

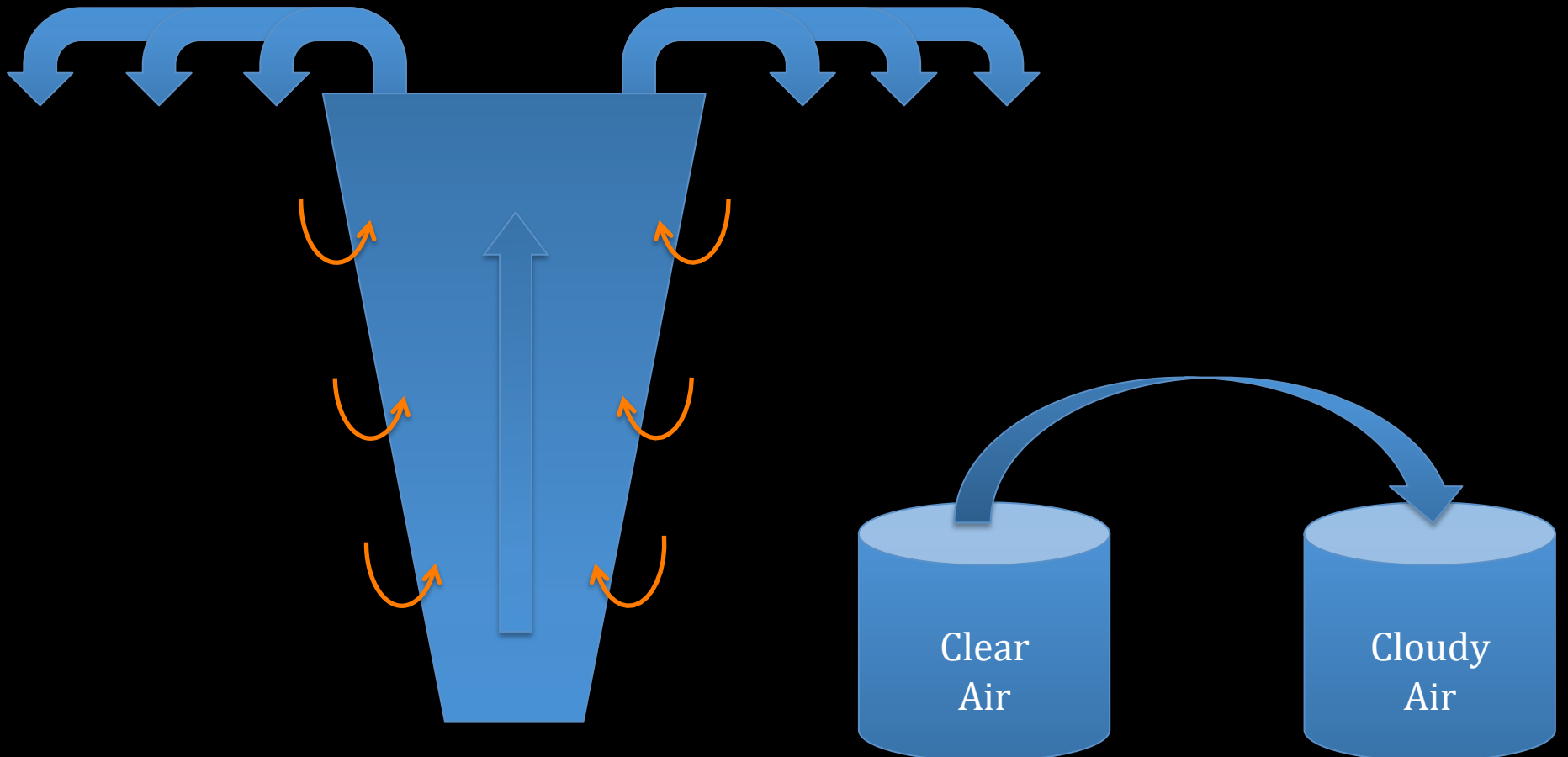
An aerial photograph of a tropical island, likely in the Pacific or Indian Ocean. The island is lush green with some buildings and a small airport. The water is a deep blue, and a large, bright white cloud formation is visible in the sky above the island. The sky transitions from a clear blue to a hazy, overcast grey near the horizon.

Entrainment in Tropical Convection

Walter Hannah
CMMAP Student Colloquium
Aug. 2013

Entrainment

en•train (v.) – to **transform** “non-cloudy” air into “cloudy” air



Entrainment

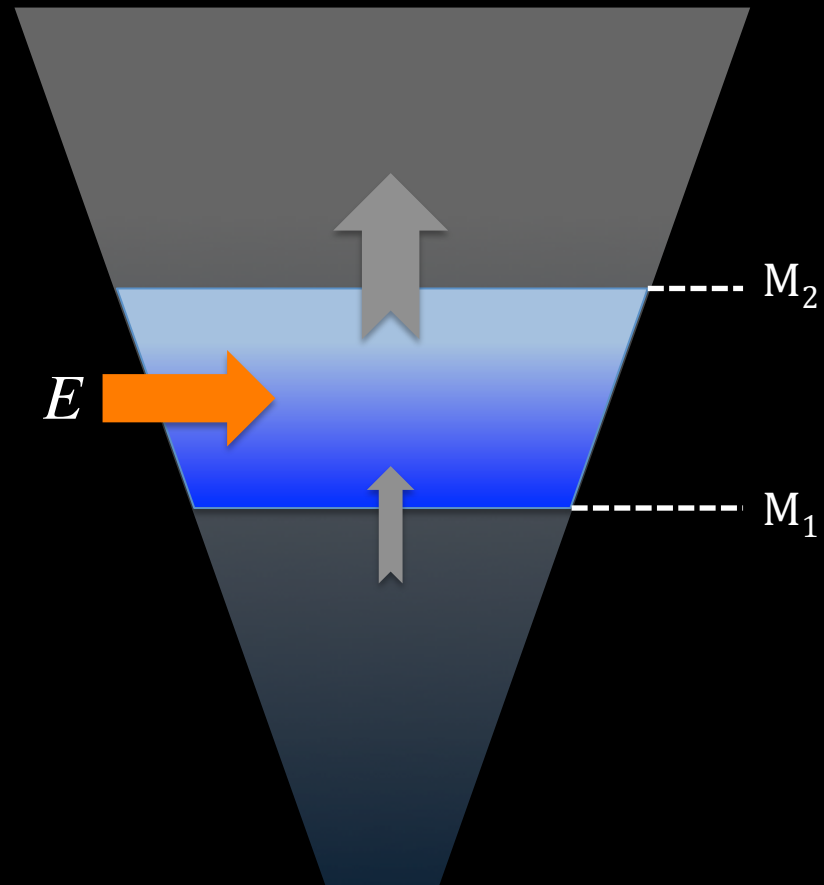
For the simplest **plume model**, ignoring detrainment, the change in mass flux with height can be used to define entrainment

