



Precipitation Assimilation

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Acknowledgements:

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Jean-François Mahfouf, Ron Errico



Contents

- Analysis system
- Observation, y
- Observation operator, $H(x)$, linearity, H
- Modelling (representativeness) error, $E + F$
- Background error, B



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At present, incremental 3D/4D-Var systems represent the only affordable approach for global applications (CMC, Met Office, Météo France, ECMWF, NCEP, JMA, NRL, ...).

Dependent of model scale (regional/global) and analysis scheme, the control variable, the regularity/linearity of physical parameterization schemes and the consistency between non-linear and linearized physics are issues that need to be addressed.

Means of making it work: experiment design, length of assimilation window, limited number of included physical processes, spatial resolution, initial conditions, choice of \mathbf{B} , $\mathbf{E}+\mathbf{F}$, \mathbf{y} , \mathbf{x} , \mathbf{J} , H vs. \mathbf{H} , etc., and performance evaluation (what is skill?).

Side-effects: clear-sky observation increments fighting cloud/rain observation increments, too large increments, little impact on dynamics, gravity wave excitation.

Ideally:

- the system works operationally and is affordable,
- a balance is defined between temperature/moisture and dynamics in the analysis,
- and that this balance is well defined inside and outside precipitation,
- \mathbf{B} can be specified or, e.g., a choice can be made that \mathbf{B} is less important (and \mathbf{Q} ?),
- the physical parameterizations used in assimilation system are as consistent as possible (NL, TL, AD),
- the observations (and H) have well behaved sensitivity and $\mathbf{E}+\mathbf{F}$ can be specified.

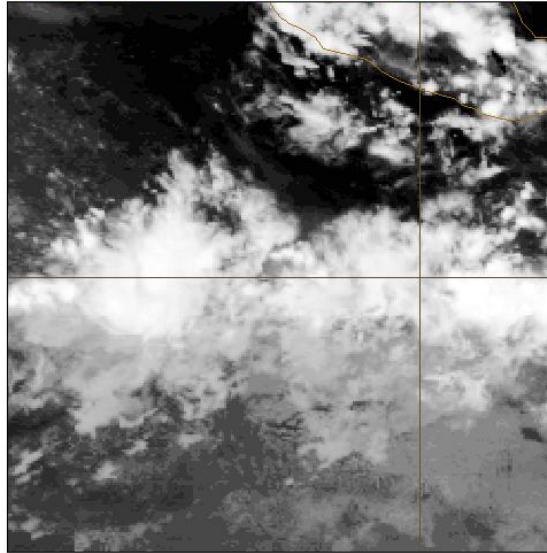


1D-/3D-/4D-Var TCWV/10m-(u,v) increments (DA-Analysis)

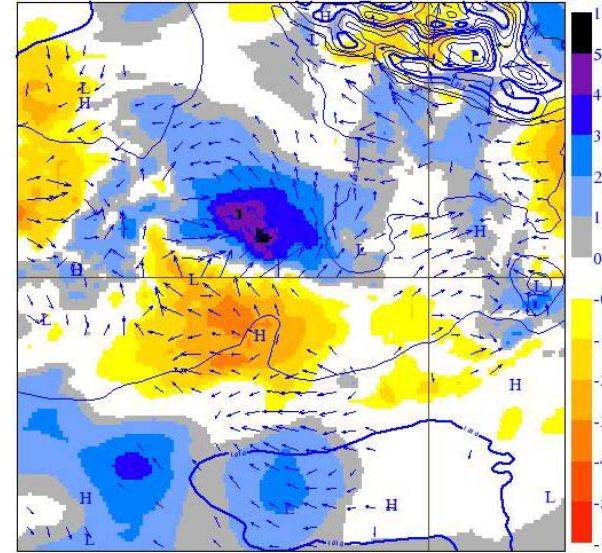
20040801
00 UTC

GOES-12
IR TB

GOES 12 First Infrared Band Sunday 1 August 2004 0000UTC

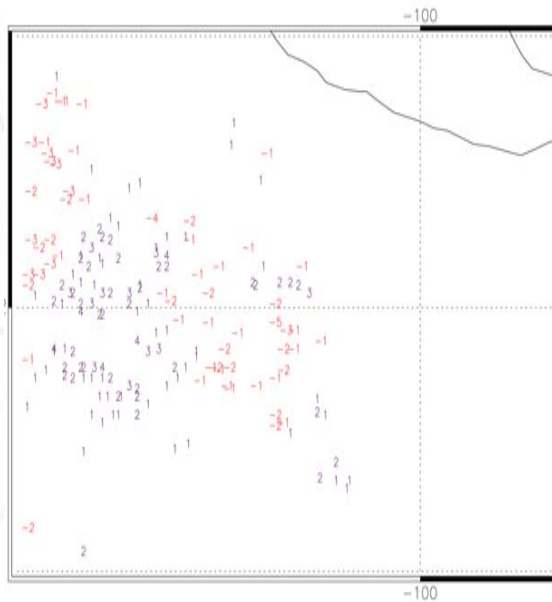


ECMWF Analysis VT: Sunday 1 August 2004 00UTC Surface: mean sea level pressure
ECMWF Analysis VT: Sunday 1 August 2004 00UTC Surface: **total column water vapour/Surf: 10 mtr v

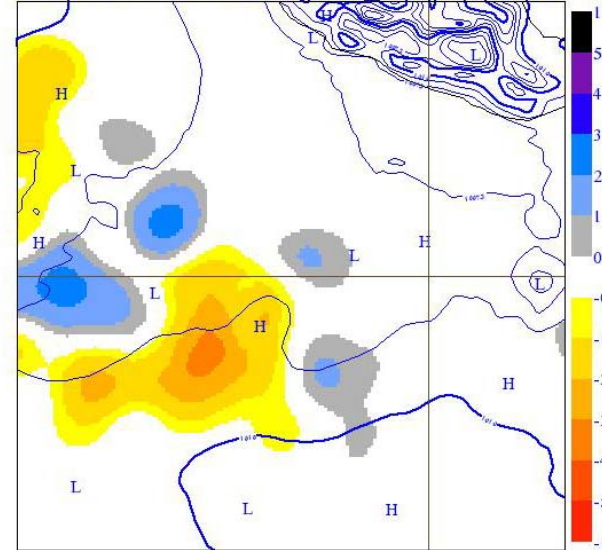


4D-Var
Increment
Differences
(exp-ctrl)

1D-Var
FG-Departures



ECMWF Analysis VT: Sunday 1 August 2004 00UTC Surface: mean sea level pressure
ECMWF Analysis VT: Sunday 1 August 2004 00UTC Surface: **total column water vapour/Surf: 10 mtr v



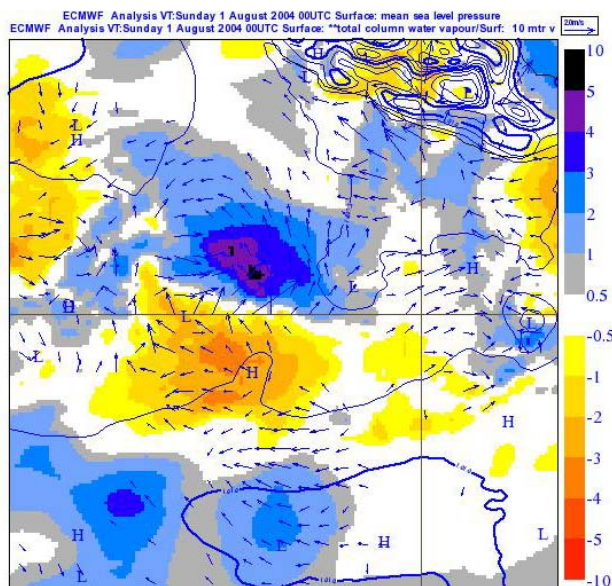
3D-Var
Increment
Differences
(exp-ctrl)



3D-/4D-Var TCWV/10m-(u,v) increments (DA-Analysis): Real vs. fake assimilation of rain affected radiances

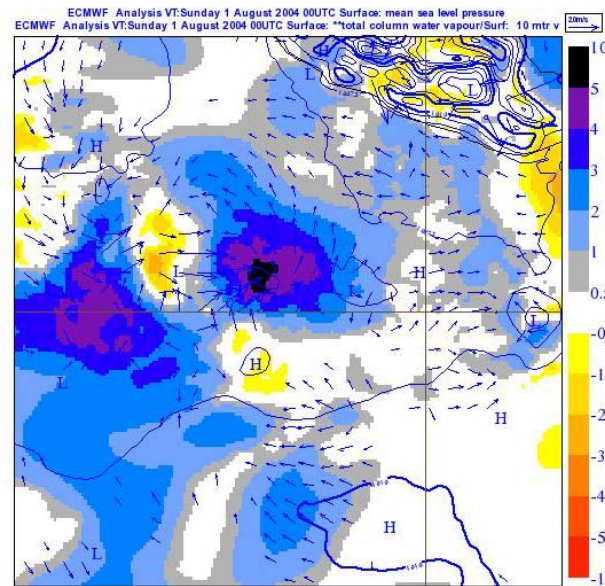
1D+4D-Var:

4D-Var
Increment
Differences
(exp-ctrl)

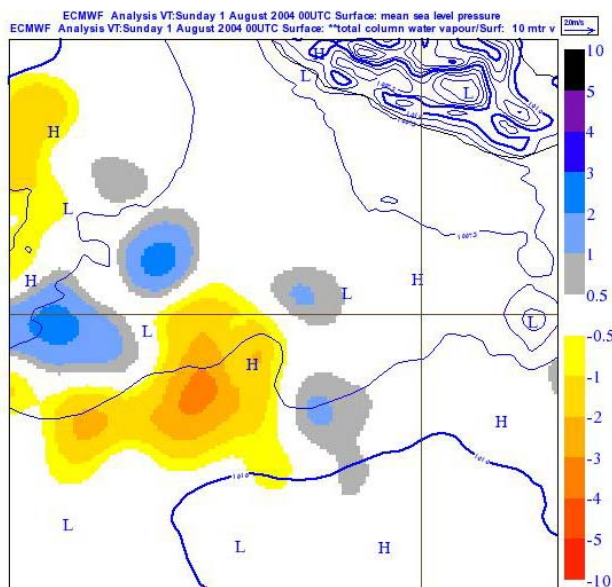


Fake 4D-Var:

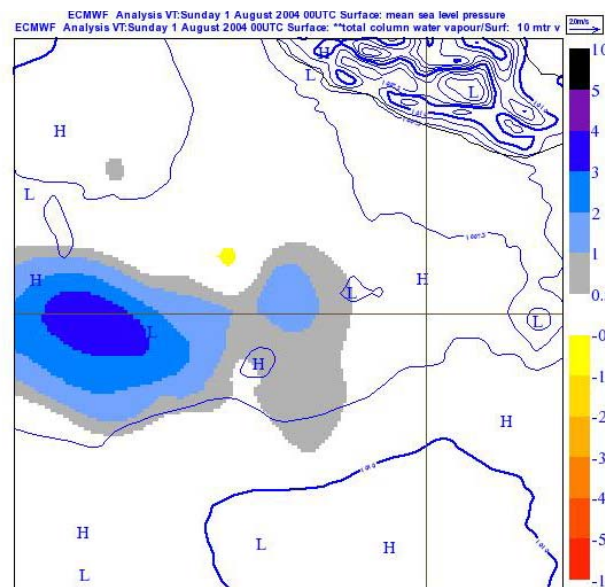
4D-Var
Increment
Differences
(exp-ctrl)



3D-Var
Increment
Differences
(exp-ctrl)



3D-Var
Increment
Differences
(exp-ctrl)





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Observation, y , must provide best trade-off between signal vs. noise (sensitivity) and regularity of response:

Rain gauge: point measurement, type-dependent accuracy, no ancillary information on rain type etc. available, measures rain accumulation (possibly DSD), operational.

