

**Physical processes in adjoint models:
potential pitfalls and benefits**

Marta JANISKOVÁ
ECMWF

marta.janiskova@ecmwf.int

Thanks to: P.Lopez, F. Chevallier A. Benedetti, J.-N. Thépaut, M. Leutbecher



Linearized models in NWP

- **different applications:**
 - **variational data assimilation** ← *incremental 4D-Var at ECMWF*
 - **singular vector computations** ← *initial perturbations for EPS*
 - **sensitivity analysis** ← *forecast errors*

- **first applications with adiabatic linearized model**

- **nowadays, including the physical processes in the linearized model**

Linearized model with physical processes

Including physical processes can:

- **in variational data assimilation:**

- reduce spin-up
- provide a better agreement between the model and data
- produce an initial atmospheric state more consistent with physical processes
- allow the use of new (satellite) observations (*rain, clouds, soil moisture, ...*)

- **in singular vector computations:**

- help to represent some atmospheric features
(*processes in PBL, tropical instabilities, development of baroclinic instabilities, ...*)

- **in sensitivity analysis:**

- allow a reduction of forecast error

- **adjoint of physical processes can also be used for:**

- model parameter estimation
- sensitivity of the parametrization scheme to input parameters

Development of a physical package

- for important applications:

- *incremental 4D-Var* (ECMWF, Météo-France),
- *simplified gradients in 4D-Var* (Zupanski 1993),
- *the initial perturbations computed for EPS* (ECMWF),

linearized versions of forecast models are run at lower resolution



the linear model can be “not tangent” to the full model

(different resolution and geometry, different physics)



simplified approaches as a way to include progressively physical processes in TL and AD models

- simplifications done with the aim to have a physical package:

- **simple** – for the linearization of the model equations
- **regular** – to avoid strong non-linearities and thresholds
- **realistic enough**
- **computationally affordable**

Full nonlinear vs. simplified physical parametrizations

In NWP – a tendency to develop more and more sophisticated physical parametrizations → they may contain more discontinuities



For the “perturbation” model – more important to describe basic physical tendencies while avoiding the problem of discontinuities

Level of simplifications and/or required complexity depends on:

- which level of improvement is expected (for different variables, vertical and horizontal resolution, ...)
- which type of observations should be assimilated
- necessity to remove threshold processes

Different ways of simplifications:

- development of simplified physics (for instance, gaining from experience with simpler parametrization schemes used in history)
- applying only part of linearization

Problems with including physics in adjoint models

- **Development** – requires substantial resources
- **Validation** – must be very thorough
(for non-linear, tangent-linear and adjoint versions)
- **Computational cost** – may be very high
- **Non-linear and threshold nature** of physical processes
(affecting the range of validity of the tangent-linear approximation)

Validation of the physical parametrizations

Non-linear model:

- Forecast runs with particular modified/simplified physical parametrization schemes

Tangent-linear (TL) and adjoint (AD) model:

- **classical validation** (TL - Taylor formula, AD - test of adjoint identity)
- **examination of the accuracy of the linearization**

Comparison:

finite differences (FD) \leftrightarrow tangent-linear (TL) integration

$$M(\mathbf{x}_{an}) - M(\mathbf{x}_{fg}) \leftrightarrow M'(\mathbf{x}_{an} - \mathbf{x}_{fg})$$

$(an = analysis, \quad fg = first\ guess)$

Singular vectors:

- Computation of singular vectors to find out whether the new schemes do not produce spurious unstable modes.

