



Using data assimilation to quantify and understand model error

Mark Rodwell

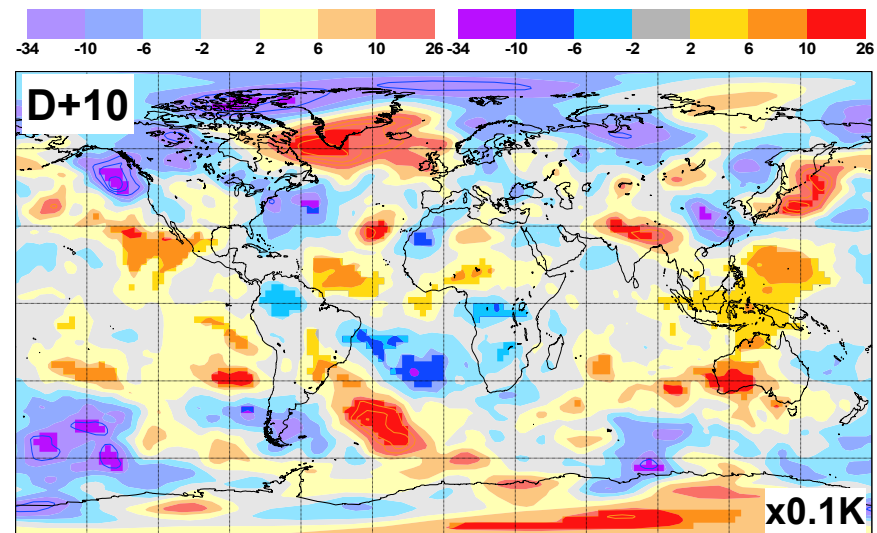