Surface radar: limited volume measurement, measures Z (possibly polarization, Doppler-spectrum), operational, accuracy of RR-retrieval depends on rain gauge calibration.

Satellite radar: profile/narrow swath measurement, calibration rather good, measures Z , experimental.

Satellite radiometer: wide swath measurement, calibration good, measures TB that is sensitive to T , q , hydrometeors, (surface emissivity,) channel selection rather flexible, operational.

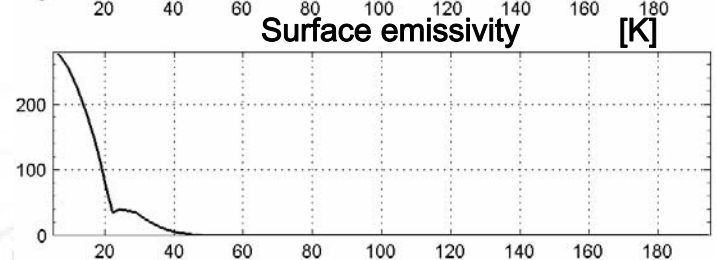
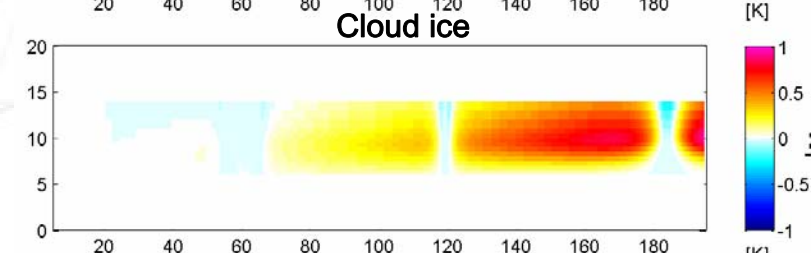
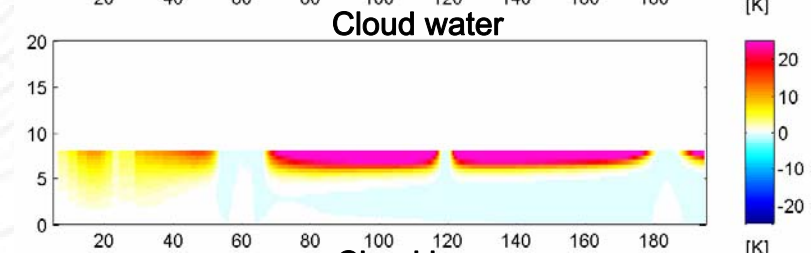
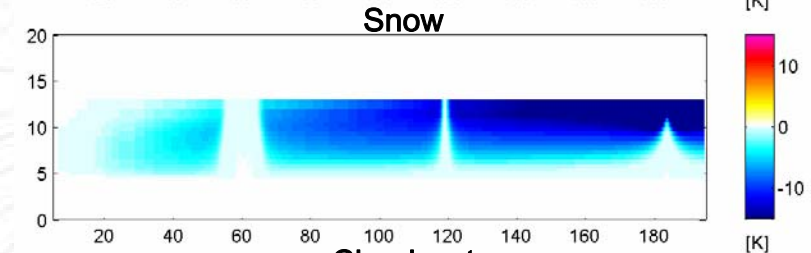
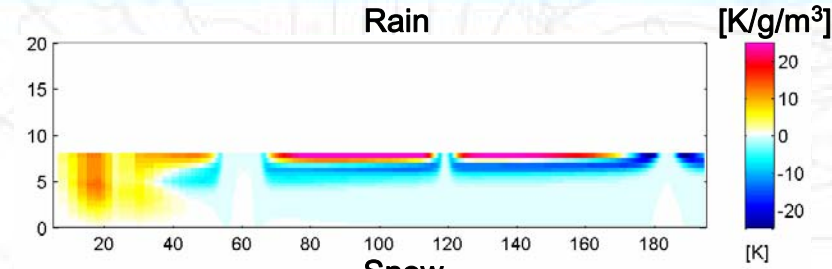
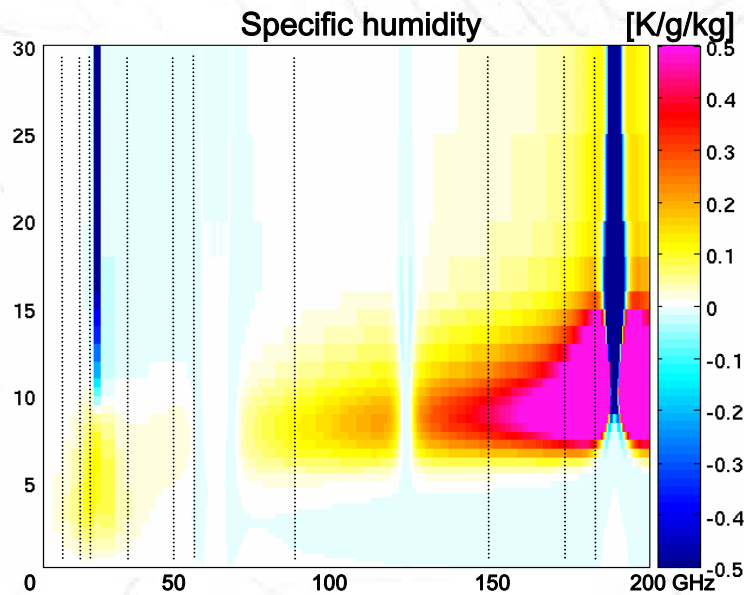
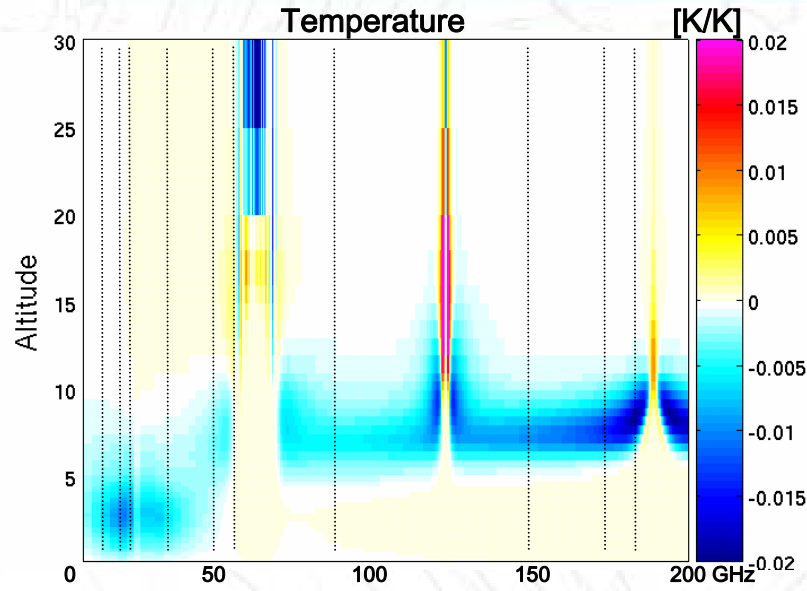
Surface measurements provide less constraints than profile/column measurements.

Rain rate measurements constrain condensation/convection less than simultaneous measurement of rain, cloud water (cover) and moisture.

Rain rate measurements suffer from irregularity (pdf) and 0-value problem (no sensitivity where FG has no rain); derived rain rates suffer from incoherent algorithm performance and mostly unknown error properties.