Total
Entrainment

$$E = \frac{dM}{dz}$$

Fractional
Entrainment

$$\varepsilon = \frac{1}{M} \frac{dM}{dz}$$



Direct Measurement

There are many methods to estimate E or ε in observations and cloud resolving models

A new method that **directly measures** entrainment in high resolution models has been introduced:

J. Dawes and P. Austin (2011)

D. Romps (2010)

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MONTHLY WEATHER REVIEW

VOLUME 139

Interpolation of LES Cloud Surfaces for Use in Direct Calculations of Entrainment and Detrainment

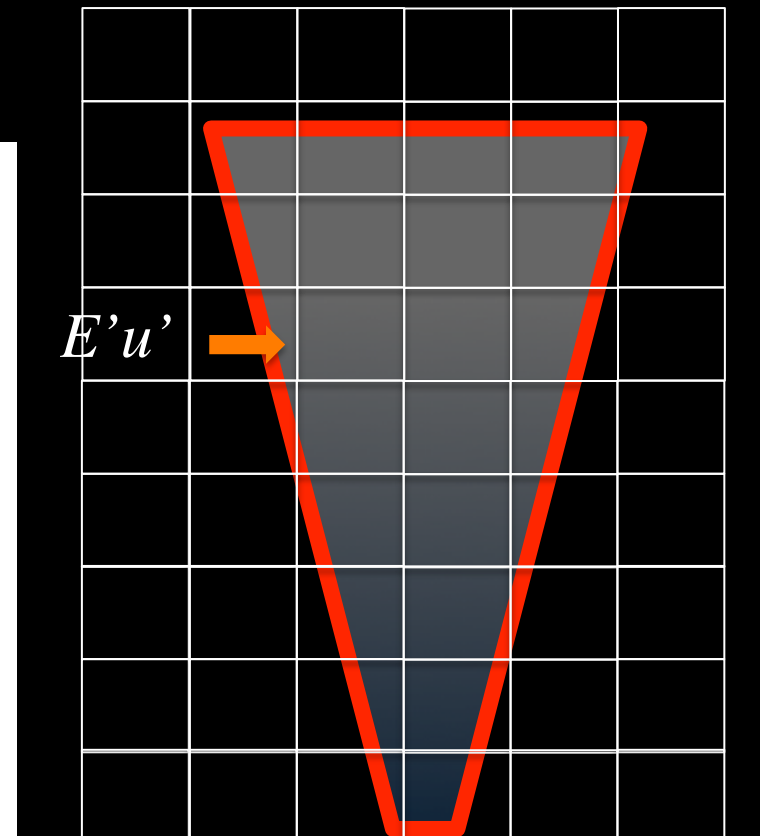
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(Manuscript received 22 April 2010, in final form 4 August 2010)

ABSTRACT

Direct calculations of the entrainment and detrainment of air into and out of clouds require knowledge of the relative velocity difference between the air and the cloud surface. However, a discrete numerical model grid forces the distance moved by a cloud surface over a time step to be either zero or the width of a model grid cell. Here a method for the subgrid interpolation of a cloud surface on a discrete numerical model grid is presented. This method is used to calculate entrainment and detrainment rates for a large-eddy simulation (LES) model, which are compared with rates calculated via the direct flux method of Romps. The comparison shows good agreement between the two methods as long as the model clouds are well resolved by the model grid spacing. This limitation of this technique is offset by the ability to resolve fluxes on much finer temporal and spatial scales, making it suitable for calculating entrainment and detrainment profiles for individual clouds.



Experiment

Using this new method we would like to find out more about entrainment to guide future convective scheme development

So let's try to answer a few questions:

- Is entrainment fundamentally different in **organized** convection? (Mapes and Neale 2011)
 - How does **shear** affect entrainment?
- Does environmental **humidity** cause less fractional entrainment? (Bechtold et al. 2008)



J. Adv. Model. Earth Syst., Vol. 3, Art. 100004, 20 pp.

Parameterizing Convective Organization to Escape the Entrainment Dilemma

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Received 1 JAMES February 2011



Land-aiding parameters in buoyancy-driven deep convection schemes are among the most sensitive and important submodels in atmosphere models. Unfortunately, there is not a true optimum value for parameter mixing ratio, but rather a dilemma or tradeoff. Excessive diffusion of updrafts leads to unrealistic stratification bias in the mean state, while insufficient diffusion allows deep convection to occur too easily, causing poor space and time distributions and variability. In this two-scale parameter space, compromises are made based on competing metrics of model performance. We attempt to escape this "entrainment dilemma" by making both parameter (shallow entrainment rate) depend on a new prognostic variable ("organization," org) meant to reflect the modified effects of subgrid-scale structure in mesoscale eddies. We test an org scheme in the Community Atmosphere Model (CAM) with a new unified shallow-deep convection scheme (UD-CM, a 2-phase version of the University of Washington scheme). Since buoyant ascent involves natural selection, subgrid structure makes convection systematically deeper and stronger than the pure unorganized case. Plumes of average (or randomly sampled) air rising in the average environment. To reflect this, org is prognostic, but we leave it dimensionless. A time scale characterizes its behavior (here ~ 3 hrs or 2 months). Carefully to avoid a rain evaporation, but other sources can be added easily. We also try org horizontally transported by advection, as a time-weighted mean over the convecting layer. Linear

with show the mean ability of evading a bias implemented in as a time-lagged but dilemma, since fully able the pioneering able variability.

