudy in one layer, and cloudy in both layers.

### Conclusions

We have built a Gaussian copulabased cloud generator that models correlations (and anticorrelations) between arbitrary separated layers. The model performs well & handles skewed layer distributions naturally. For more details see Norris et al., QJRMS, **134**: 1843-1864 (2008).



synthetic cloud field. The  $\Delta s_{W_c}(\%)$  are the mean biases in the sdev of IWP for the regenerated clouds compared to the ARM clouds. Further work on the copula analysis is required.



# Using ARM Data to Model Cloud Overlap with Copulas

Copulas in the News Copulas have gained some recent notoriety (e.g. WIRED, "The Secret Formula that Destroyed Wall Street", 3/09) as a cause of the "economic meltdown." In reality, copulas serve a very useful role in the analysis of multivariate correlations if used properly, not treated as "black boxes".

## Analysis of ARM Microbase Data

inIn the case of Microbase IWC retrievals, only the condensed tail of the total water distribution is observable. To what extent can this data be used to constrain the total water PDFs and their rank correlations between layers? In the case study below, Microbase IWC retrievals for 1/1/04 are segmented into 2hr (72km @ 10m/s) windows, of similar scale to a GCM gridcolumn. Layers are fitted with skewed GEV distributions, assuming the PDF smoothly extends into the clear fraction and that its probability mass is replicated. Preliminary inspection of the fits look reasonable, but quantitative analysis indicates that better modeling of upper tail behavior may be needed. A fully bounded distribution such as the Beta distribution may be required. For now,

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study the most appropriate marginal PDFs and copula over longer sequences of data with differing synoptic conditions. Analysis and parameterization of the correlation matrices is also needed.