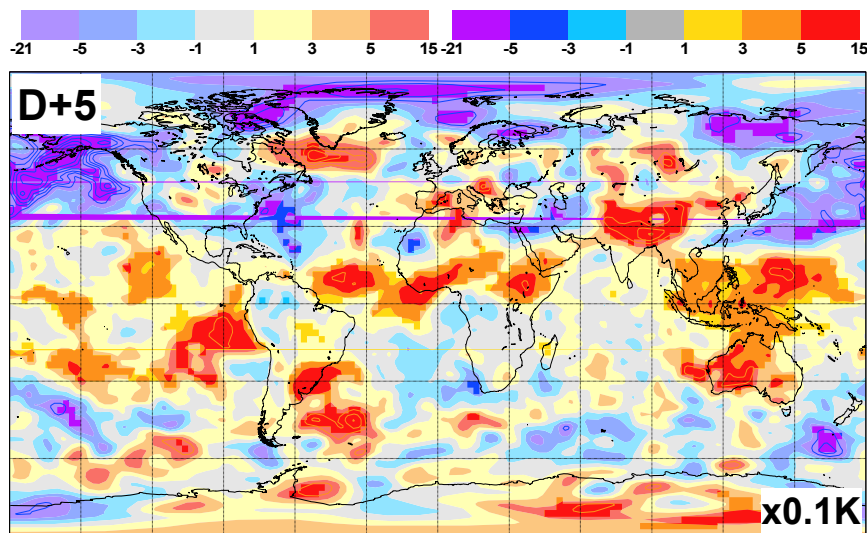
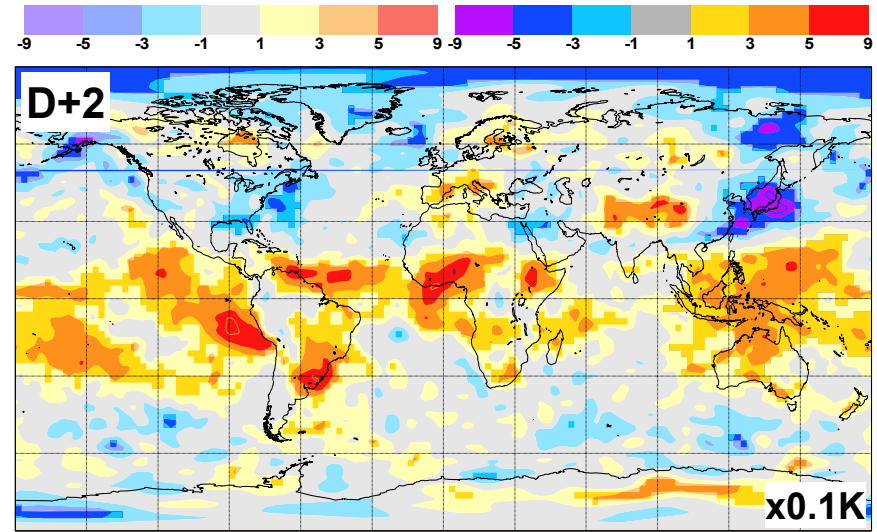
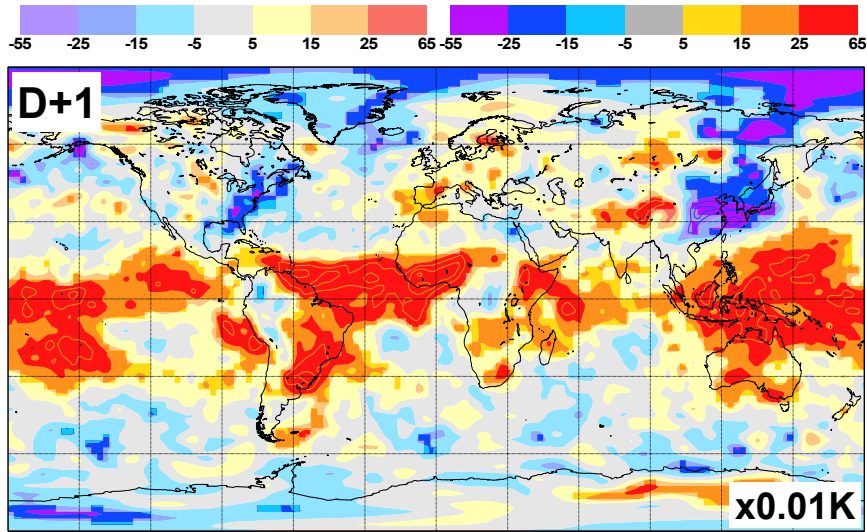
EUCLIPSE kick-off meeting