QUARTERLY JOURNAL OF THE ROYAL METEOROLOGICAL SOCIETY
© J. A. Met. Soc. 134: 1117–1131 (2008)
Revised online 15 May 2008
www.met.rdg.ac.uk/journals/qjrmets



Advances in simulating atmospheric variability with the ECMWF model: From synoptic to decadal time-scales

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European Centre for Medium-Range Weather Forecasts, Reading, UK

ABSTRACT. Advances in simulating atmospheric variability with the ECMWF model are presented that stem from revisions of the convection and diffusion parameterizations. The revisions concern in particular the introduction of a variable convective adjustment time scale, a convective entrainment rate proportional to the environmental relative humidity, as well as the tropospheric diffusion coefficients for heat and momentum based on moist–(Peltier) functional dependencies. The forecasting system is evaluated against analyses and observations using high-resolution reanalysis, decadal reanalysis and ensemble forecasts, monthly and seasonal averages, and decadal averages with coupled atmosphere–ocean models. The results show a significantly higher and more realistic level of monthly activity in terms of the amplitude of tropical and extratropical monsoons, synoptic and planetary perturbations. Importantly, with the higher variability and reduced bias not only the precipitation system are improved, but also the multidecadal interannual variance in the short and medium ranges. Furthermore, for the first time the model is able to reproduce a realistic spectrum of convectively coupled equatorial Kelvin and Rossby waves, and maintains a realistic amplitude of the Madden–Julian oscillation (MJO) during monthly forecasts. However, the propagation speed of the MJO is slower than observed. The highest tropical tropospheric wave activity also results in better atmospheric compositions and ozone through deposition of momentum.

The partitioning between convective and resolved precipitation is unaffected by the model changes with roughly 65% of the total global precipitation being of the convective type. Finally, the changes in convection and diffusion parameterizations resulted in a larger spread of the ensemble forecasts, which allowed the amplitude of the initial perturbations in the ensemble prediction system to decrease by 30%. Copyright © 2008 Royal Meteorological Society

Key words: atmospheric variability; model climate; tropical waves; convection; vertical diffusion; seasonal weather prediction

Received 28 February 2008; Revised 16 June 2008; Accepted 18 June 2008

1. Introduction

Forecasting the state of the atmosphere involves the prediction of a slowly evolving mean equilibrium state (also referred to as the climate) and the temporal and spatial variations about this state. Predicting the amplitude and phase of the planetary, synoptic and mesoscale perturbations is a deterministic and ensemble prediction system (EPS) is a challenging task.

The main physical processes that lead to the generation of kinetic energy in the atmosphere are the differential radiative heating and the convective heating (Hasselberger and Hendt, 2000; Stanburton et al., 2005). Latent heat release in deep convection excites among others equatorial Rossby, Kelvin and mixed Rossby–gravity waves. It has been demonstrated in numerous modelling studies (Simpf et al., 1994; Schirova and McFarlane, 2004; Lin et al., 2008) that the simulation of these waves in global models is very sensitive to the type of convective parameterization employed. Typically in these studies the simulated tropical variability increases with increasing resolved precipitation at the expense of the parameterized convection. An inadequate representation of the convective forcing and the associated equatorial waves has numerous consequences on errors present on the wavenumber 1 and 2 Madden–Julian oscillation (MJO) as well as on the stratospheric interdecadal variability (the quasi-biennial oscillation (QBO) since these waves propagate upward and interact with the mean flow (Rienecker and Garcia, 2001; Horendahl et al., 2003). These errors even affect radiative meteorology through interaction with the midlatitude Rossby waves. In contrast to convection parameterization schemes, explicit models of deep convection tend to excite a larger and more realistic spectrum of waves. However, these models are very expensive numerically, and other still have problems in representing a good mean atmospheric state (Khairoutdinov et al., 2005), or an overly run on an aquaplanet (Gochis et al., 2007) or on limited domains (Shaw, 2007), and therefore are not applicable yet to global numerical weather prediction.

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E-mail: peter.bechtold@ecmwf.eu

Experiment Setup

$$\Delta x = 500 \text{ m}$$

72 km² domain

Initialized with **tropical mean** sounding

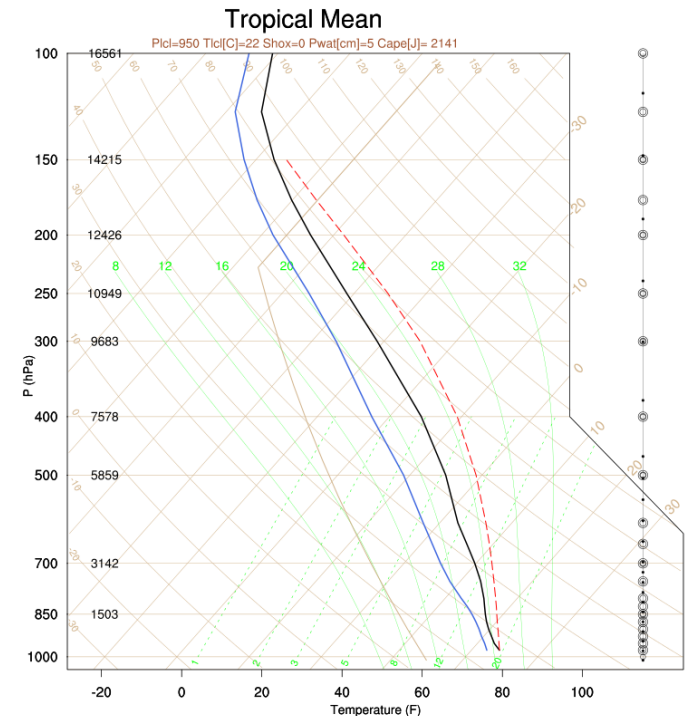
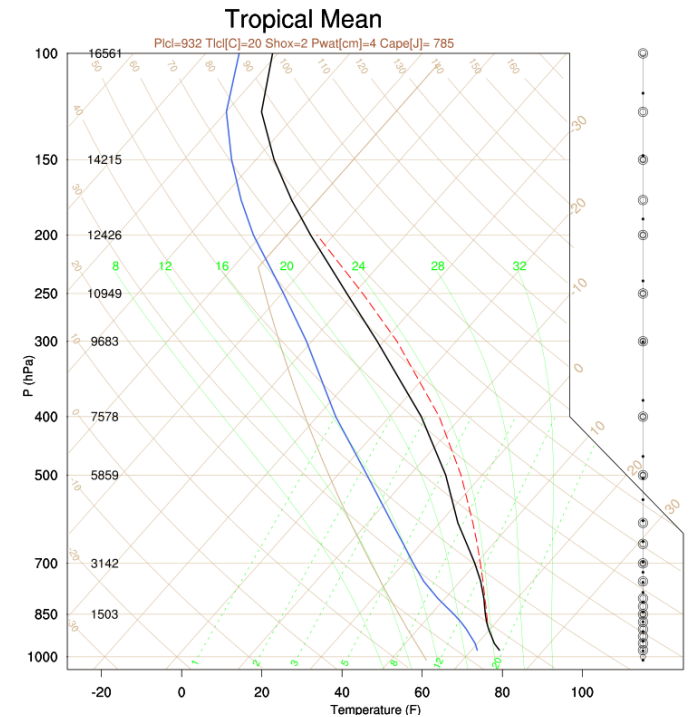
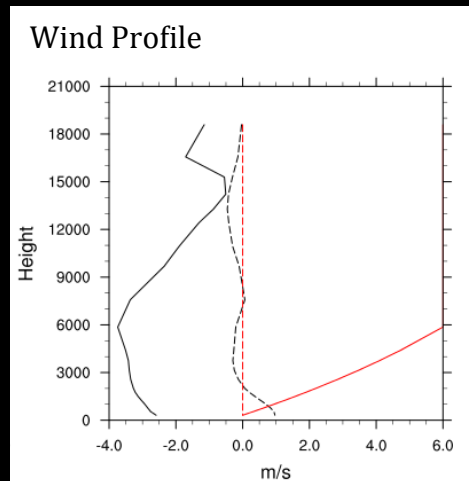
warm/wet bubble was also added:

$$dT = 1 \text{ K}$$

$$dq = 3 \text{ g/kg}$$

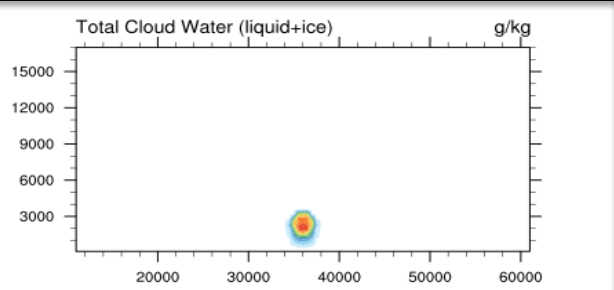
$$X \text{ rad} = 4 / 8 / 12 \text{ km}$$

$$Z \text{ rad} = 2 \text{ km}$$

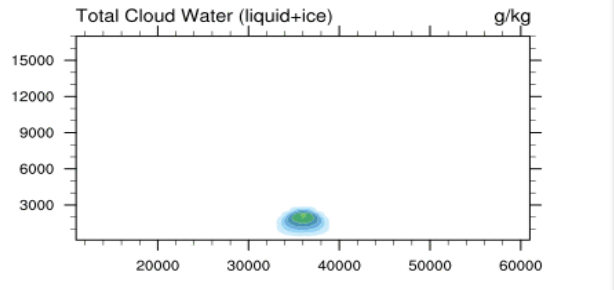


Initial Radius

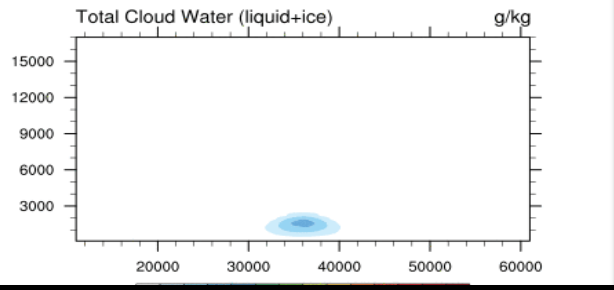
4 km



8 km



12 km



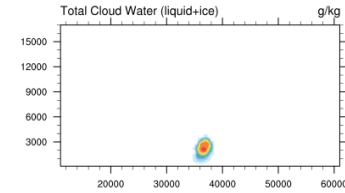
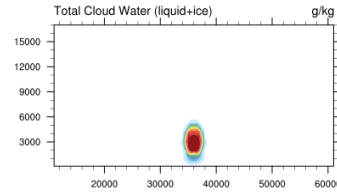
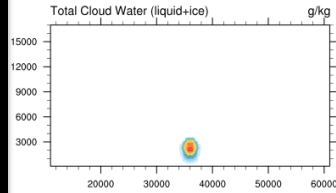
Initial Radius

Control

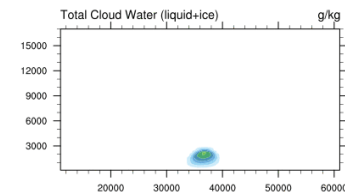
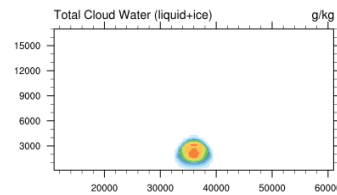
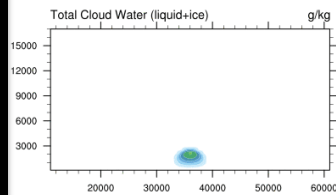
Higher Humidity

w/ Shear

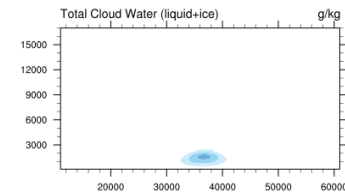
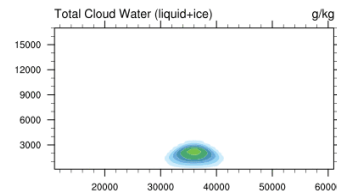
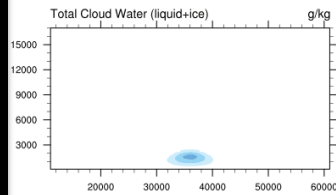
4 km



