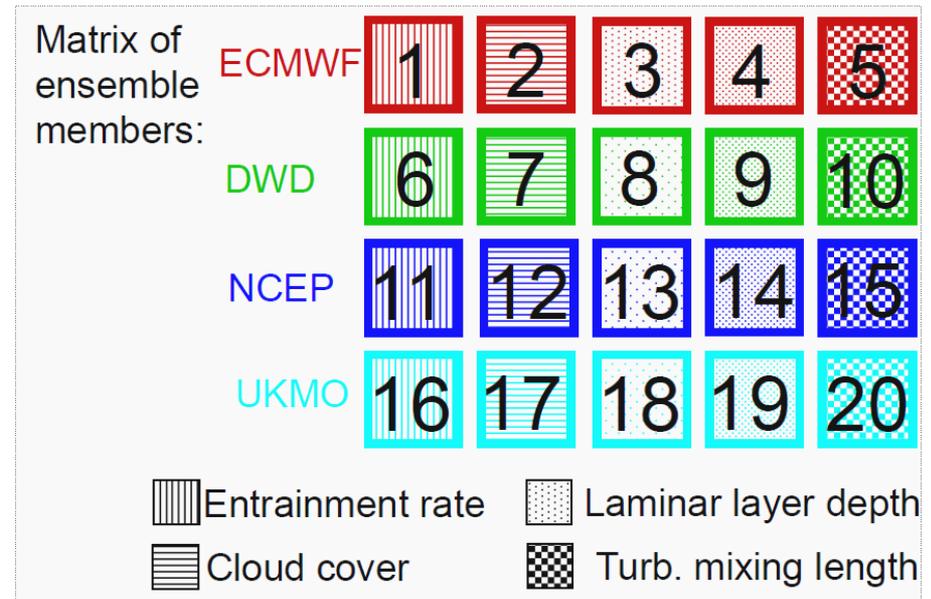


The convective time-scale as indicator of predictability

Christian Keil, Florian Heinlein and George C. Craig
Meteorologisches Institut, Universität München, Germany



- Predictability is flow-dependent
- Sub-groups of EPS perform differently (LBC, PHYS)
- Spread in forecast quality gives information on predictability
- Convective time-scale serves as predictor