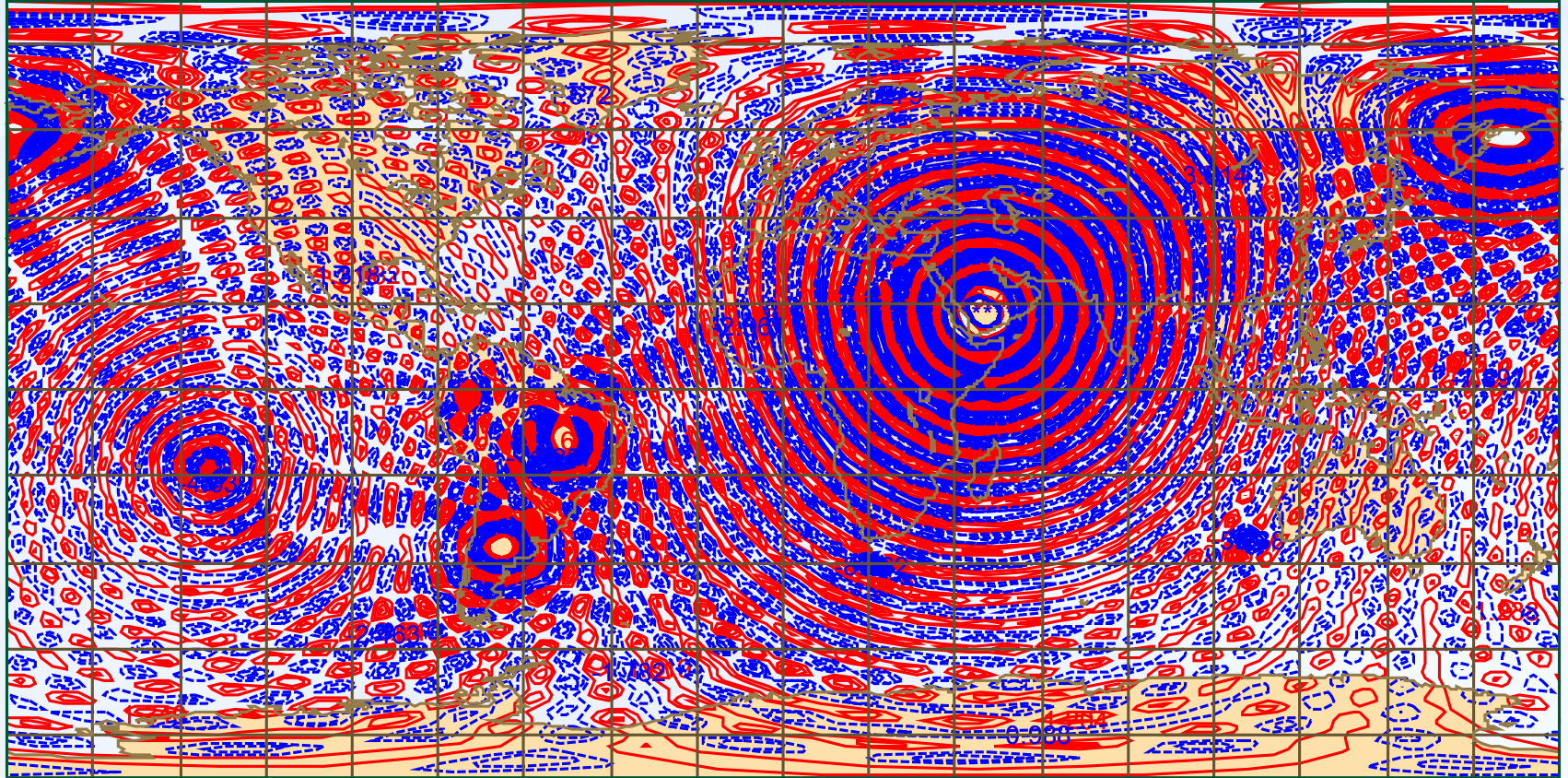
Importance of the regularization of TL model

- physical processes are characterized by:
 - * threshold processes:
 - discontinuities of some functions describing the physical processes
(*some on/off processes*)
 - discontinuities of the derivative of a continuous function
 - * strong nonlinearities

WHY REGULARIZATION IS IMPORTANT



lv31 T* 1999-03-15 12h fc t+6 - TL with vdif (no regularization applied) [cont.int: 0.5e+07]