Utrecht

27-28 September 2010



T500 forecast error as function of lead-time

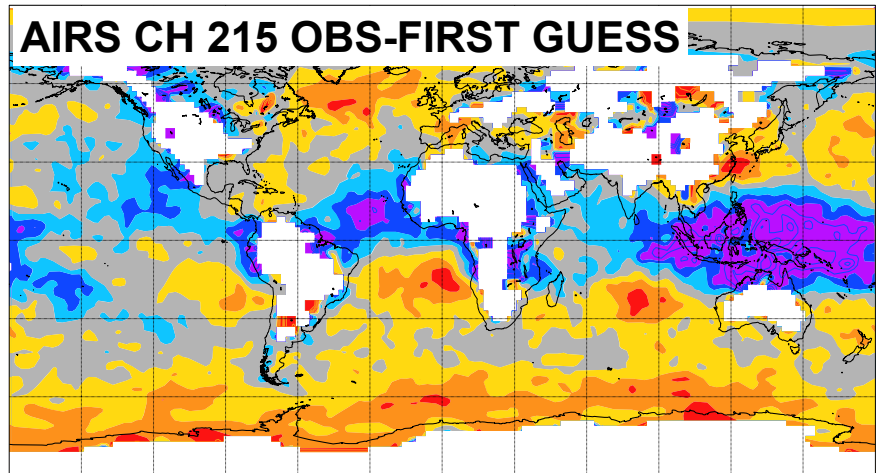
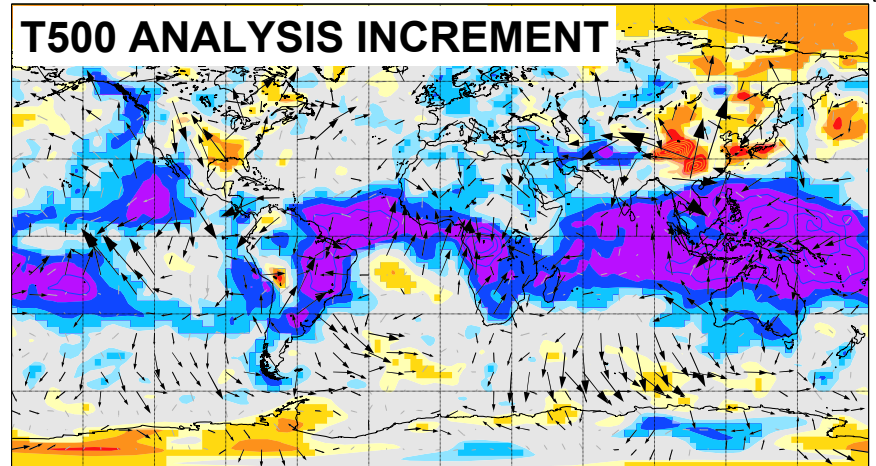
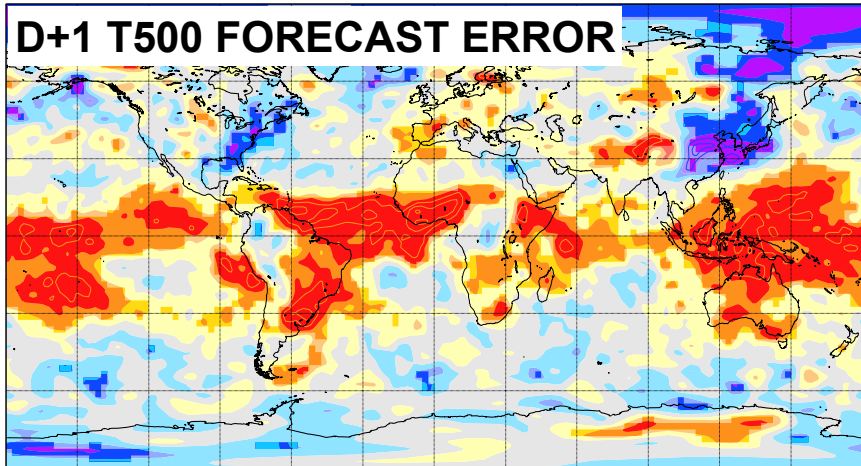
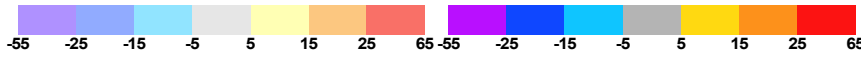