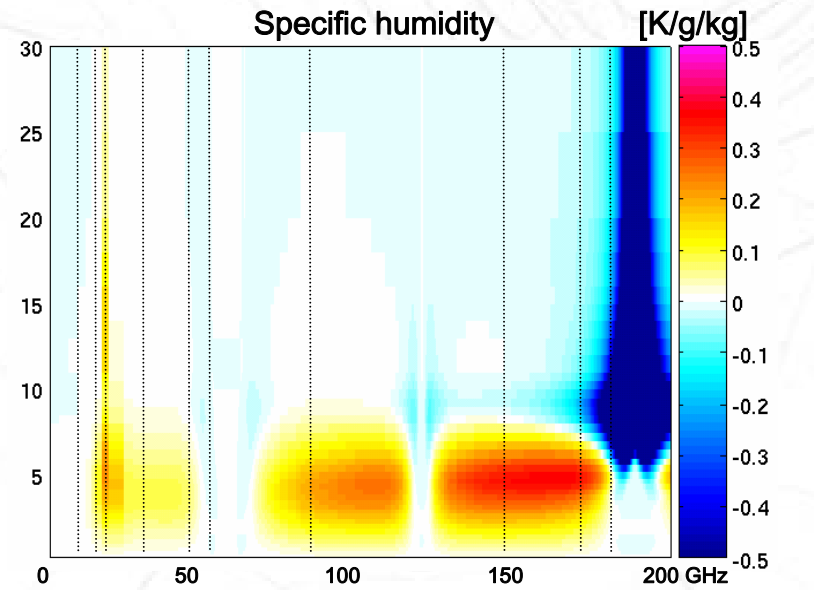
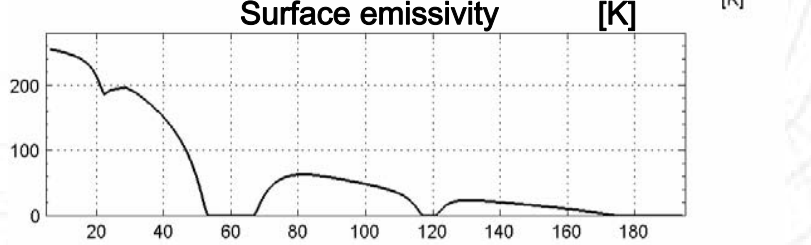
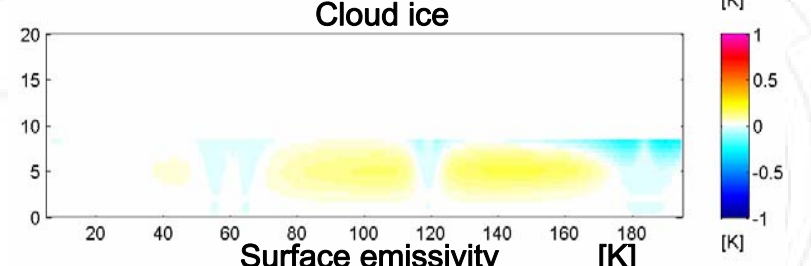
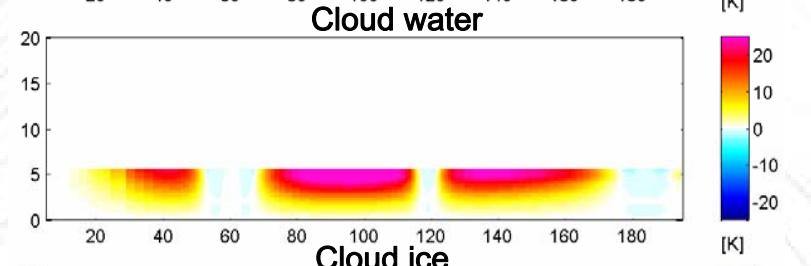
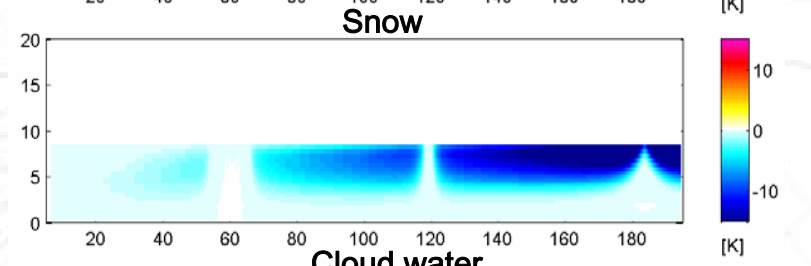
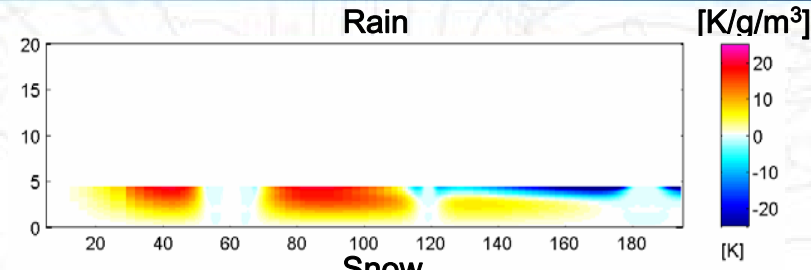
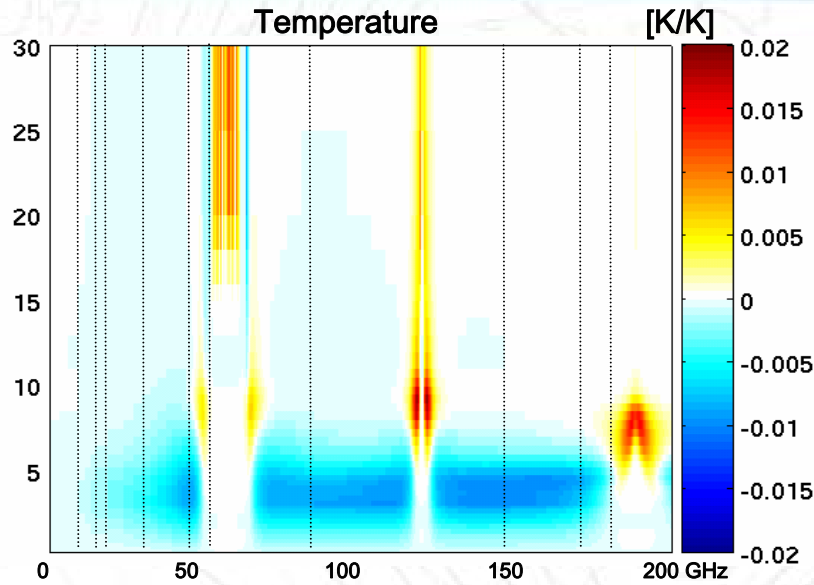


Microwave TB-Jacobians over ocean



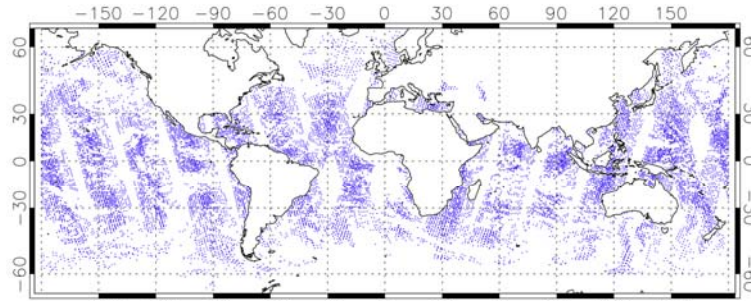


Microwave TB-Jacobians over land



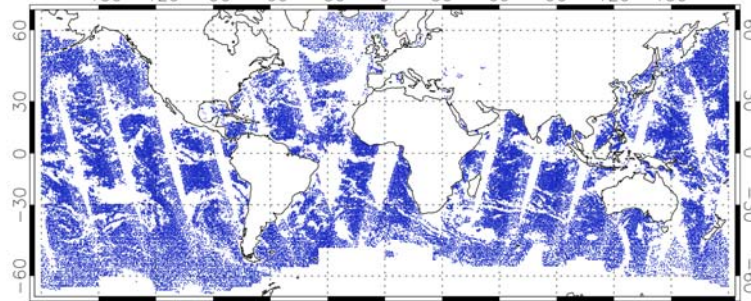


FG SSM/I Radiance Departures (03/08/2004 00 UTC)



← SSM/I clear-sky radiance data coverage

DCDA: 70,000 observations/cycle
DA 50,000 observations/cycle

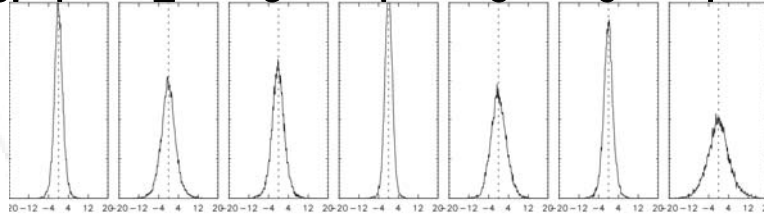


← SSM/I precipitation TCWV data coverage

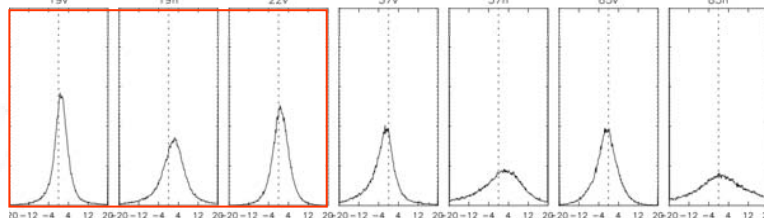
DCDA: 50,000 observations/cycle
DA 25,000 observations/cycle