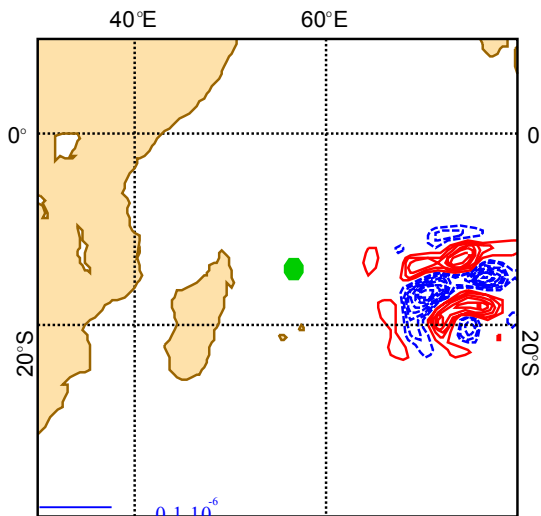


Without any treatment of most serious threshold processes, the TL approximation can turn to be useless.

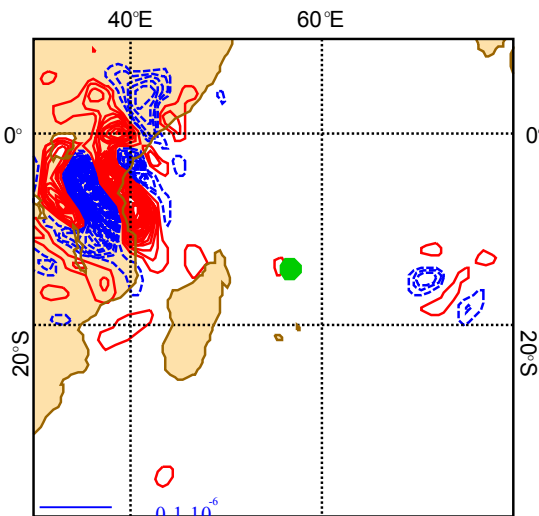
SPURIOUS UNSTABLE MODES PRODUCED BY THE LINEARIZED PHYSICS

The first singular vectors located around the cyclone (58°E, 18°S) computed at the resolution T95 (Barkmeijer, 2002)

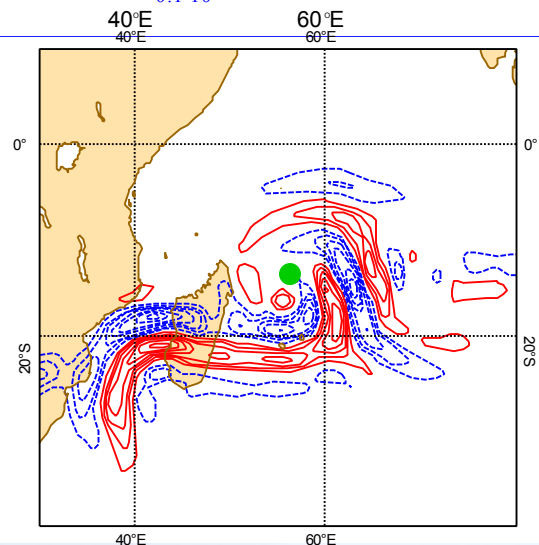
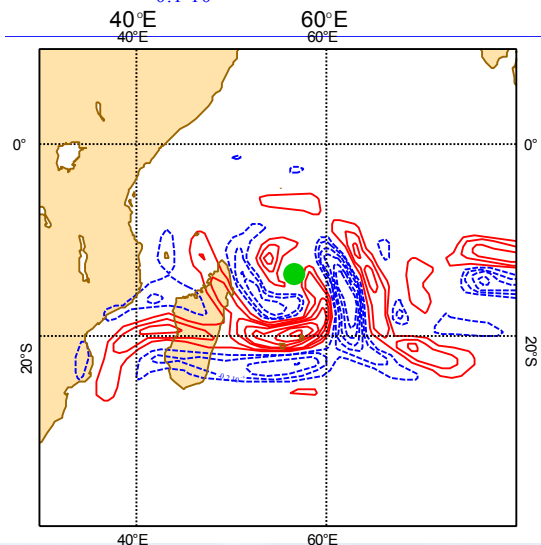
SL diabatic T95