Based on DJF 2007/8 operational analyses and forecasts. Significant values (5% level) in deep colours.



Confronting models with observations

UNIT=0.01K

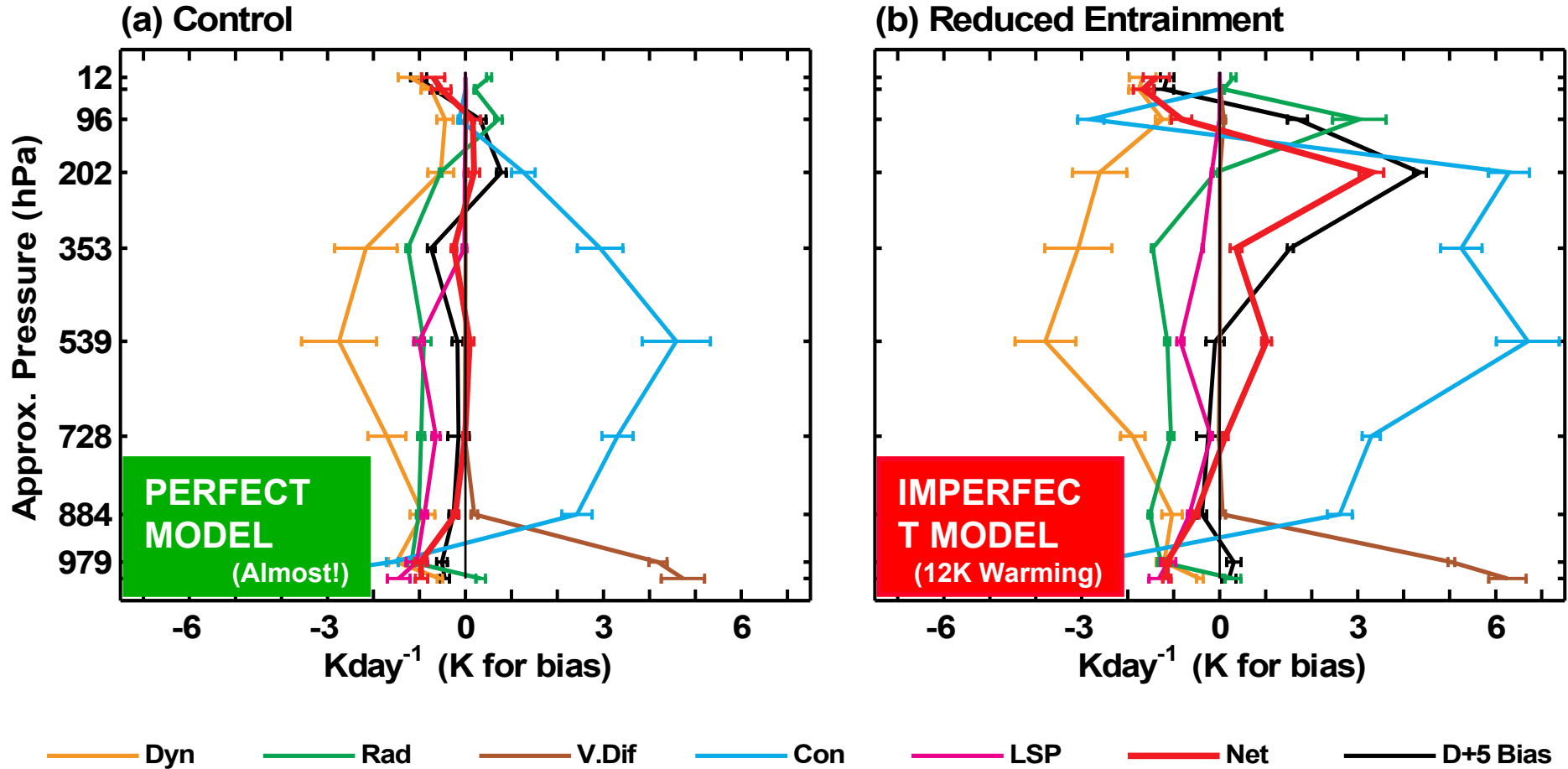


- Every 1° square has data every cycle
 - ~6 Million data values
- Independent vertical modes of information:
 - IASI / AIRS: ~ 15
 - HIRS / AMSUA: ~ 5 (~ 2 IN TROP)
- Anchors (not bias corrected):
 - Radiosonde
 - AMSUA-14
 - Radio Occultation

*Based on DJF 2007/8 operational analyses and forecasts. Significant values (5% level) in deep colours.
AIRS CH 215 BRIGHTNESS TEMPERATURE ~T500*