SSM/I

Channels: 1 2 3 4 5 6 7



← SSM/I clear-sky FG radiance departure pdf's

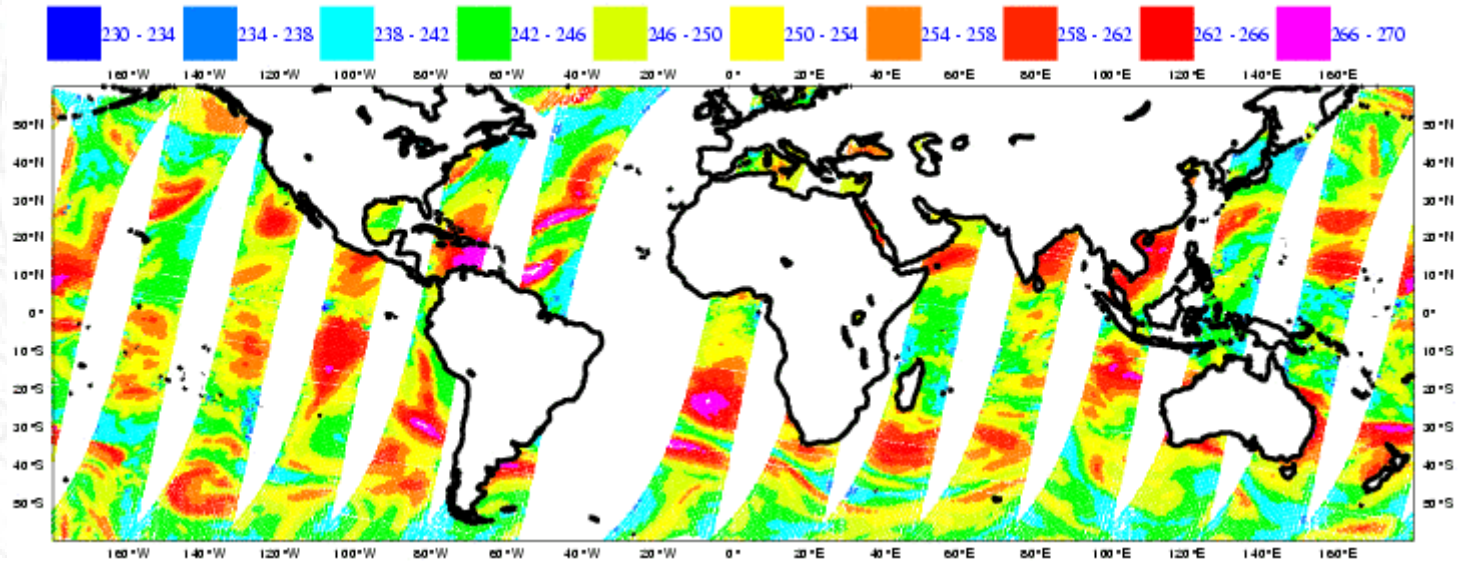


← SSM/I precipitation FG radiance departure pdf's

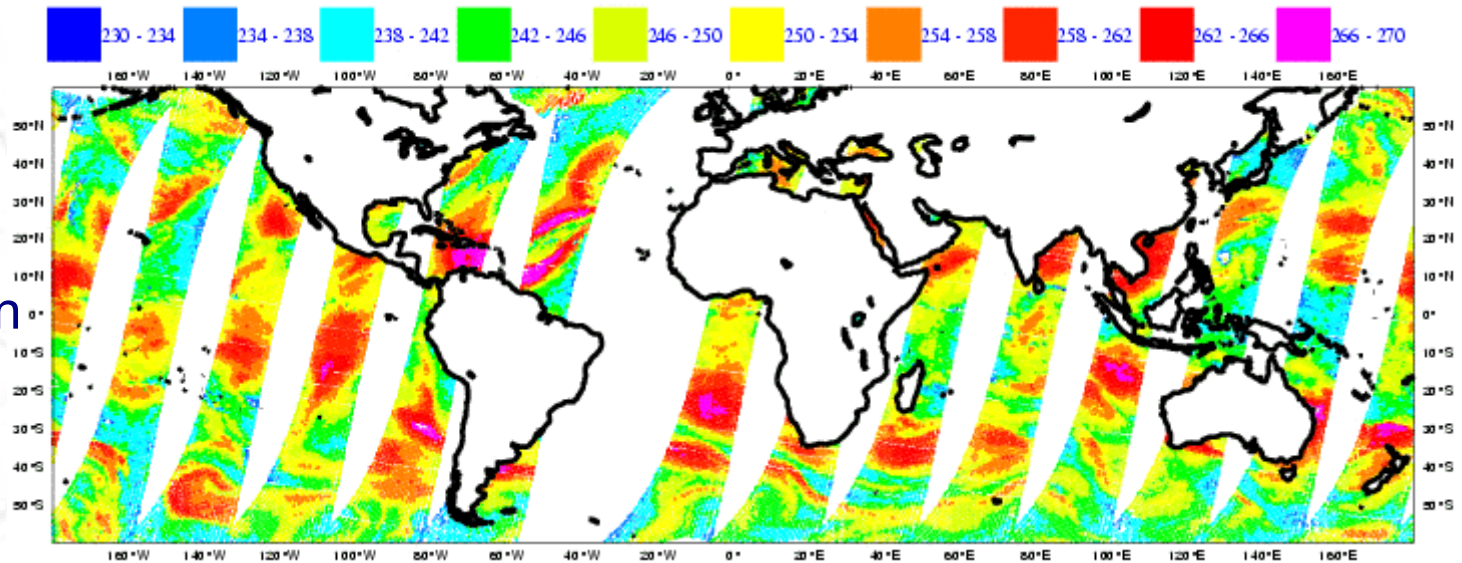


Simulated vs. observed AMSU-B 183±7 GHz TB's

Model

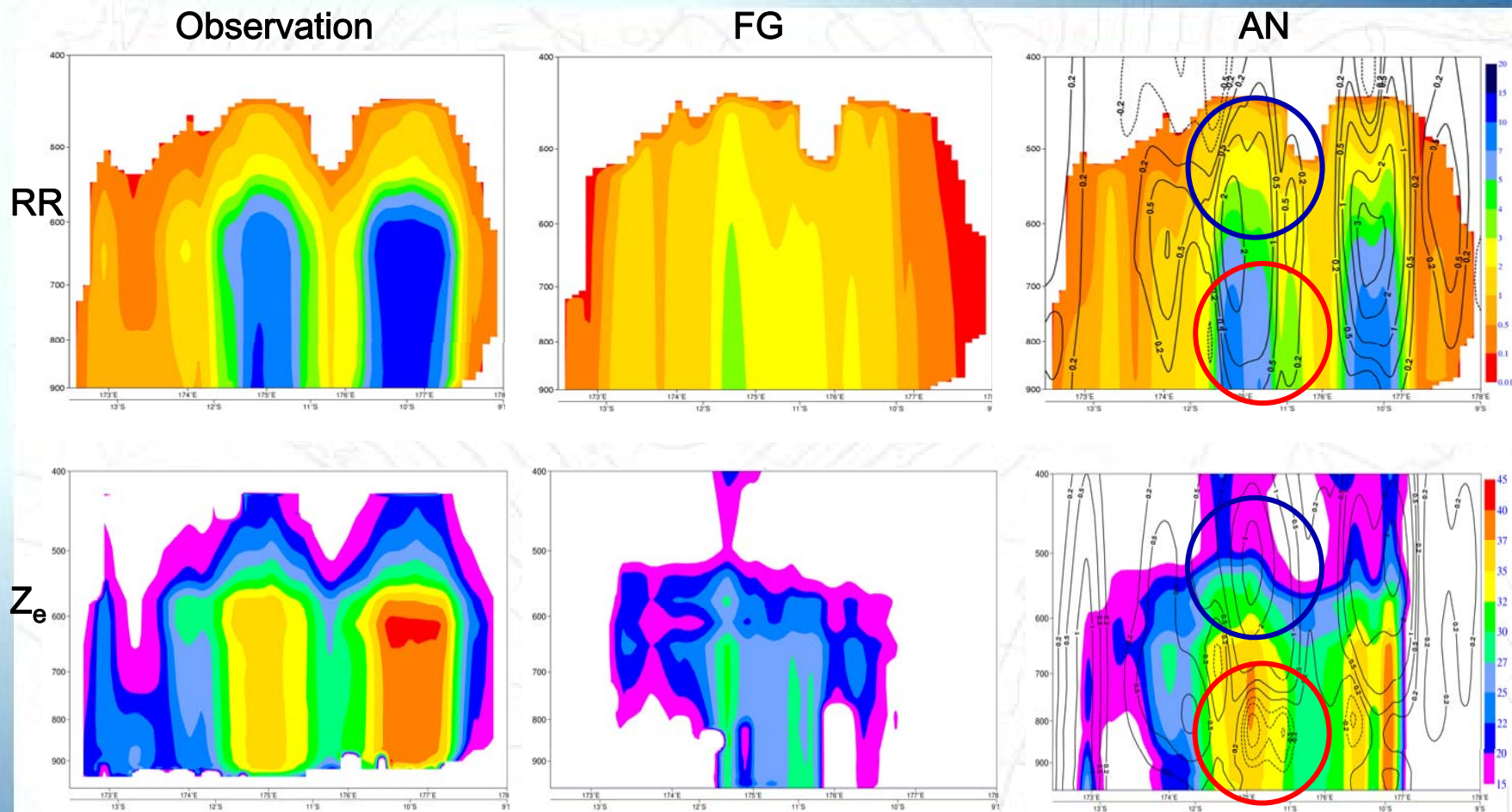


Observation





1D-Var Assimilation of TRMM PR RR vs. Z_e



Tropical Cyclone Zoe (26 December 2002 @1200 UTC)

Vertical cross-section of rain rates (top, mm h⁻¹) and reflectivities (bottom, dBZ), observed (left), background (middle), and analyzed (right).

Black isolines on right panels = 1D-Var specific humidity increments [g/kg].



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Observation operator: Moist Physics (convection)

Linearized parameterization of convection:

- based on Tiedtke (1989)
- single closure assumption based on CAPE
- uncoupling of formulation for convective mass flux and updraft
- unified entrainment/detrainment formulation
- forward/tangent-linear/adjoint models

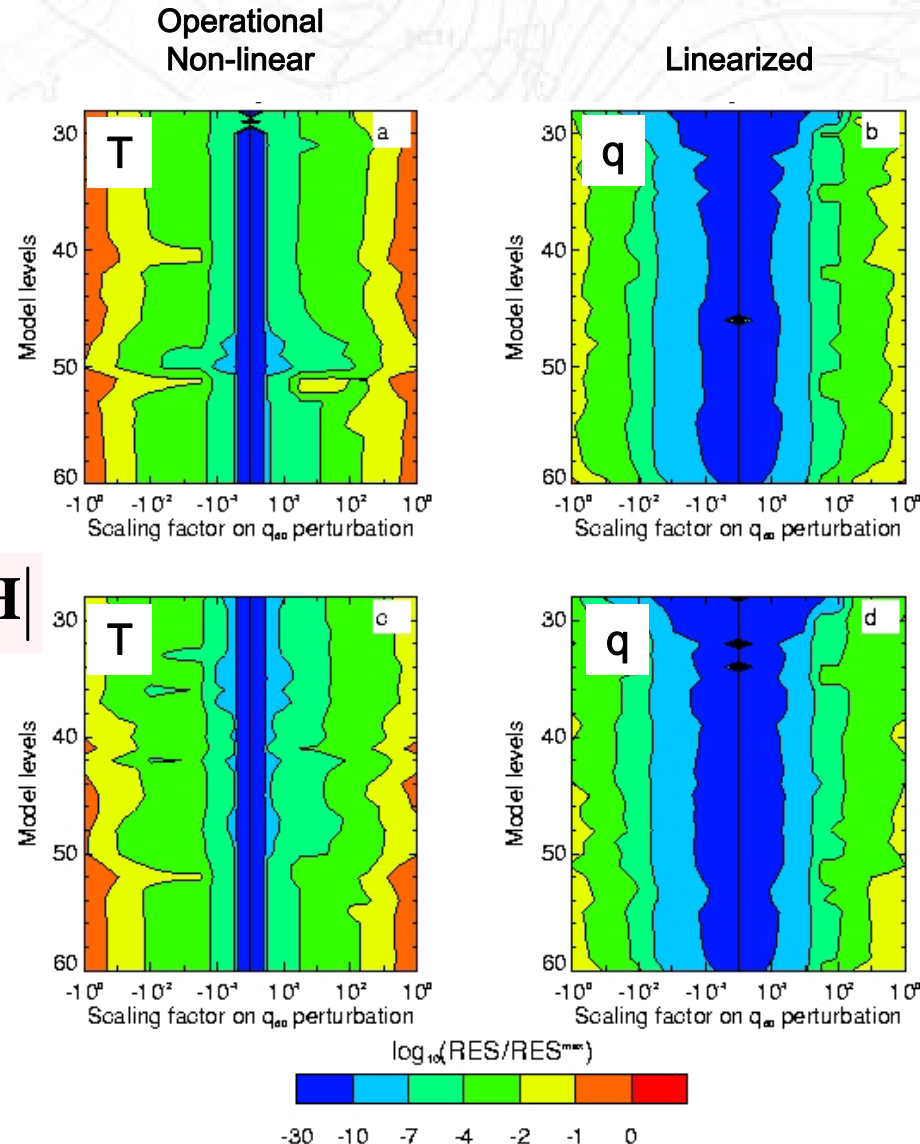
Issues: Linearity

$$RES = |H(\mathbf{x} + \lambda\delta\mathbf{x}) - H(\mathbf{x}) - \lambda\delta\mathbf{x}\mathbf{H}|$$

- H: non-linear observation operator
- H: TL of H
- \mathbf{x} : state vector
- $\delta\mathbf{x}$: perturbation
- λ : scaling factor

Lopez and Moreau (2005)

Response of (T,q)-tendencies to perturbations of q:





Observation operator: Radiative Transfer

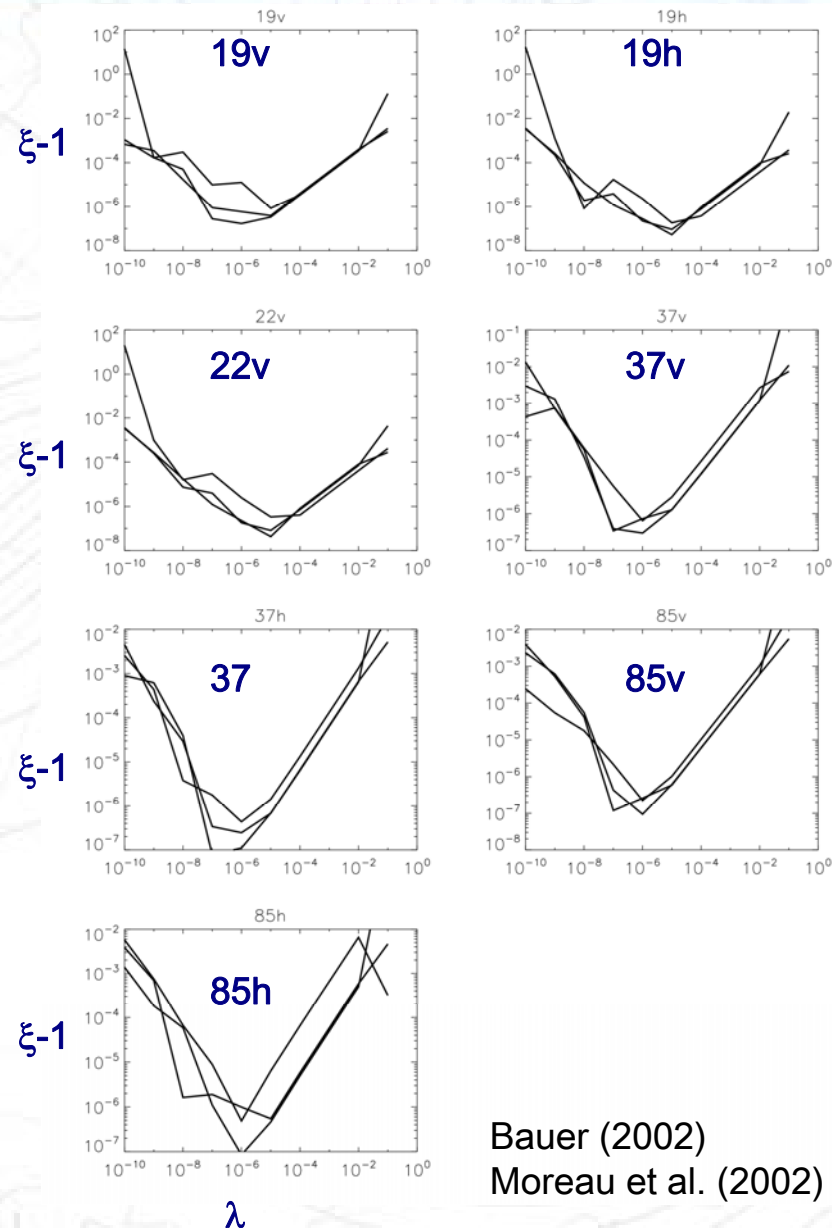
Multiple scattering radiative transfer:

- δ -Eddington solution (Eddington's 2nd approximation)
- plane-parallel with clear and cloud/rain column
- forward/tangent-linear/adjoint models
- part of NWP-SAF RTTOV-8 package
- atmospheric absorption / surface emission from RTTOV

Issues: Linearity

$$\xi - 1 = \frac{H(\mathbf{x} + \lambda \delta \mathbf{x}) - H(\mathbf{x})}{\lambda \delta \mathbf{x} \mathbf{H}}$$

- H: non-linear observation operator
- H: TL of H
- \mathbf{x} : state vector
- $\delta \mathbf{x}$: perturbation
- λ : scaling factor



Bauer (2002)
Moreau et al. (2002)



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E+F = modelling (representativeness) error + measurement error covariance:

- magnitude and quantifiability depends crucially on choice of y (H , and H vs. H)
- moist physical parameterizations may constitute biggest contribution:
 - multi-model ensembles, ensemble analysis, EnKF
 - single-column model forecasts with IC/BC from (accurate) observations
- bias-correction

SSM/I

Clear-sky

Clouds + precipitation

	19v	19h	22v	37v	37h	85v	85 h
<i>Clear sky:</i>							
Mean FG-departure	0.2	0.2	0.1	0.1	0.3	0.0	0.2
FG-Departure standard deviation	1.8	3.1	2.7	1.6	3.1	2.1	4.6
Mean AN-departure	0.1	0.1	-0.1	0.1	0.2	0.0	0.1
AN-Departure standard deviation	1.3	2.0	1.9	1.2	2.4	1.7	3.6
Observation error : E+F	3.0	4.5	4.0	3.5	4.0	4.0	6.0
Mean background error : HBH^T	1.7	3.1	2.7	1.4	2.7	1.7	4.0
<i>Precipitation:</i>							
Mean FG-departure	0.3	0.5	0.5	0.4	0.8	0.5	1.1
FG-Departure standard deviation	4.2	7.7	3.9	6.4	13.3	5.8	13.0
Mean AN-departure	0.1	0.1	0.1	0.1	0.0	0.3	-0.2
AN-Departure standard deviation	1.0	1.7	1.1	3.0	5.9	4.9	9.1
Observation error : E+F	4.0	6.0	4.0	-	-	-	-
Mean background error : HBH^T	4.1	7.5	4.8	-	-	-	-

$$\sigma_{E+F}^2 \neq \sigma_{\Delta FG}^2 - \sigma_{HBH^T}^2$$



Observational Method

Hollingsworth and Lönnberg (1986):

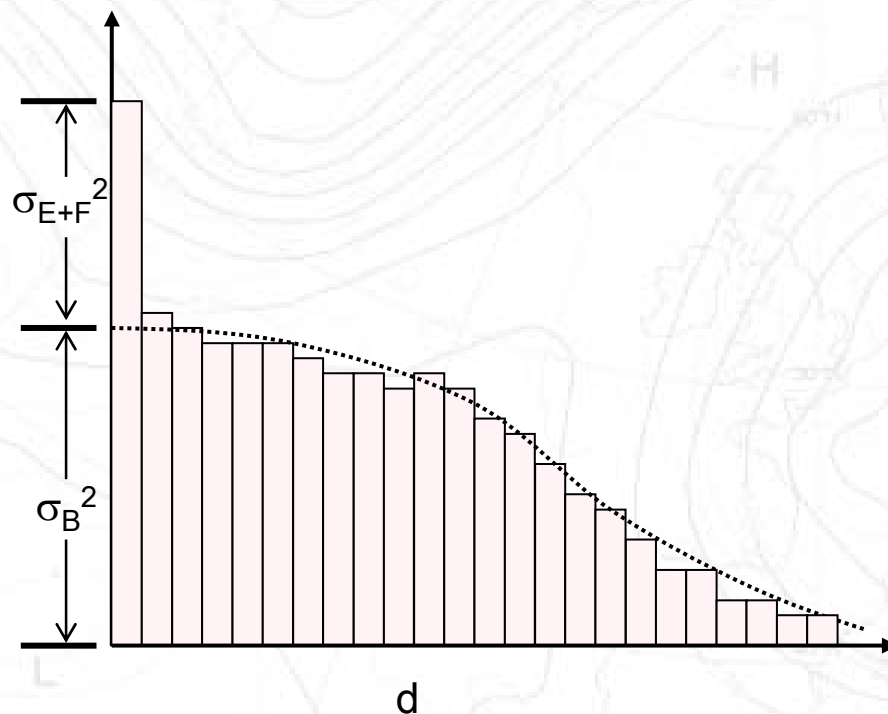
Estimation of σ_B vs. σ_{E+F} from spatial covariance structure of first-guess departures

Assumptions:

- observation network is rather dense
- observations are spatially uncorrelated (and discrete)

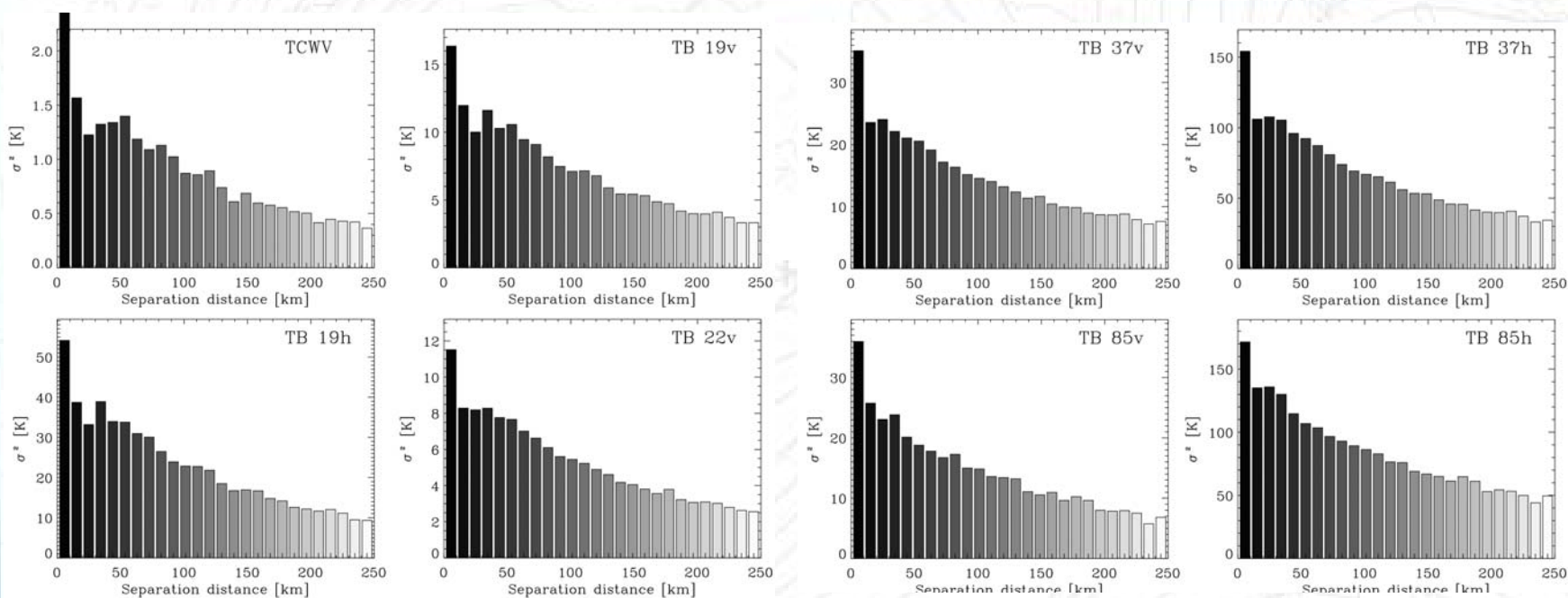
Conclusions:

- at 0-separation distance, the variance is $\sigma_B^2 + \sigma_{E+F}^2$
- at >0-separation distance, the variance is covariance $\sigma_B^2(d)$





Observational Method: TCWV/TB in Precipitation



	σ_B	σ_{E+F}	
TCWV	1.5	1.5	kg/m ²
TB 19v	3	2.5	K
TB 19h	6	4.5	K
TB 22v	3	2	K
TB 37v	5	3	K
TB 37h	10	7	K
TB 85v	5	3.5	K
TB 85h	11	6	K

$$\sigma_{E+F}^2 = \sigma_{\Delta FG}^2 - \sigma_{HBH^T}^2$$



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B determines:

- weight of background,
- structure of increments,
- interaction between temperature/moisture and, e.g., dynamics.