lv19 VO 2001-01-04 12h

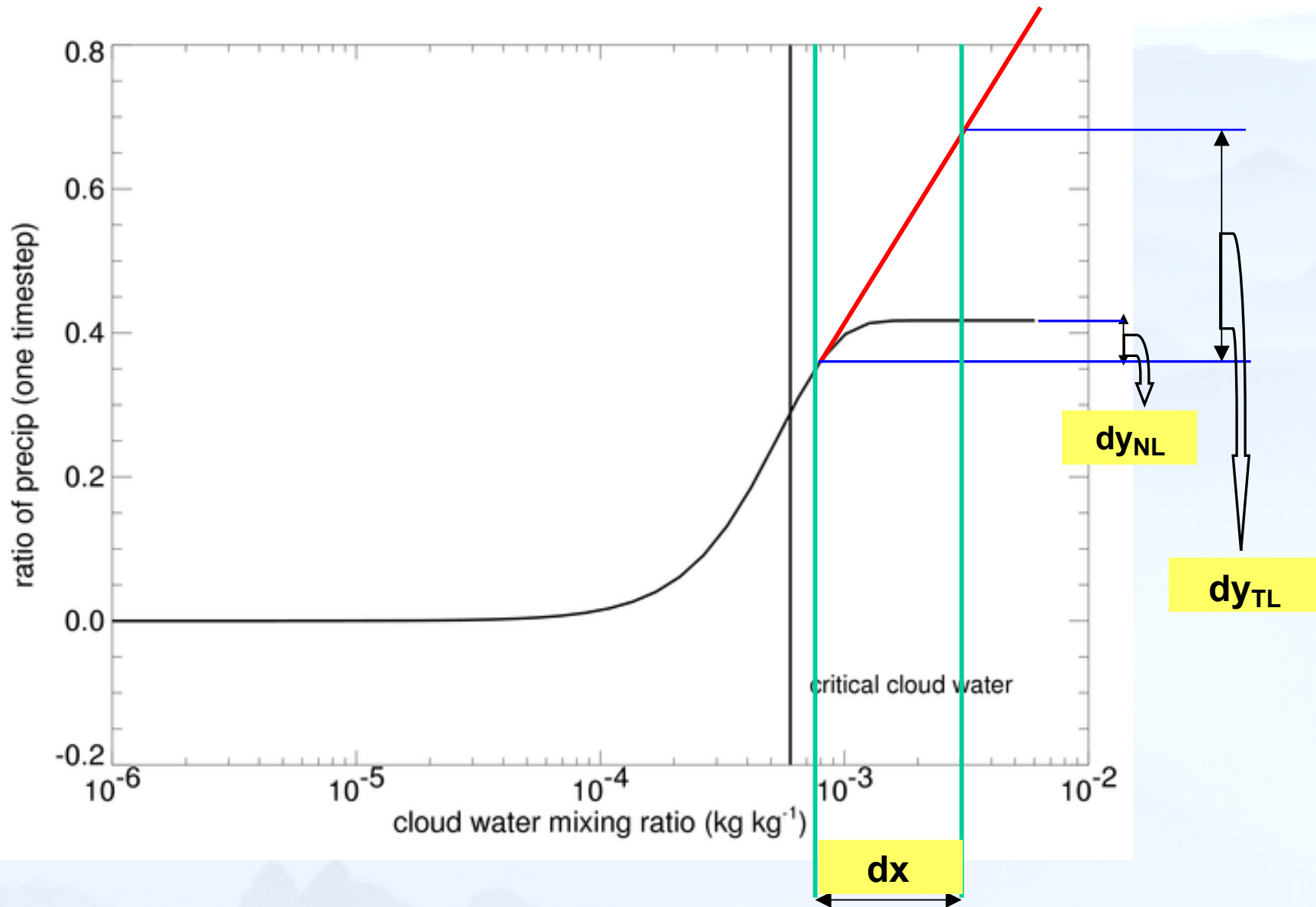


all linearized
physical
parametrization