8 km



12 km



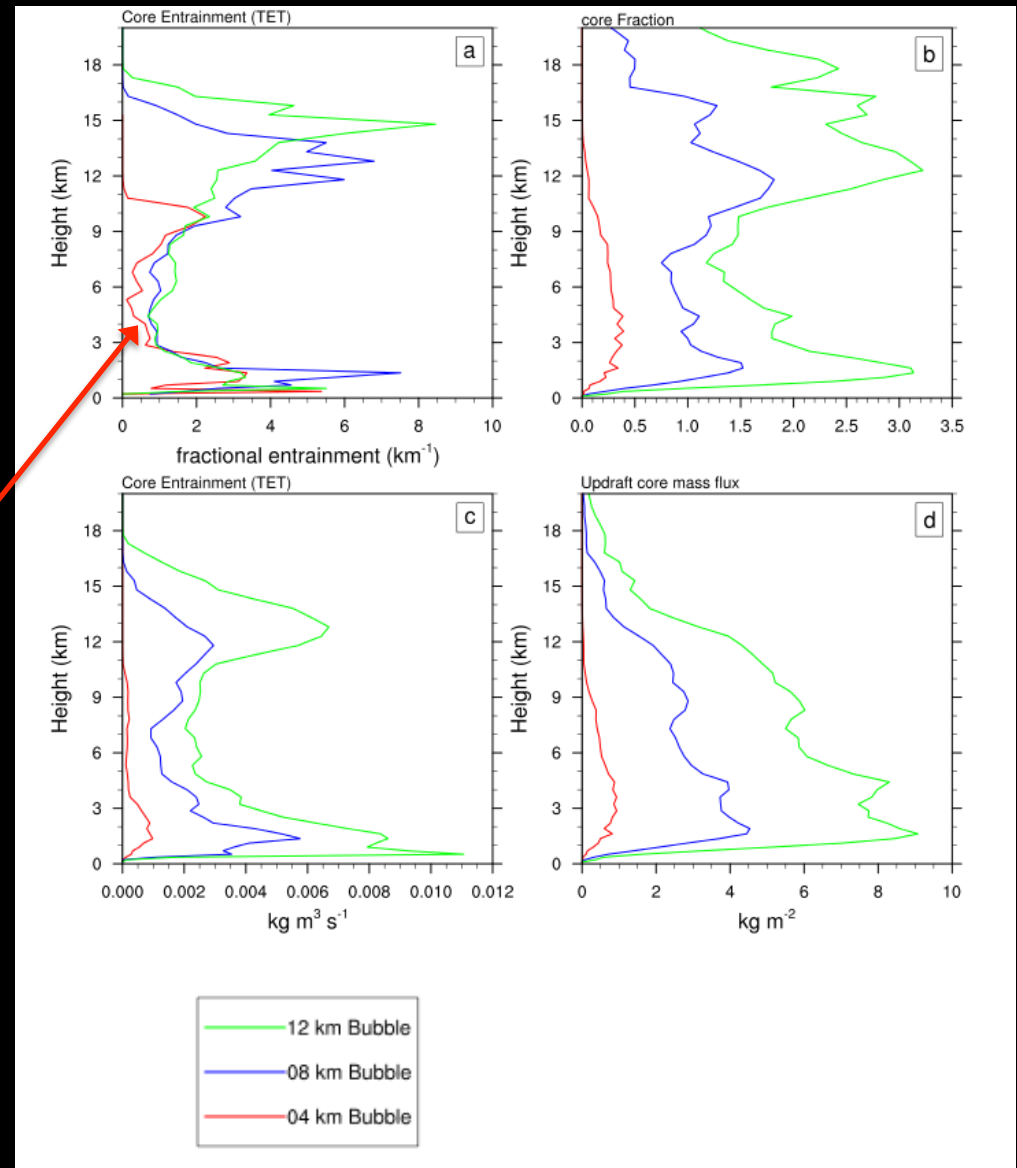
Control Simulations

$$\varepsilon = \frac{E}{M}$$

Larger bubble radius:

- Higher cloud top
- Larger cloud fraction
- Stronger updrafts
- **Larger entrainment?**

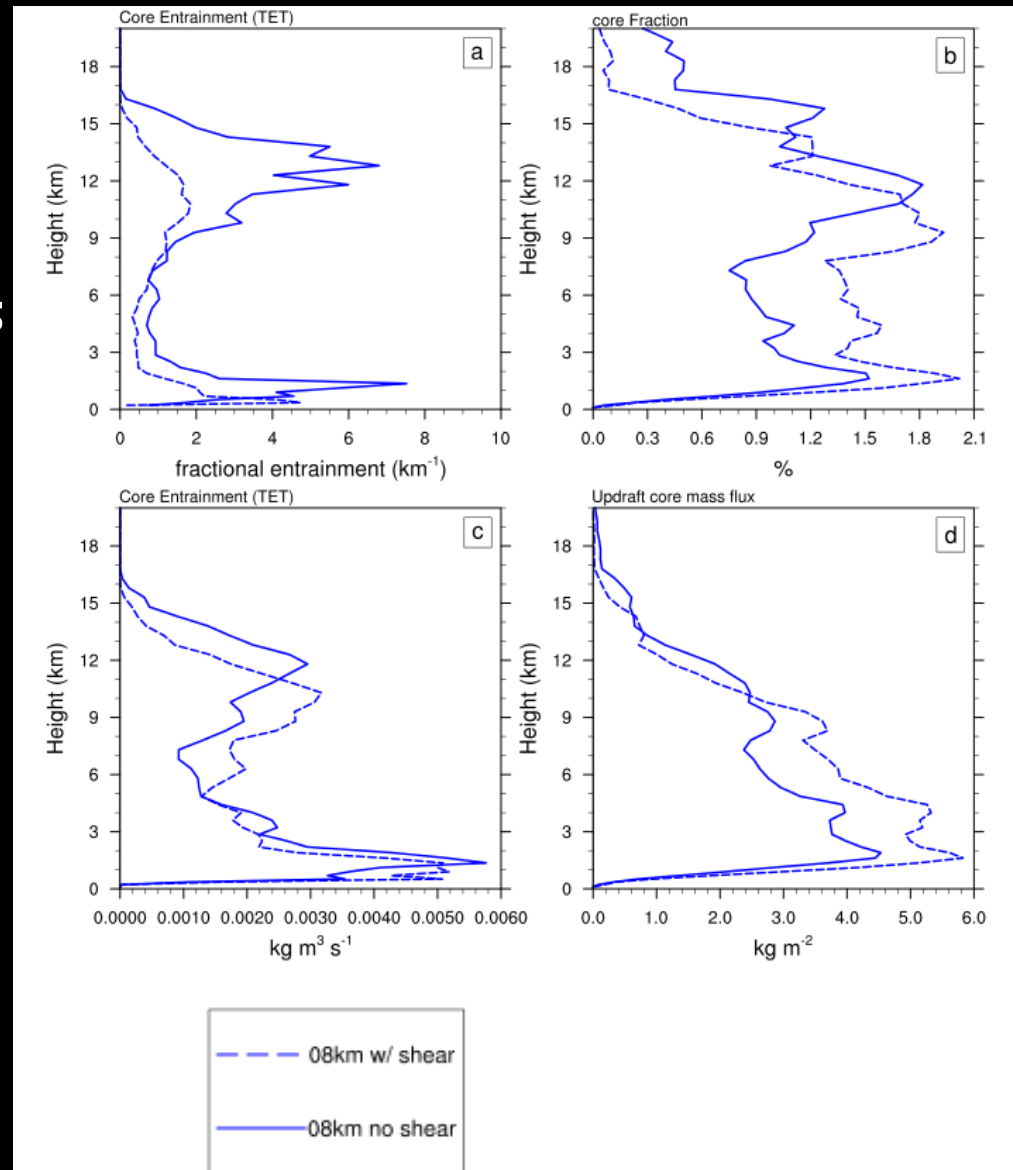
Traditional plume model predicts weaker entrainment with larger radius