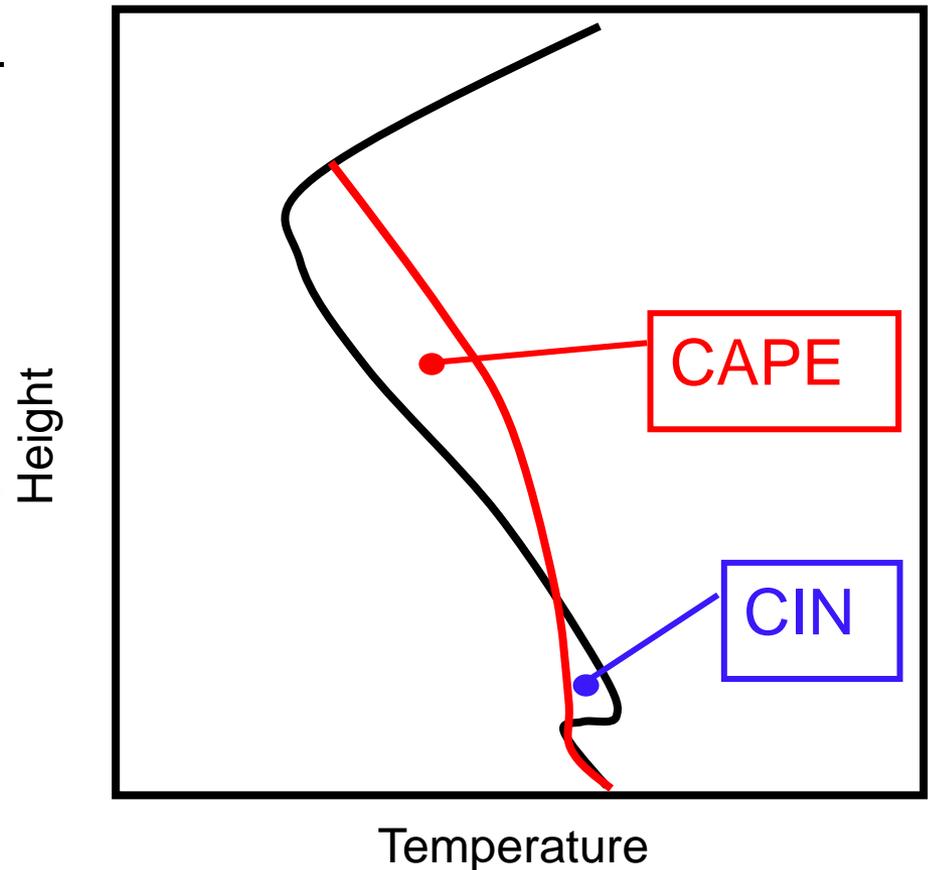
COSMO-DE-EPS



Two mechanisms for control of convection

Many properties of cumulus convection depend on the large-scale environment.

1. **Strong forcing** (equilibrium):
Dynamical production of CAPE
- convection removes CAPE rapidly in comparison to the rate it is being generated
2. **Weak forcing** (triggered): Local perturbations to overcome CIN
- large amounts of CAPE can build up if triggers not present



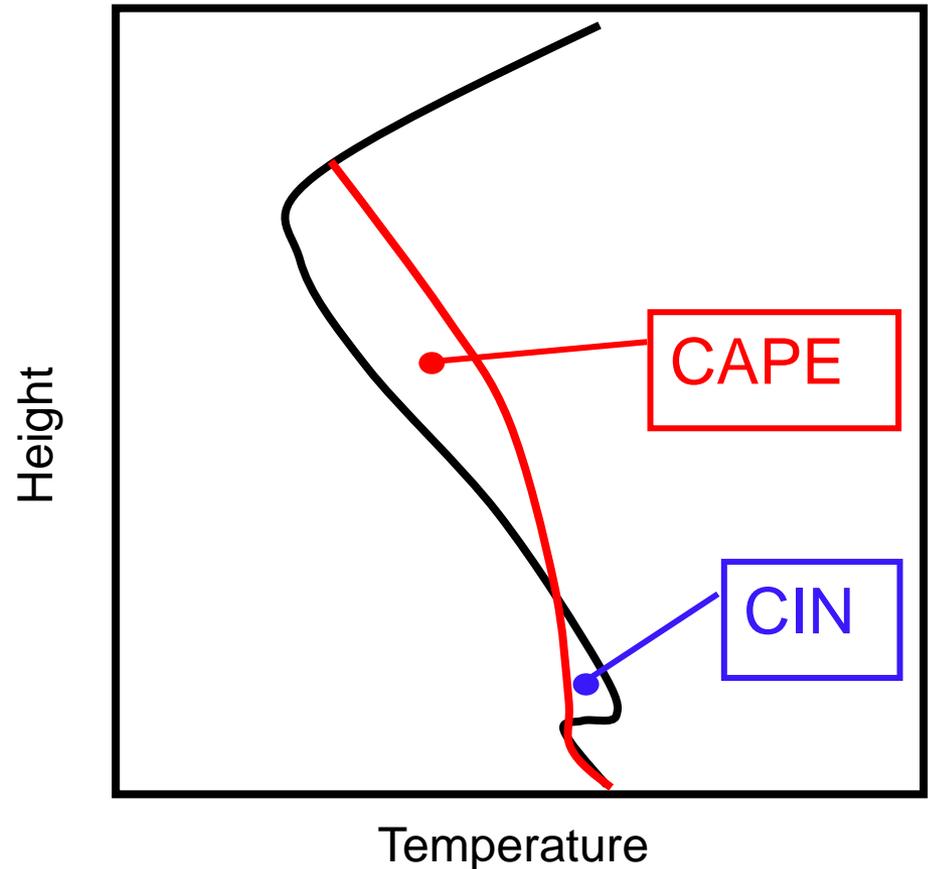
The convective time-scale τ_c

To identify regime, consider time-scale over which convection removes CAPE:

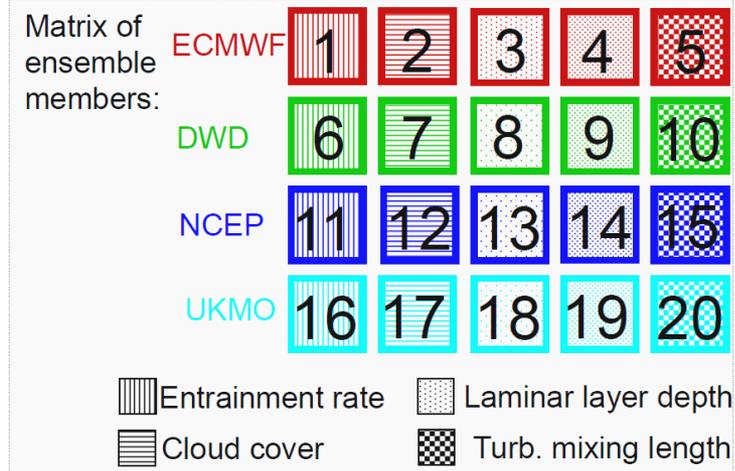
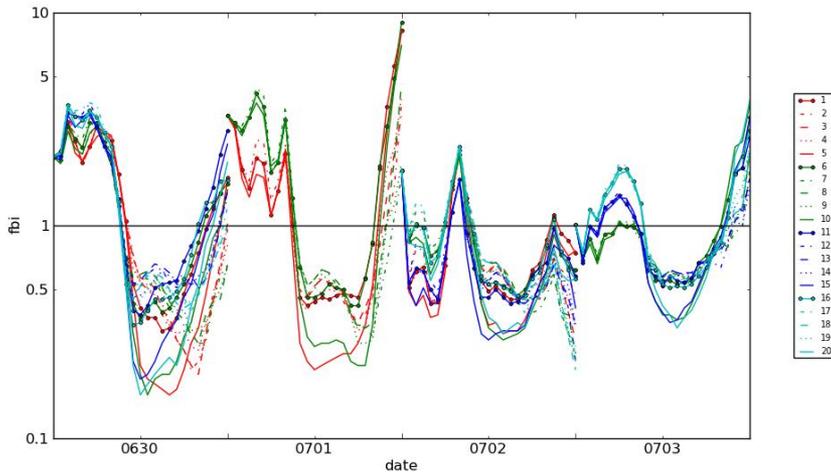
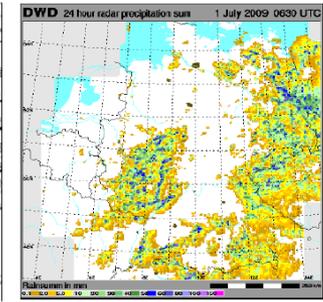
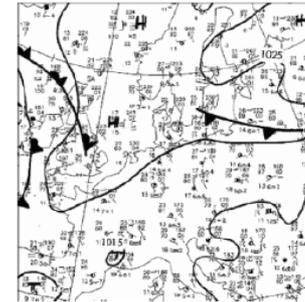
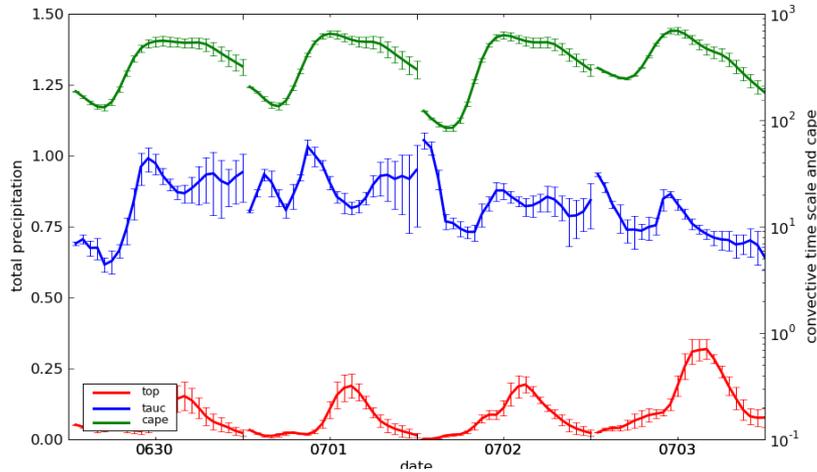
$$\tau_c = \frac{CAPE}{\frac{dCAPE}{dt}}$$