Profile of increments \neq observation (weighting function) [x moist physics] sensitivity

Shape of **B** is not precipitation specific:

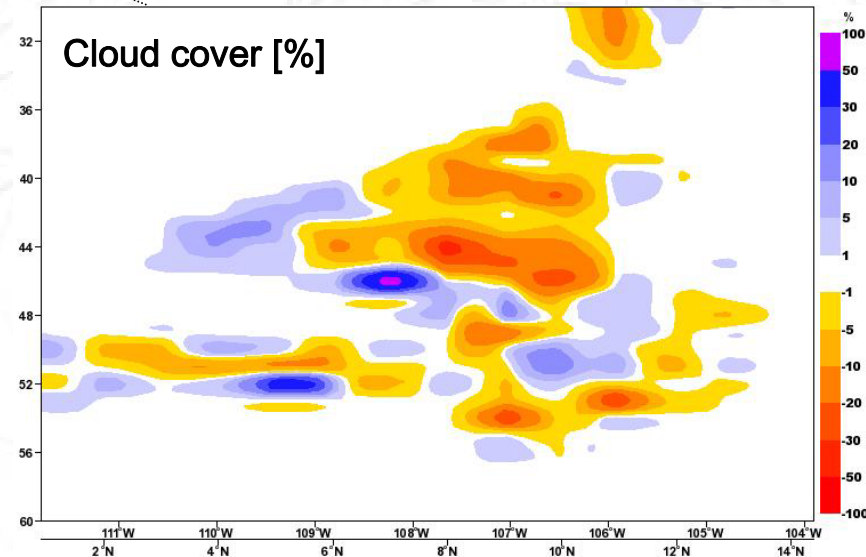
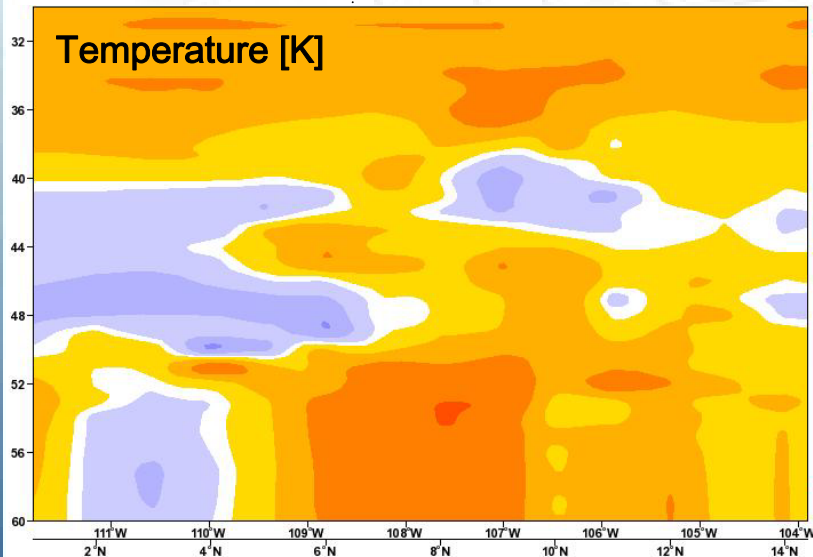
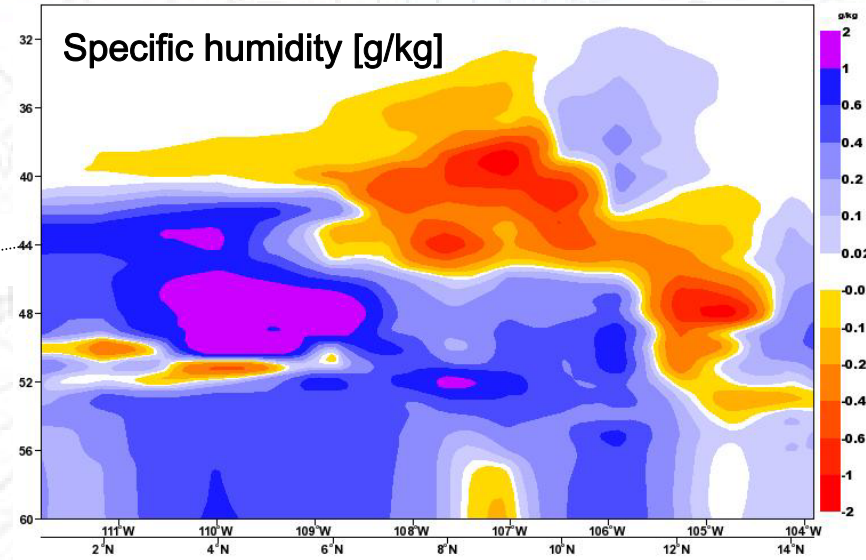
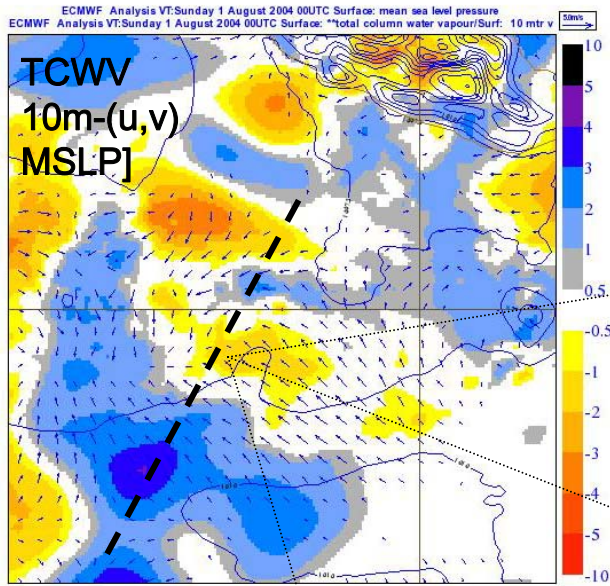
- σ_b profile drives increments similarly inside/outside precipitation
- spread of information due to horizontal structure functions may be disproportional

Definition of **B** may include other mechanisms (e.g. limitation of increments in dry environments)

Development of more specific formulations required: wavelets, ENKF etc.



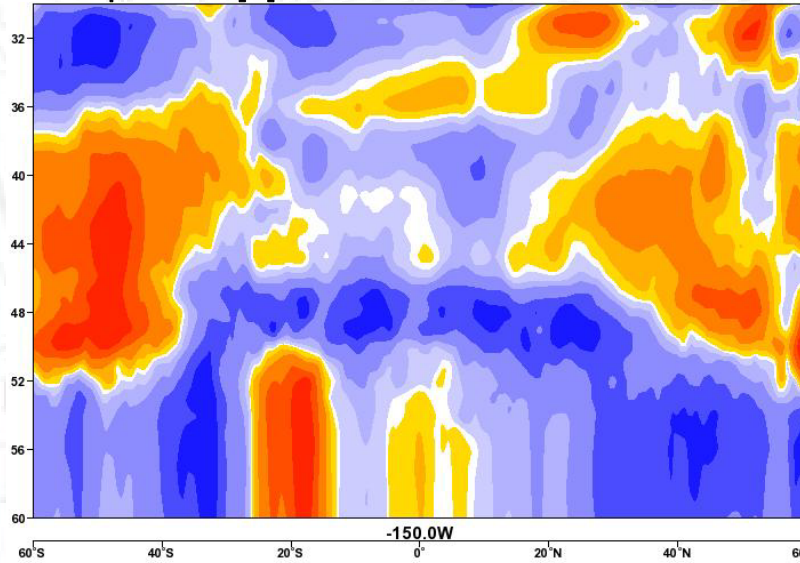
4D-Var increment cross-section (DA-Analysis)



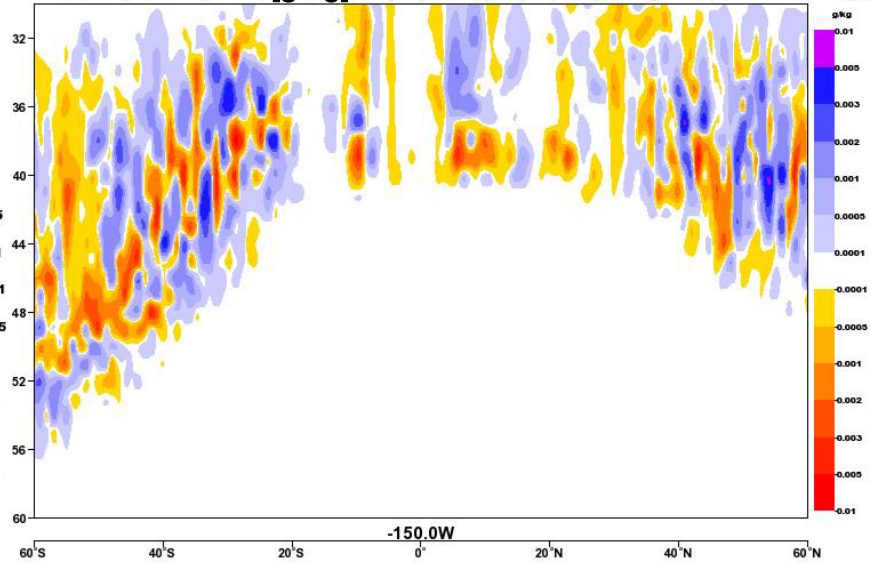


4D-Var mean increment cross-section (DA-Analysis) 200408 - 150° W

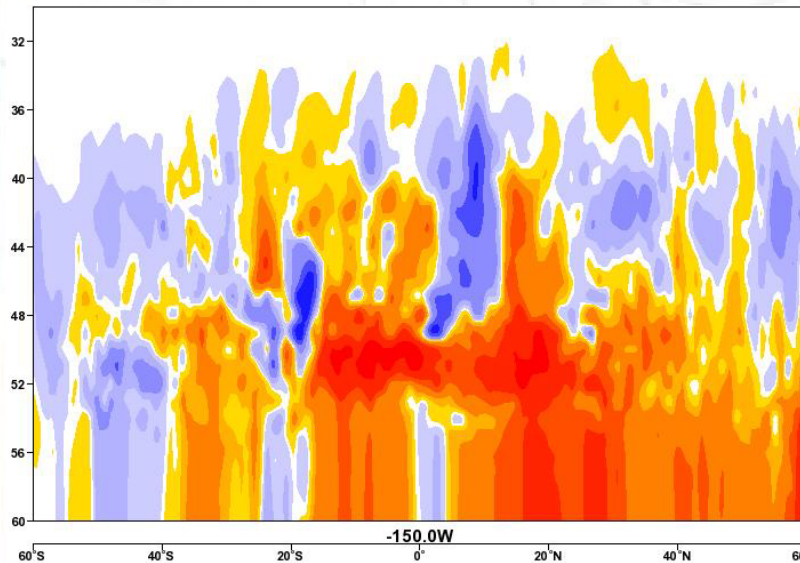
Temperature [K]



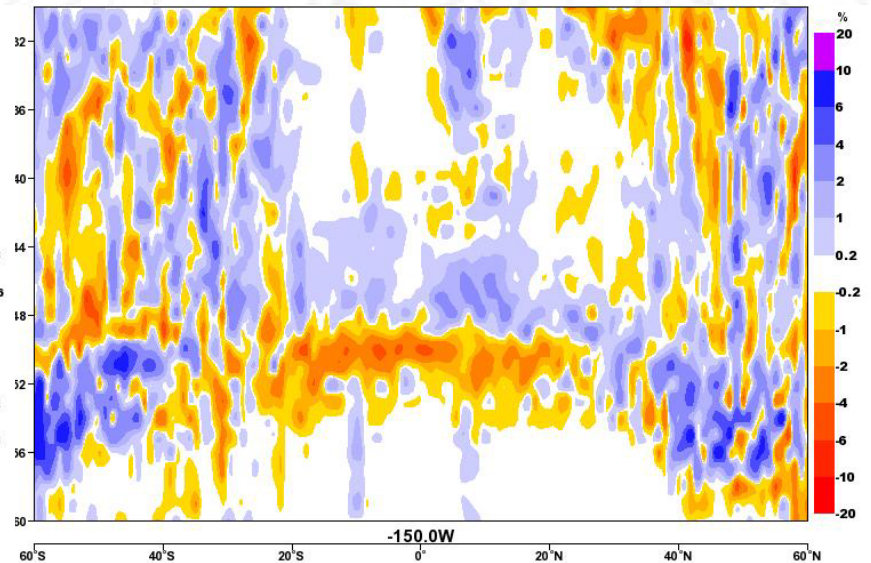
Cloud water/ice [g/kg]