Shear Effects

Adding shear:

- Lower cloud top
- Slightly stronger updrafts
- **Weaker entrainment**



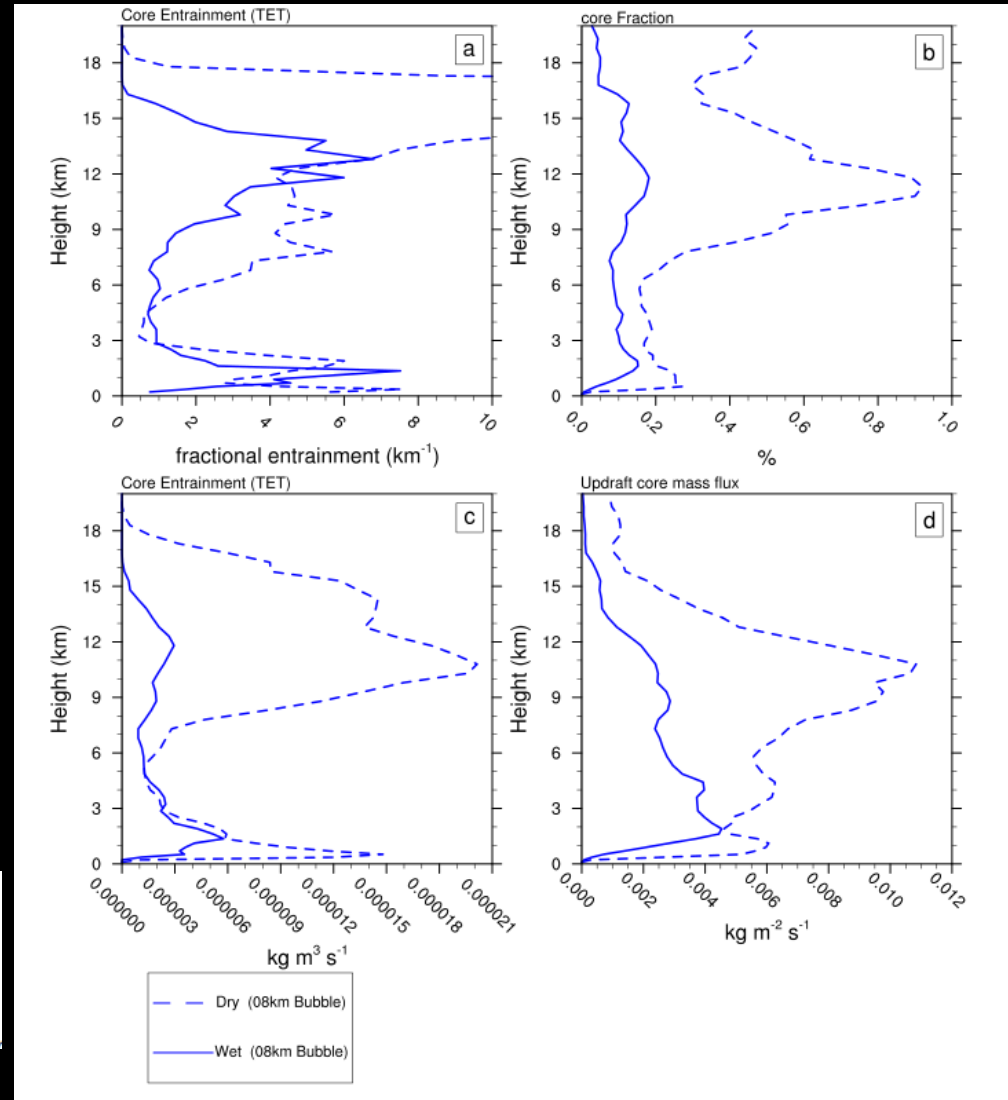
Humidity Effects

Higher Humidity:

- Higher cloud top
- Larger cloud fraction
- Stronger updrafts
- **Stronger entrainment?**

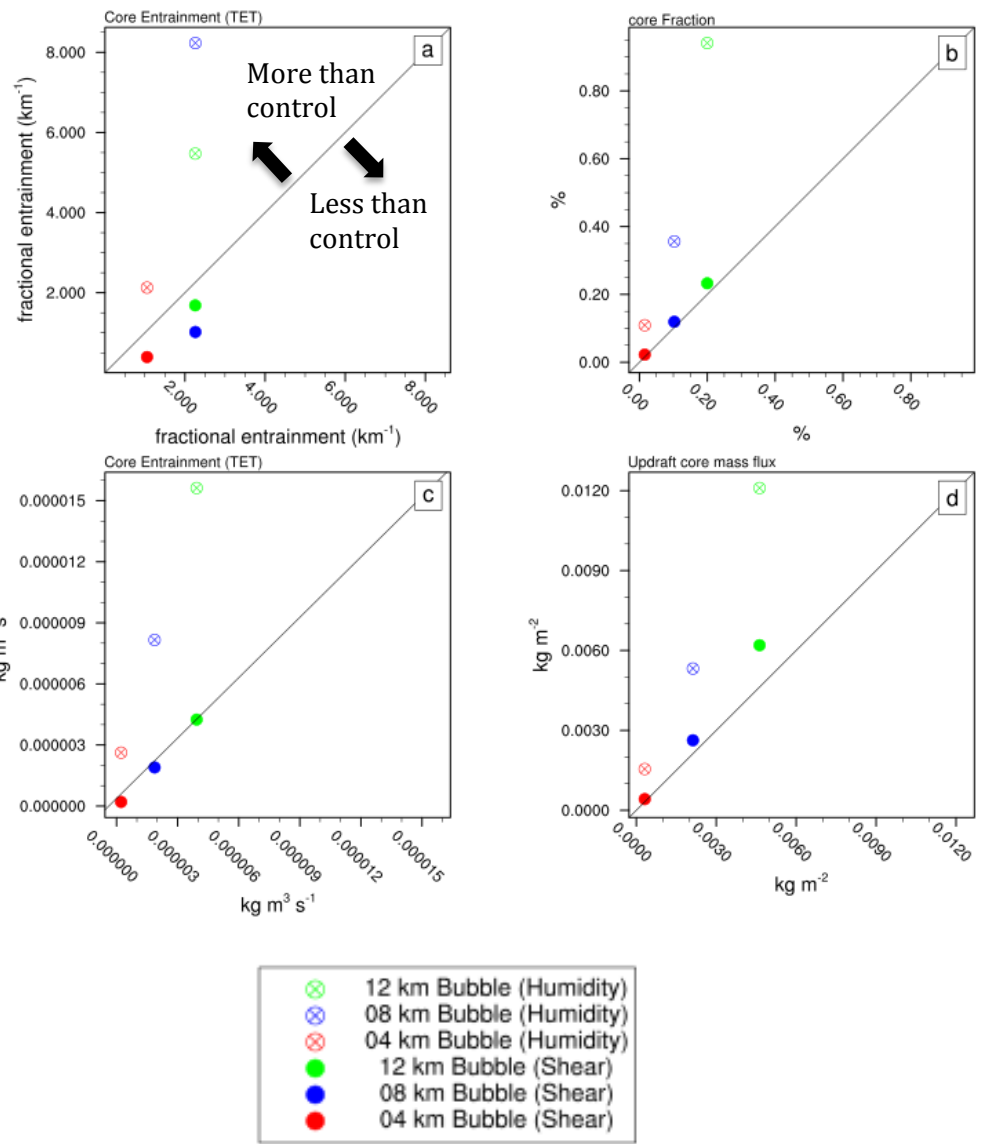
Bechtold et al. (2008)
predicted the opposite