$$\frac{dCAPE}{dt} = \frac{1}{3600} \cdot \frac{L_v}{c_p} \frac{g}{\rho_0 T_0} P$$

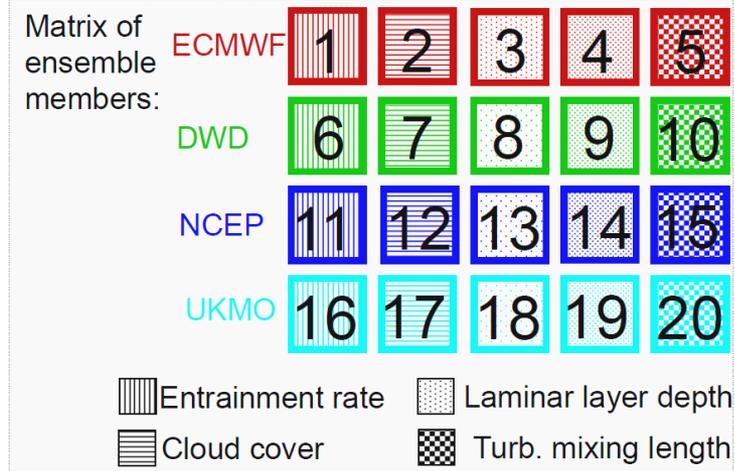
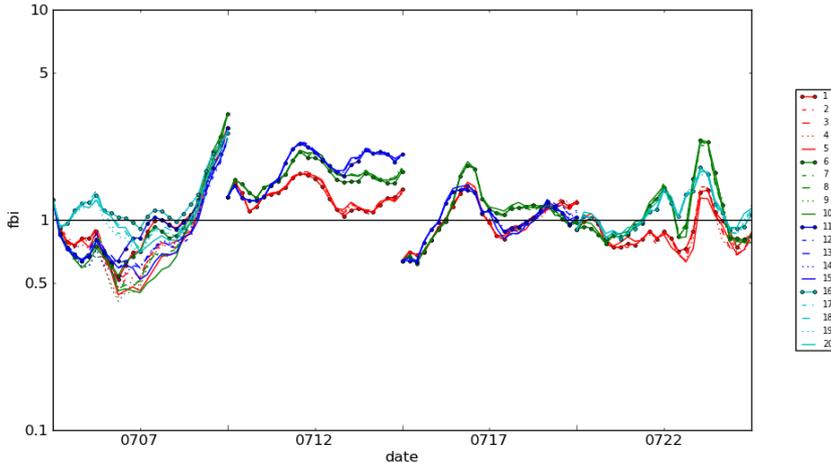
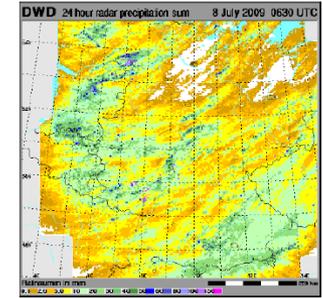
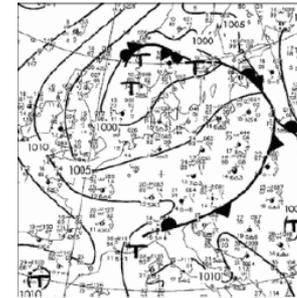
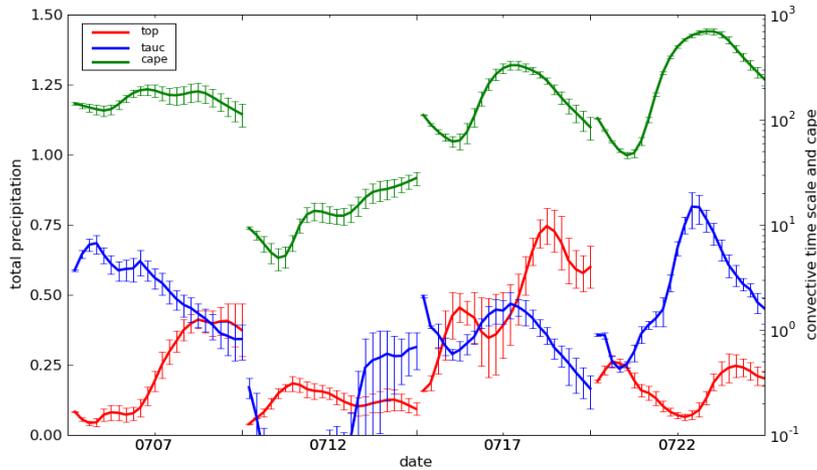
$$\tau_c \sim \frac{CAPE}{P}$$