Specific humidity [g/kg]



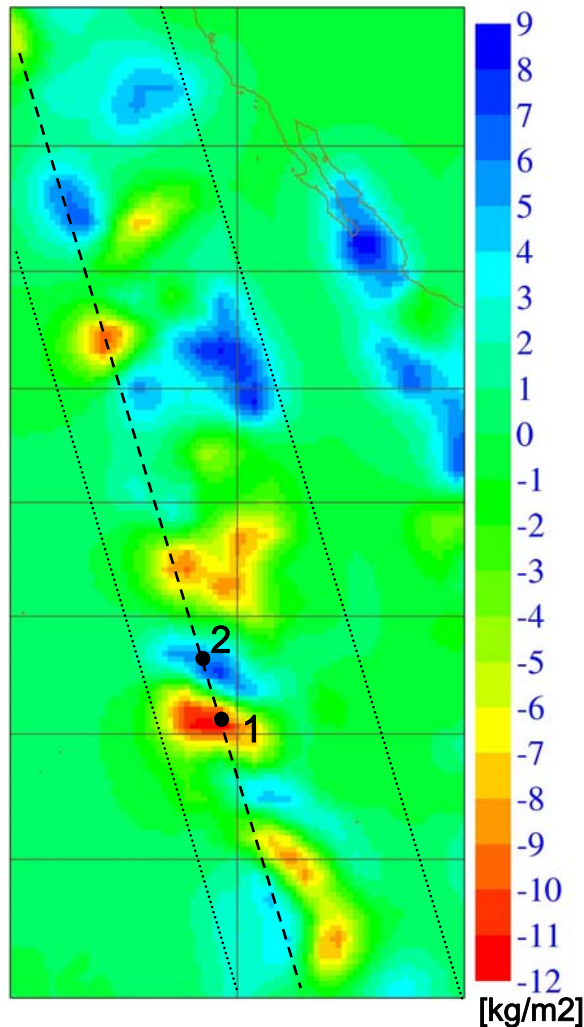
Cloud cover [%]



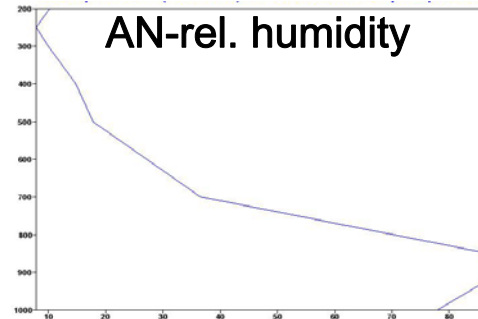
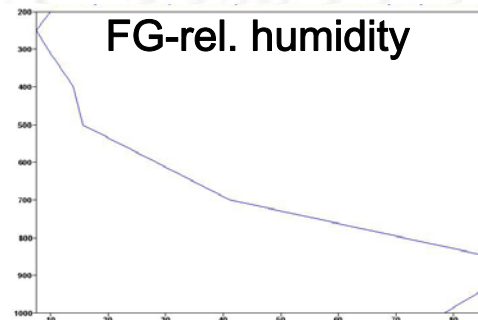
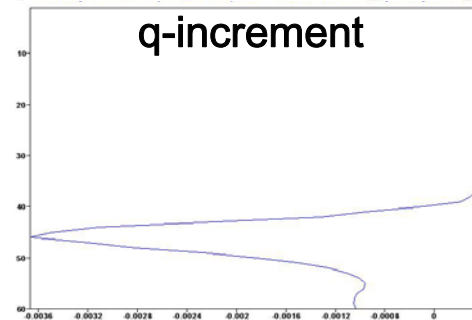


Example q-increment profiles

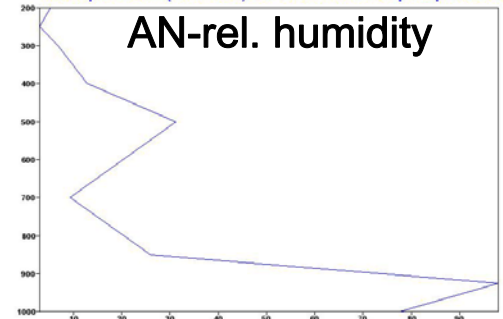
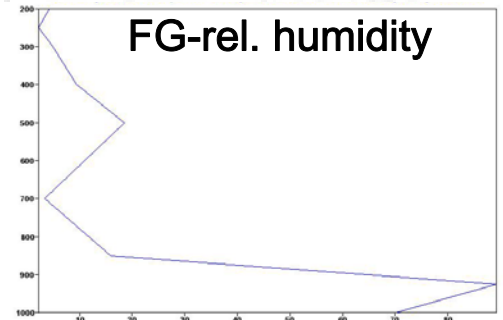
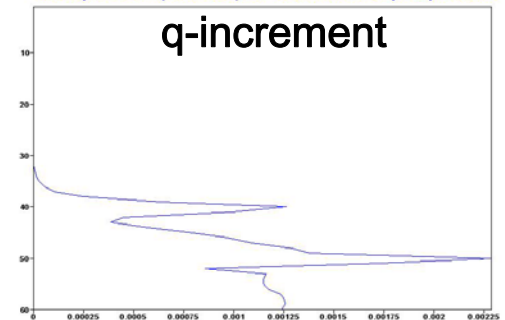
SSM/I 3D-Var TCWV increments
from SSM/I clear-sky radiance
assimilation



Profile 1



Profile 2

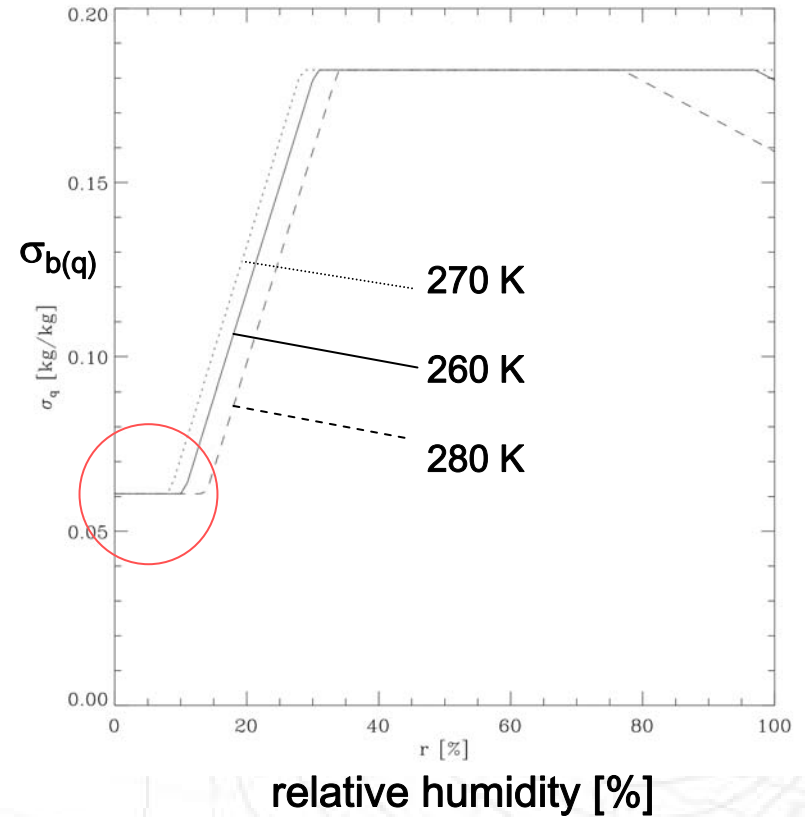
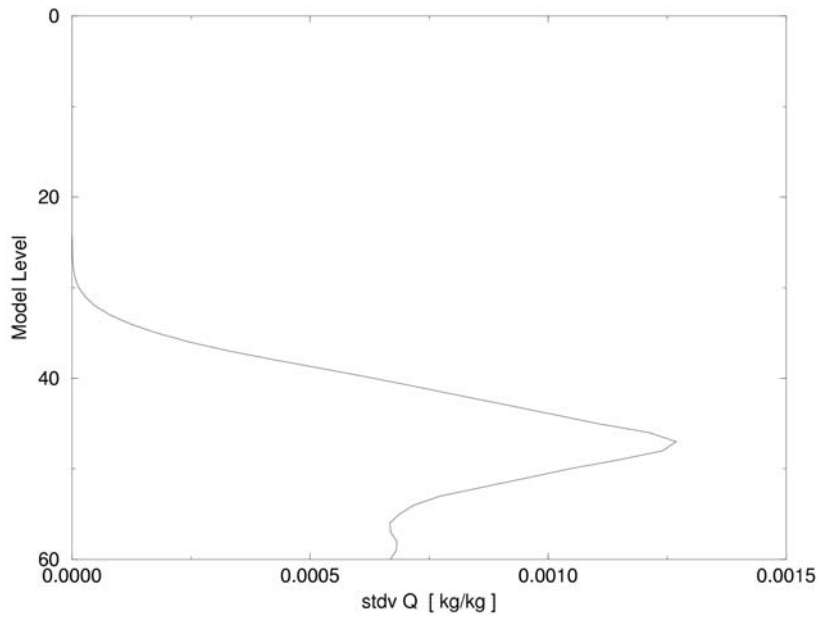