Amazon January 2005 initial T tendencies

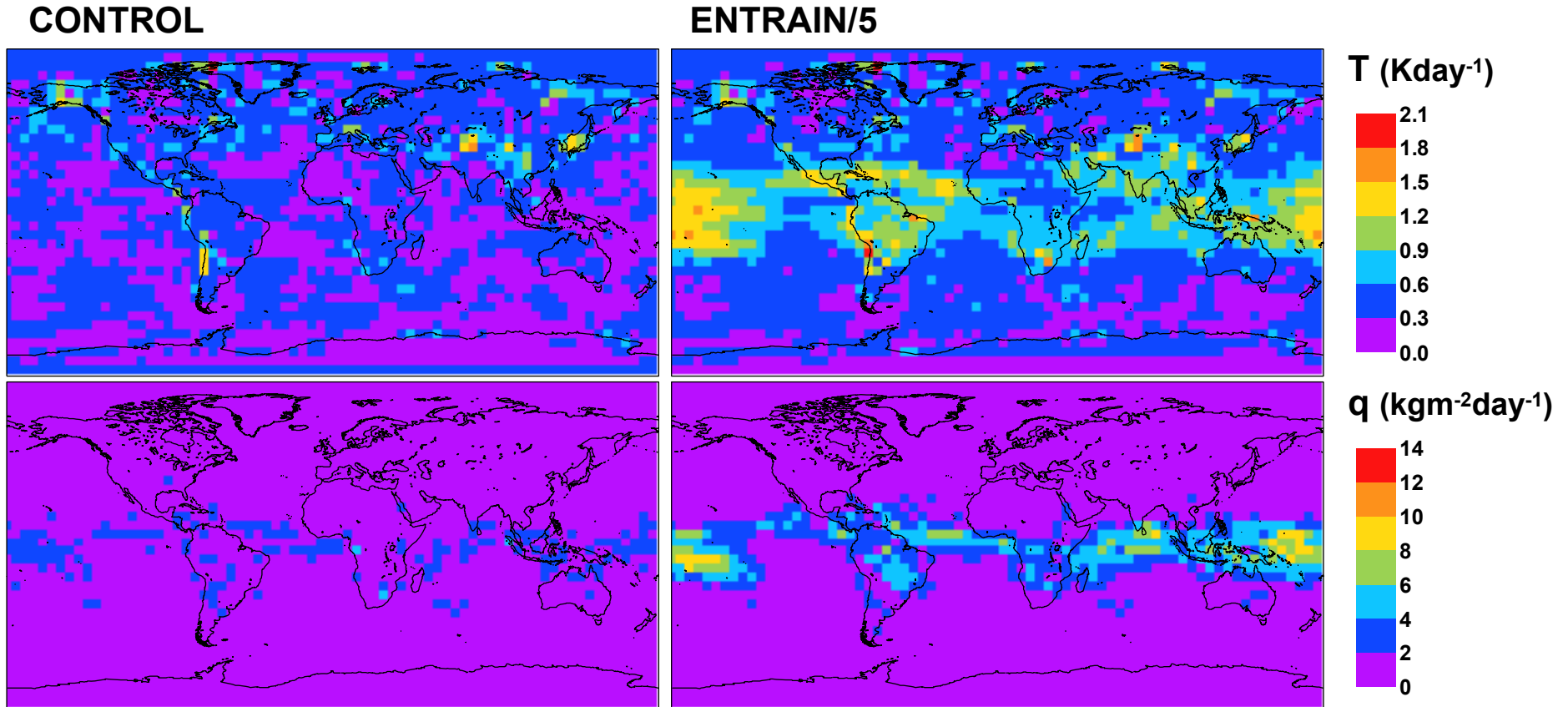


**Reduced Entrainment model is out of balance.
Reject or down-weight?**

Amazon = [300°E-320°E, 20°S-0°N]. Mean of 31 days X 4 forecasts per day X 12 timesteps per forecast. 70% confidence intervals are based on daily means. CONTROL model = 29R1,T159,L60,1800S.



Vertically integrated absolute net tendencies



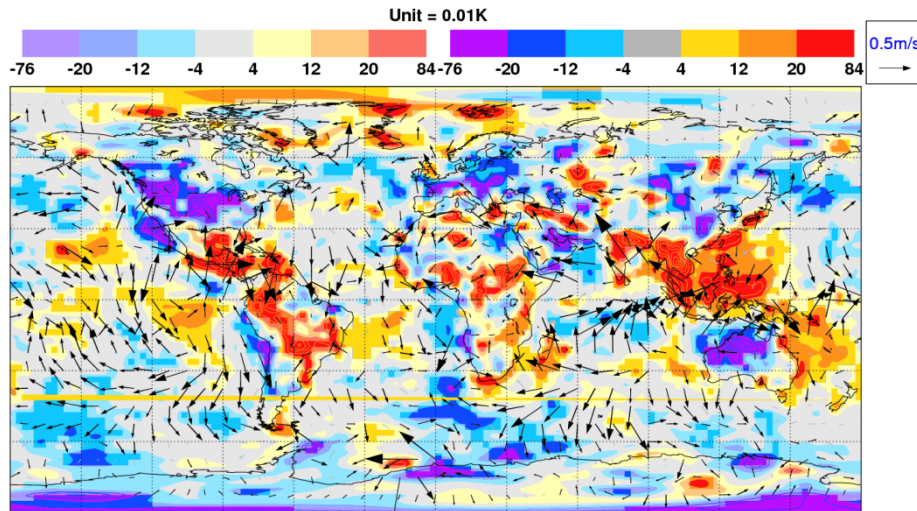
**Can be the basis for a “process metric”
(as opposed to the more widely used circulation metrics)**

*Mean of 31 days X 4 forecasts per day X 12 timesteps per forecast. Mass-weighted vertical integrals.
CONTROL model = 29R1,T159,L60,1800S.*