$$\varepsilon = \underbrace{c_0}_{\text{turb}} F_{\varepsilon,0} + \underbrace{c_1 \frac{\bar{q}_s - \bar{q}}{\bar{q}}}_{\text{org, deep, buoy} > 0} F_{\varepsilon,1}; \quad F_{\varepsilon,i} = \left(\frac{\bar{q}_s}{\bar{q}_{s,b}} \right)^{\beta_i}$$



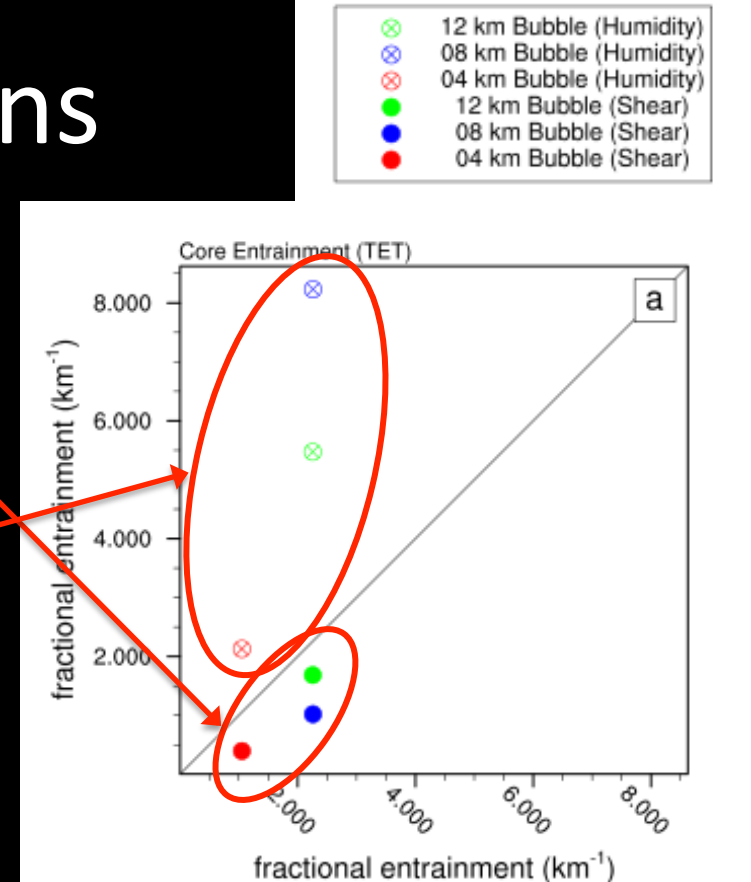
Column Average

$$\epsilon = \frac{E}{M}$$



Conclusions

- Shear effects
 - Bulk of the cloud has **weaker** ϵ
 - Clouds are slightly **larger** and **lower**
- Humidity effects
 - ϵ is **stronger**
 - Clouds are **higher** and **larger**
 - Larger mass flux