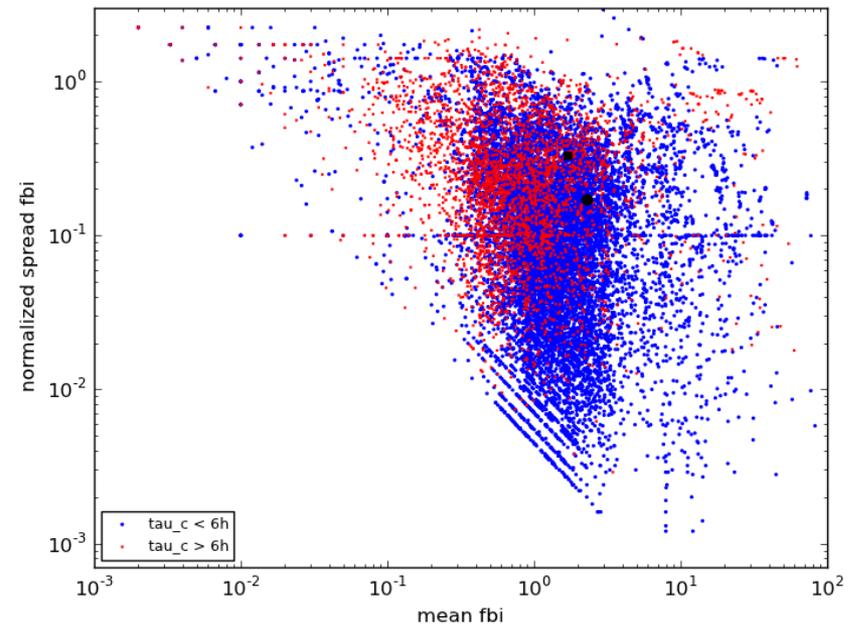
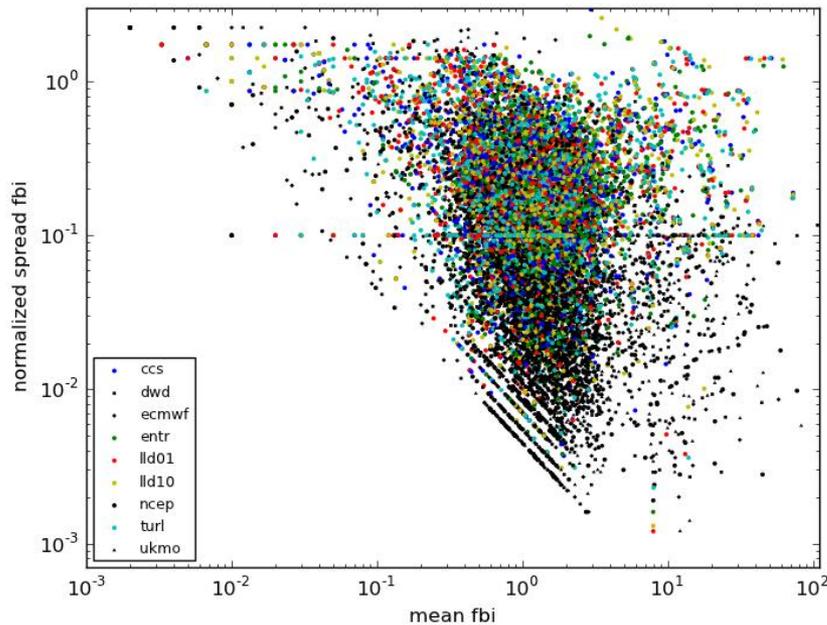
Weak forcing (locally-forced, non-equilibrium)