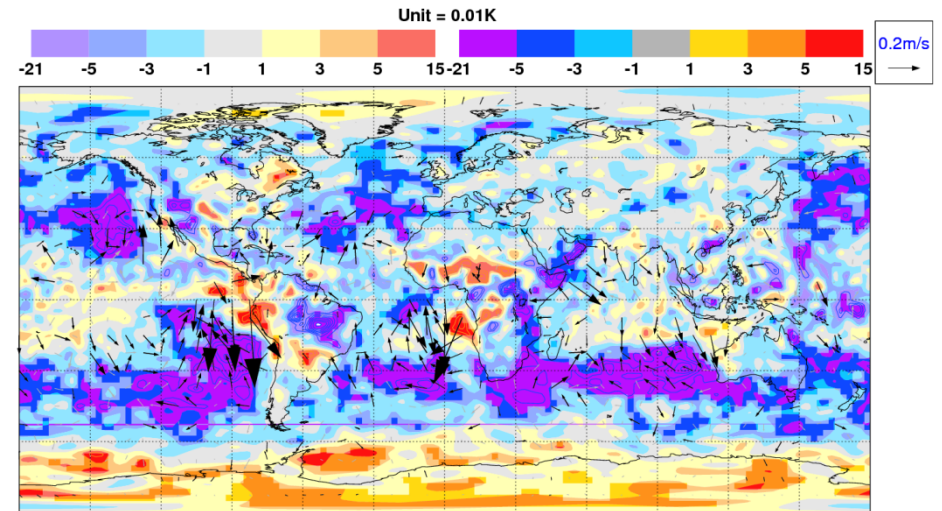


DUALM Boundary-Layer Mass-Flux Scheme

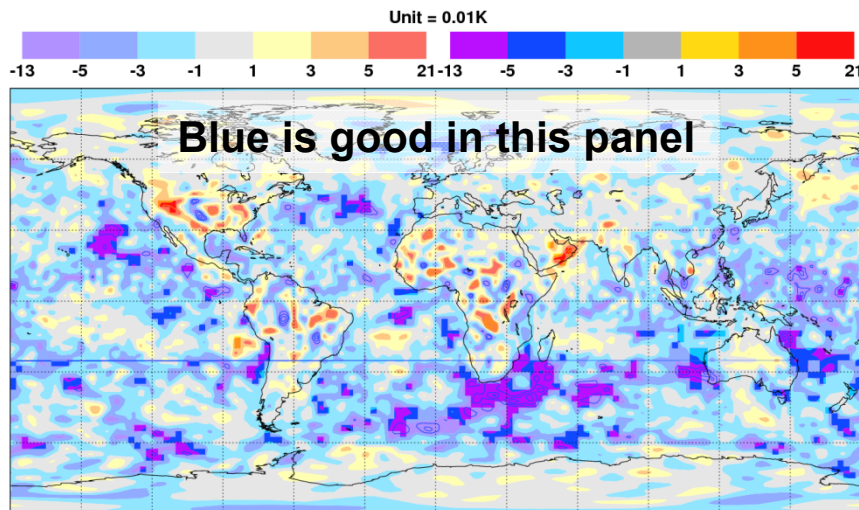
MEAN CONTROL



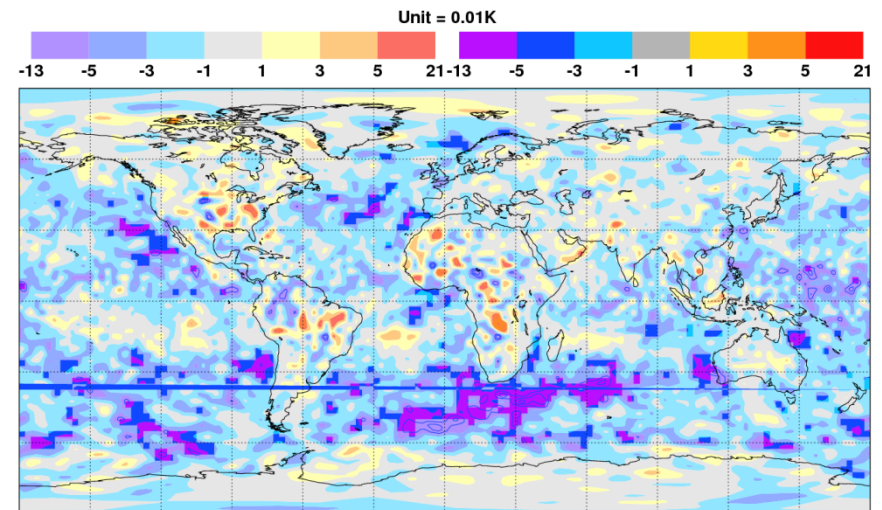
MEAN DUALM-CONTROL



RMS DUALM-CONTROL



STD DUALM-CONTROL



Based on analysis Increments of T1000 20090715 – 20091031. 36R1 with and without DUALM M8.