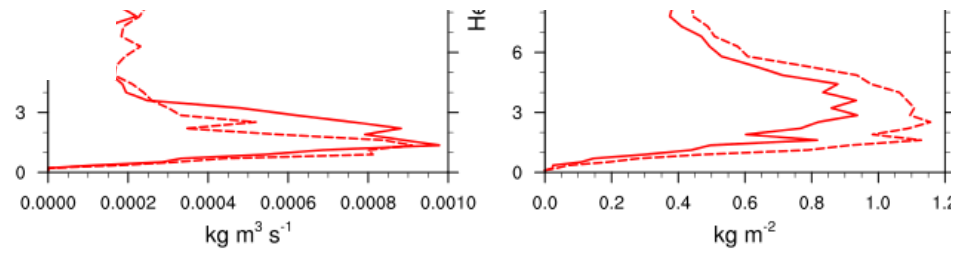
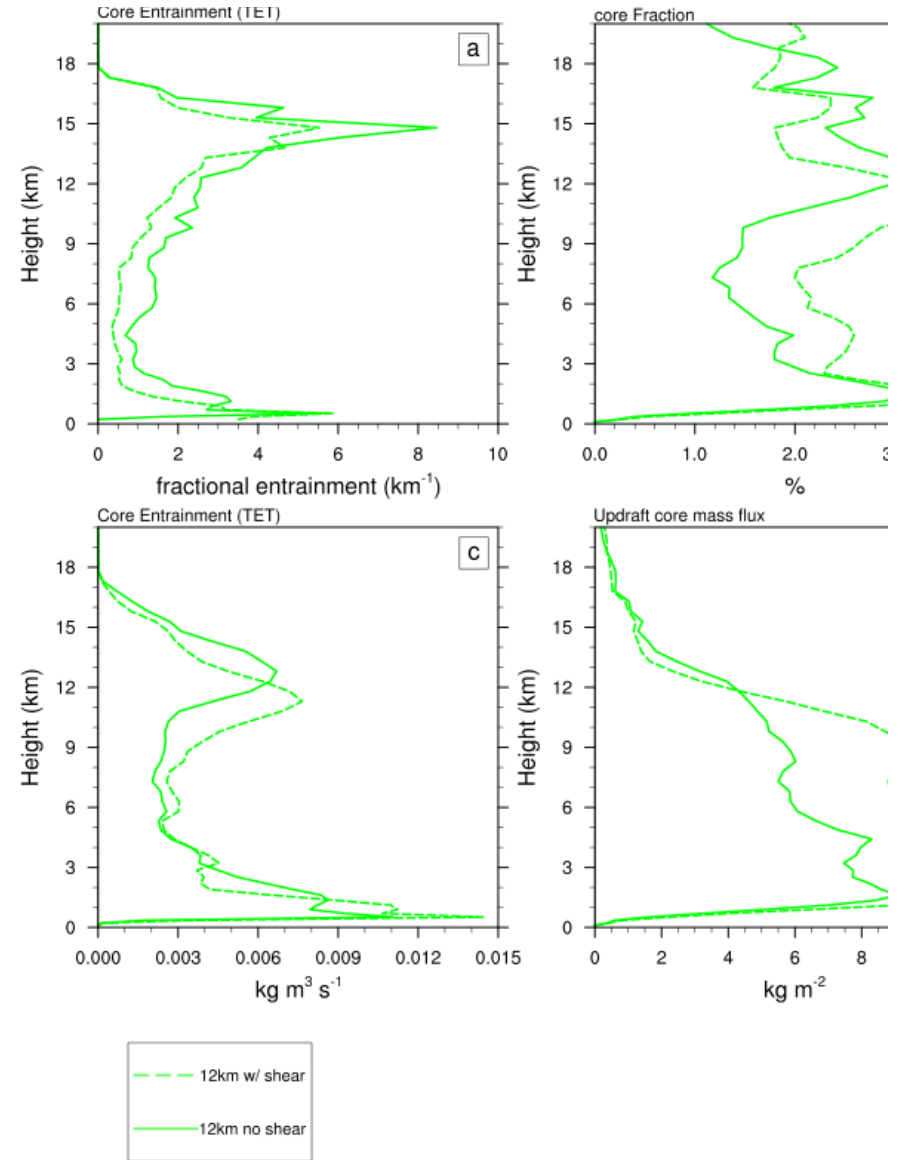
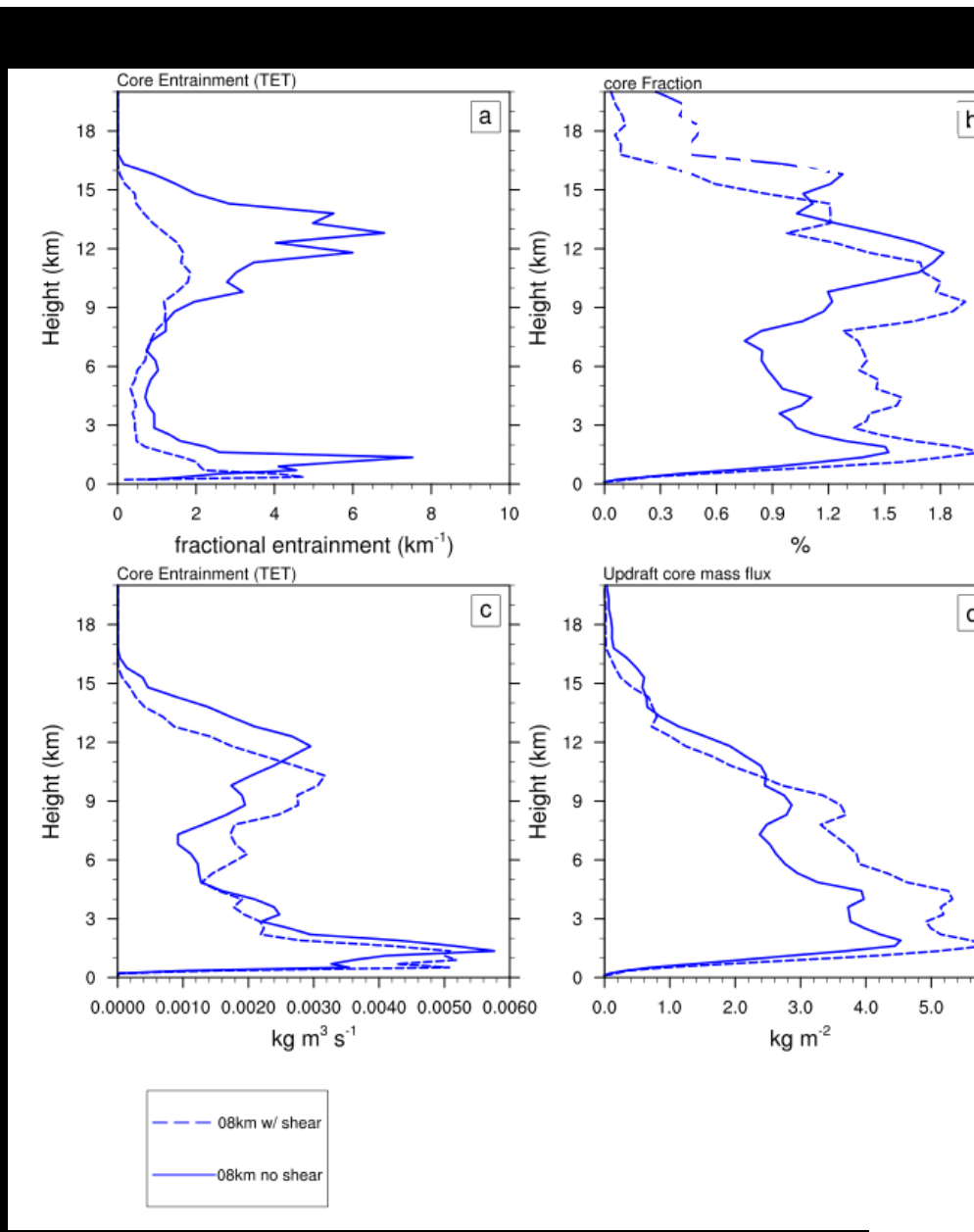


- Caveats
 - Highly **idealized** experiment
 - Profile of **shear** was not realistic
 - Large-scale **forcing** not considered
 - No cloud **interaction**
- Further Work
 - Do entrainment and **dilution** mean the same thing?
 - Is the impact of **mesoscale** organization similar to shear?

Questions?



- Intro
 - What is entrainment?
 - Direct entrainment
- Simulation
 - Setup
 - Animations
- Entrainment comparison
 - Impact of bubble width
 - Shear vs non-shear
 - Impact of more humid environment
 - Scatter plots
 - E vs sheared E
 - E vs Humid E



Effects