Strong forcing (synoptically-forced, equilibrium)



Spread-skill relationship stratified by unc. and τ_c



➤ Two predictability regimes:

- **strong forcing:** sensitive to LBC, τ_c less than a few hours, more predictable
- **weak forcing:** sensitive to PHYS, τ_c more than a few hours, less predictable

Hans-Ertel-Center for data assimilation

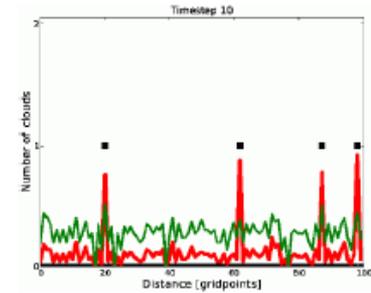
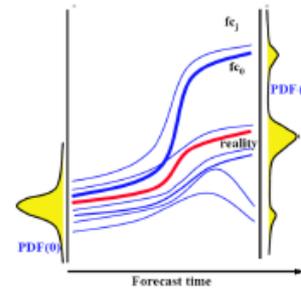
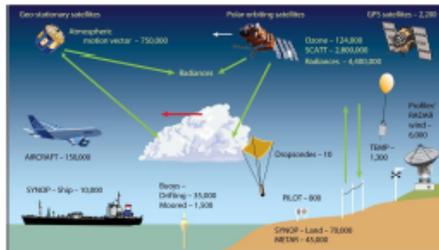
Ensemble-based convective-scale data assimilation and the use of remote sensing observations

- Research group at LMU Munich, collaboration with DLR
- Strong interaction with DWD research department
- Funding from DWD: 2011-2014

Goals:

- Fundamental research in the areas of data assimilation (DA) and ensemble forecasting
- Training of young researchers and students

Research areas



Observation impact

Tools to quantify the analysis and forecast impact of observations in EnDA

Monitoring of observations

Optimized use of observations

Satellite observations

Direct assimilation of MSG SEVIRI VIS+NIR radiances in KENDA

AMV height correction with lidar observations

(Lightning)
(ADM-Aeolus)

Ensemble forecasts

Improved representation of forecast uncertainty

KENDA initial perturbations

Flow-dependence and impact time of perturbations

DA methods

Methods for convective-scale DA

Idealized tests with non-Gaussian error statistics (toy models)

Robust methods for highly non-linear systems