Future

- Initial tendencies to assess strato-cumulus / DUALM
- Transpose-AMIP to assess strato-cumulus & remote influence of deep convection
- Close collaboration with Physical Aspects Section

ECMWF Vacancy

Consultant (Scientist) position to work in the Predictability and Diagnostics Section on EUCLIPSE Project

<http://www.ecmwf.int/newsevents/employment/en/AP10-29-en.html>

Closing Date: 5 November 2010

Start Date: 1 February 2011



M.J. Rodwell

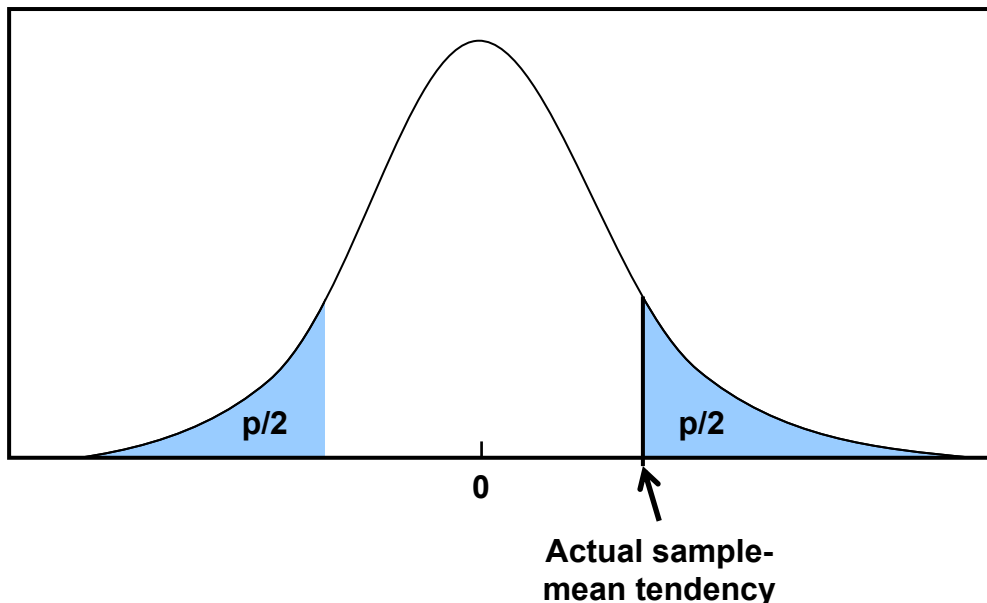
Extra slides

8



Weighting models within a perturbed ensemble

pdf of sample-mean tendency for perfect model



Null hypothesis: “model is perfect”
(i.e. expected net tendency is zero)

p = probability of incorrect rejection of null hypothesis

$p = p(\text{param}, x, y, z)$

Average p over parameters T, q, u, v ,
 Vertically integrate
 Integrate over tropics and extra-tropics

$$p_{\text{PERF}} \equiv p_{\text{TROP}} \times p_{\text{EX-TR}}$$

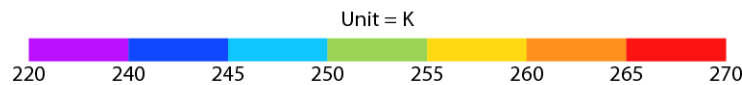
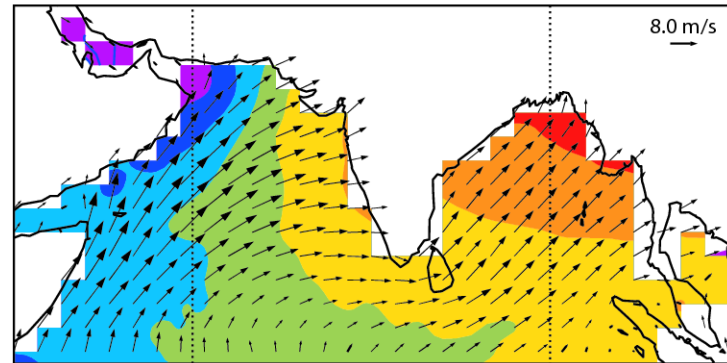
	Probability that model fast physics is perfect
CONTROL	0.20
ENTRAIN/5	0.12

p is a metric that quantifies how good a model’s fast physics (and dynamics) is and could be used to weight models (after being appropriately normalised)