Increments vs. Background errors

σ_b (RH, T) formulation

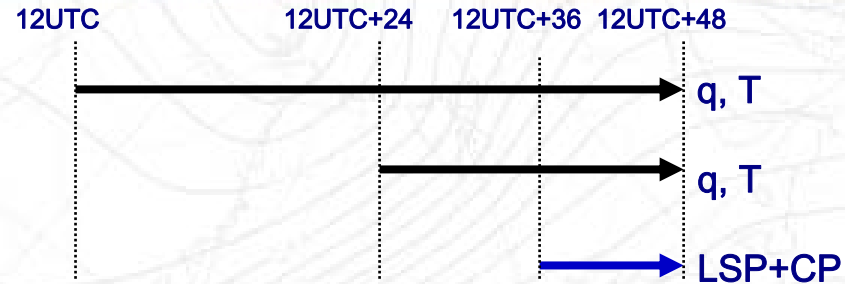
Typical σ_b profile



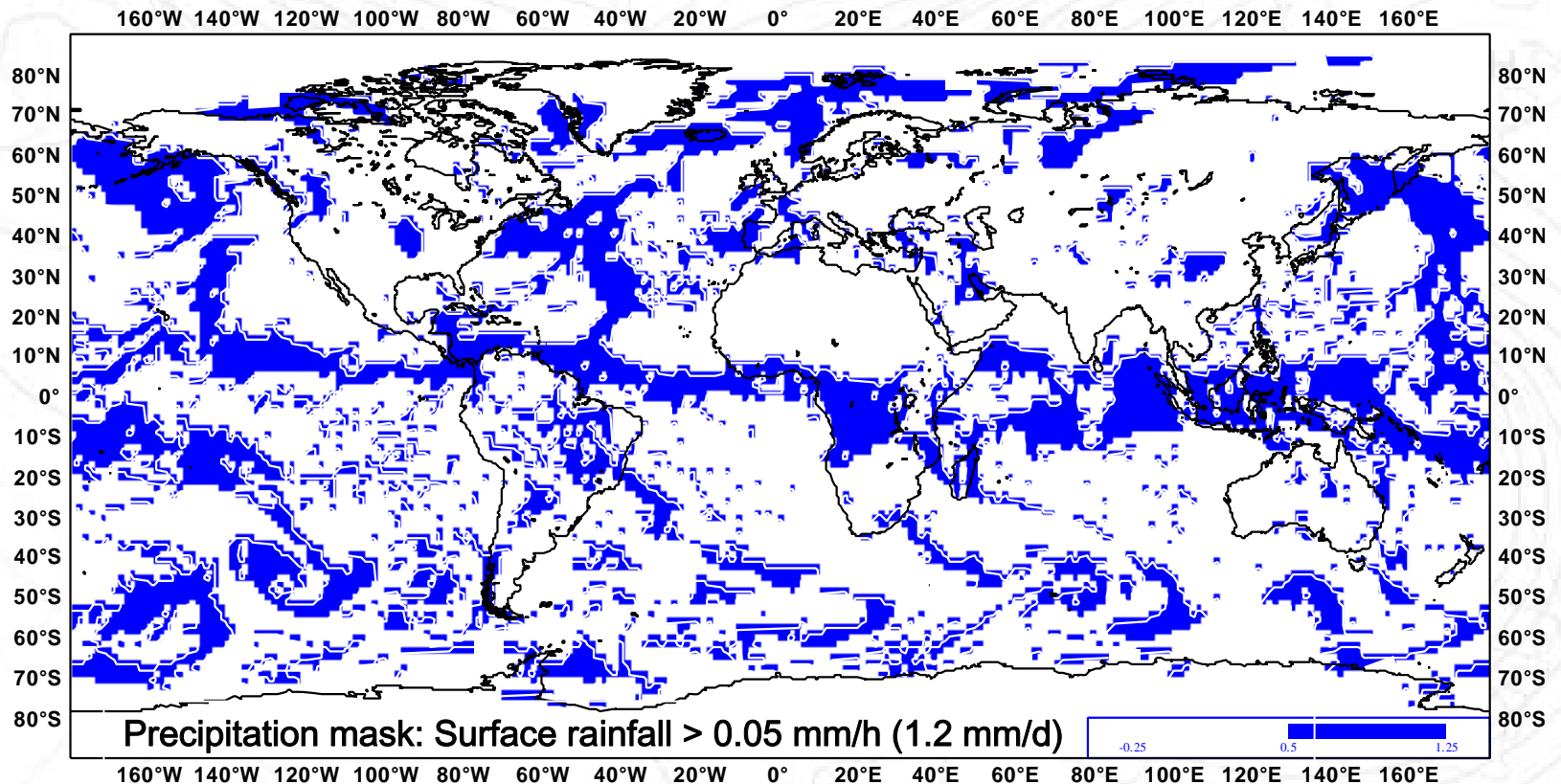


BG-error vertical correlations and error standard deviations

NMC-method:
Statistically, the difference between 48h and 24h forecasts corresponds to forecast errors, i.e. FG-errors of next forecast



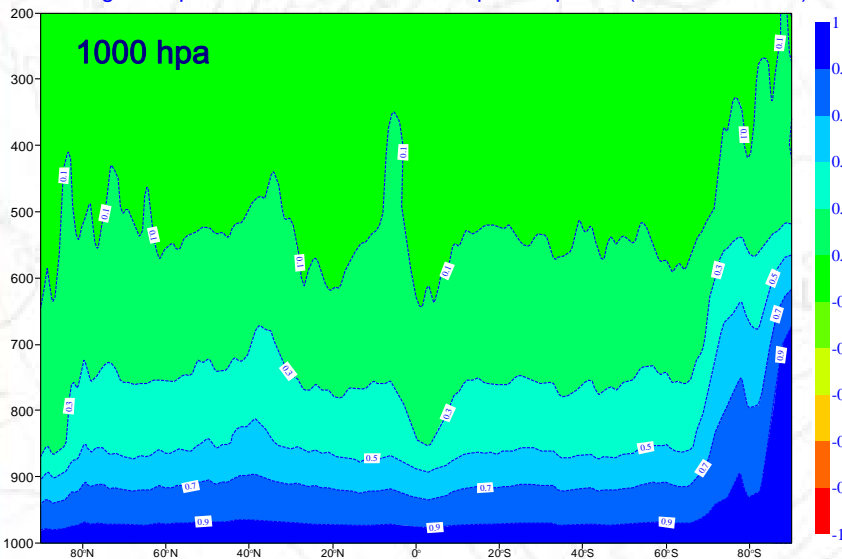
Wednesday 3 December 2003 00UTC ECMWF Forecast t+12 VT: Wednesday 3 December 2003 12UTC Surface



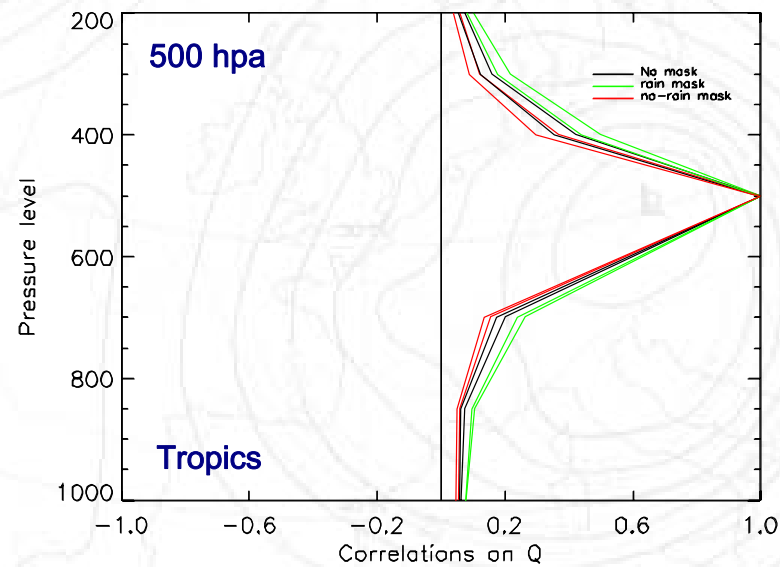
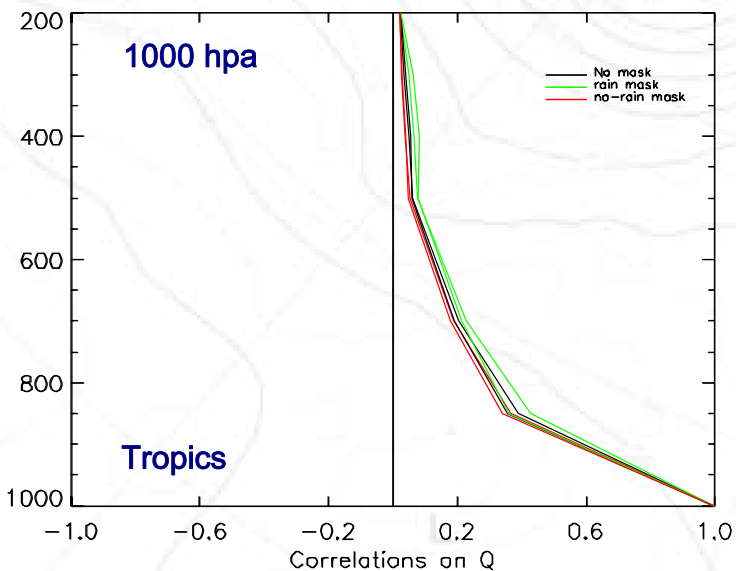
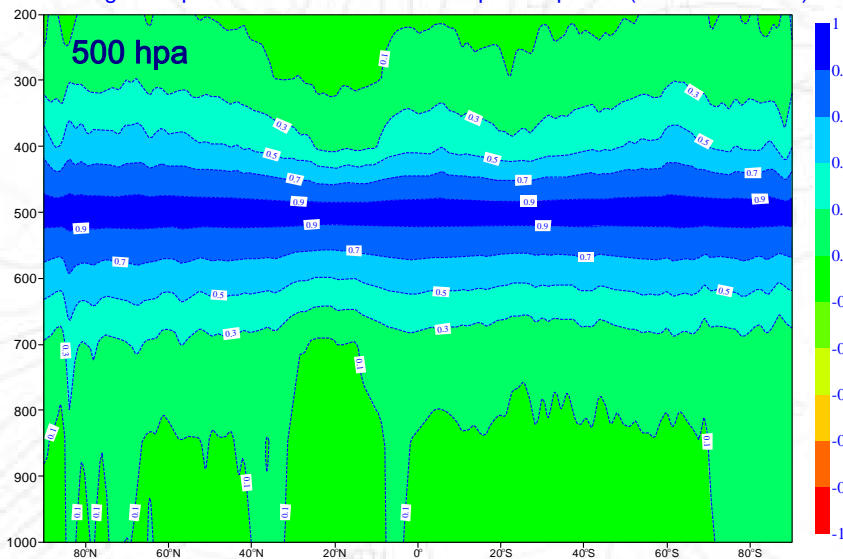


Zonal means of vertical correlations for specific humidity (DJF 2004, global)

Average of spec hum 20031202 1200 step 24 Expver 1 (180.0W-180.0E)



Average of spec hum 20031202 1200 step 24 Expver 1 (180.0W-180.0E)





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Issues for Discussion

In view of the importance of departure-pdf's (and error pdf's) what is the optimum choice of control variable (e.g. total water content) and precipitation-related observation (e.g. microwave TB)?

Biases: how can systematic model (observation operator) errors be reduced (1st choice: bias-correction schemes) and separated from observation biases?

Non-linearity of observation operators/convergence - how non-linear is it and can we produce criteria for avoiding the problematic situations?

How can (moisture) background errors be defined in precipitation (this includes horizontal structure functions)? Longer-window 4D-Var to reduce importance of B for analysis (loss of memory but model error contribution, Q , increases)? Ensemble Kalman Filter, i.e., perturb f/c's for cloud profiles to have more specific adaptation of local conditions?

Design experiment for quantifying representativeness error (E+F).

How can the link between moisture analysis and dynamics be improved (through B and/or 4D-Var and/or otherwise)?

Experiment evaluation - Since we deal with variable, inhomogeneous, synoptic-system-rather-than-large-scale-wave-pattern parameters, do we need other ways of scoring precipitation assimilation performance?