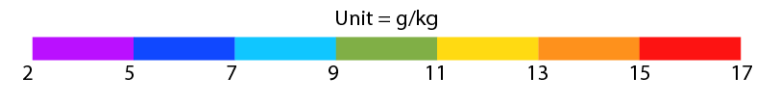
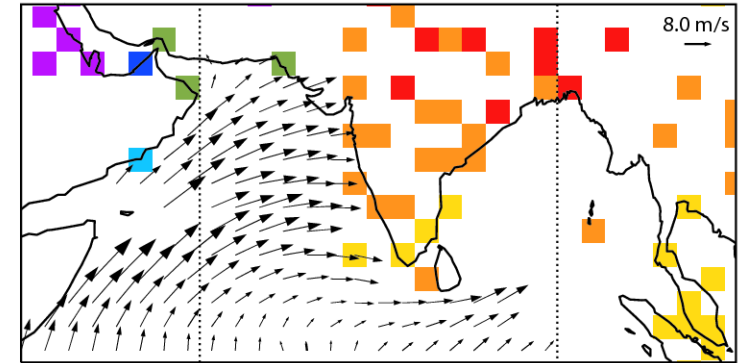


Observations and Departures of low-level v and q

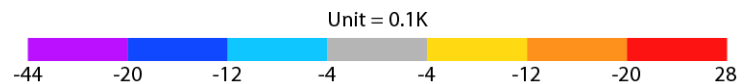
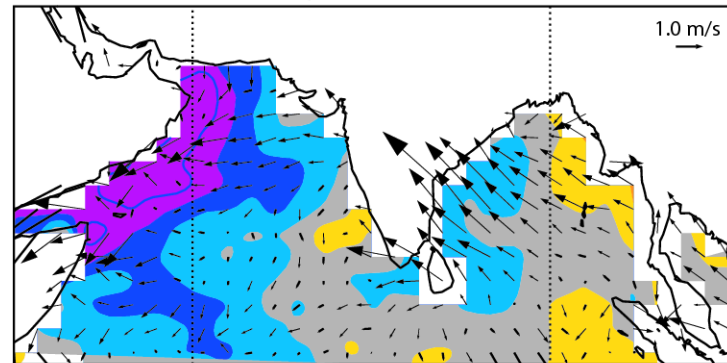
a) ASCAT and SSMI observations



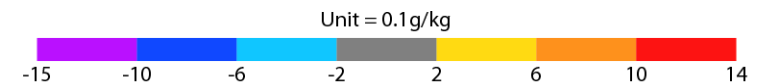
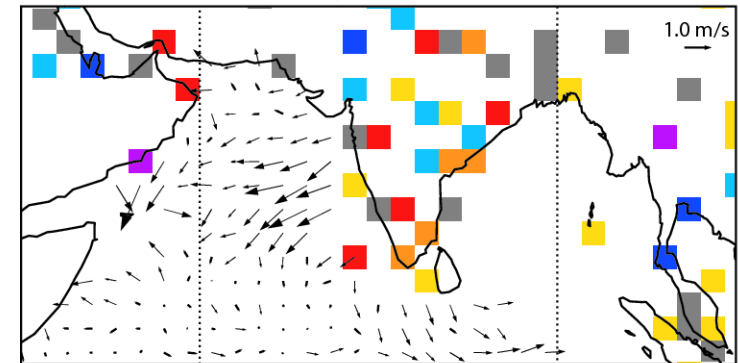
b) AMV and radiosonde observations



c) ASCAT and SSMI FG departures



d) AMV and radiosonde departures



Agreement of departures suggests model issues

Mean JJA 2009 diagnostics of ASCAT surface winds, AMV 950 hPa winds, SSMI brightness temperature (positively correlated with specific humidities at ~850 hPa) and radiosonde specific humidities at 850 hPa