schemes

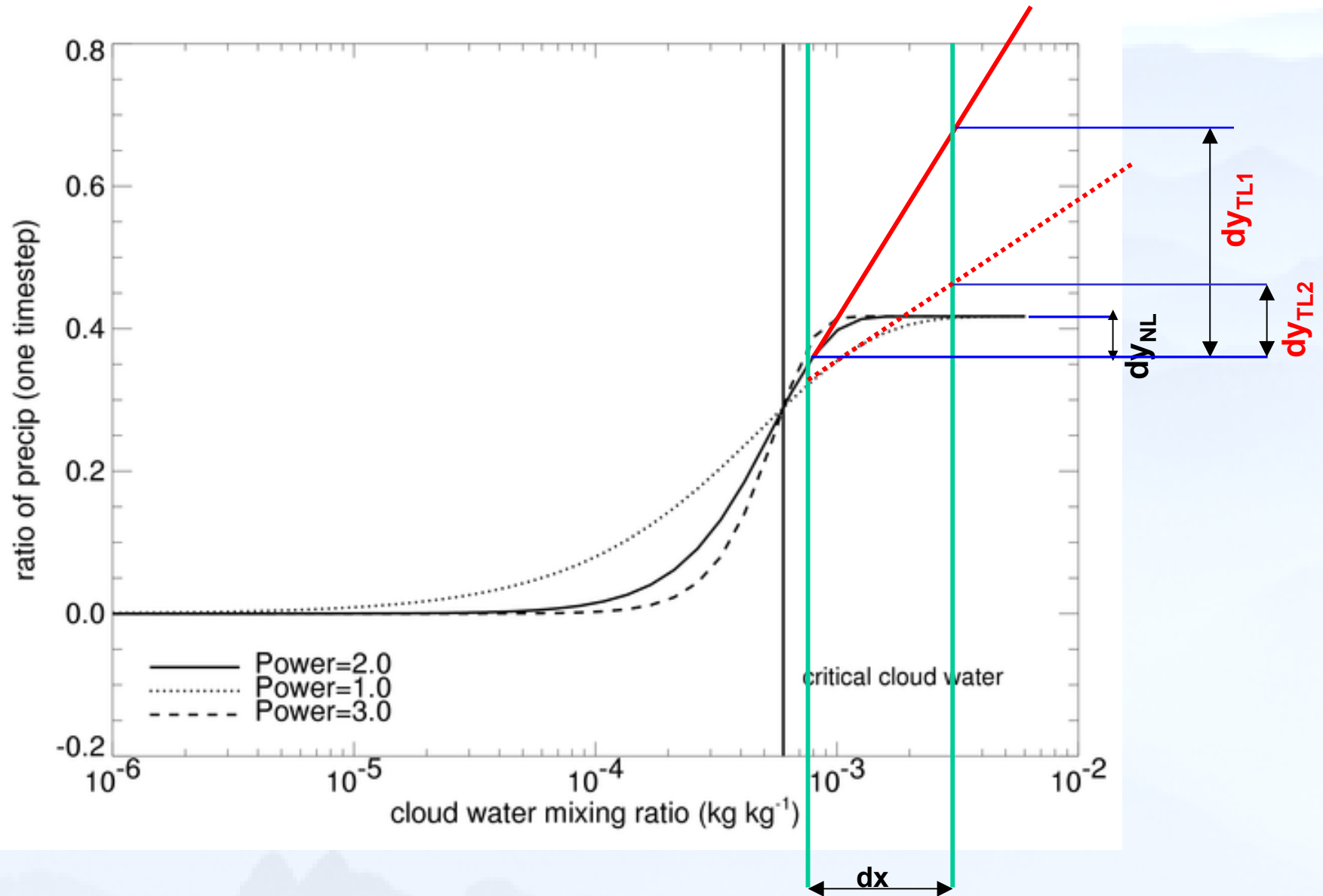


without
the linearized
large-scale
precipitation
scheme

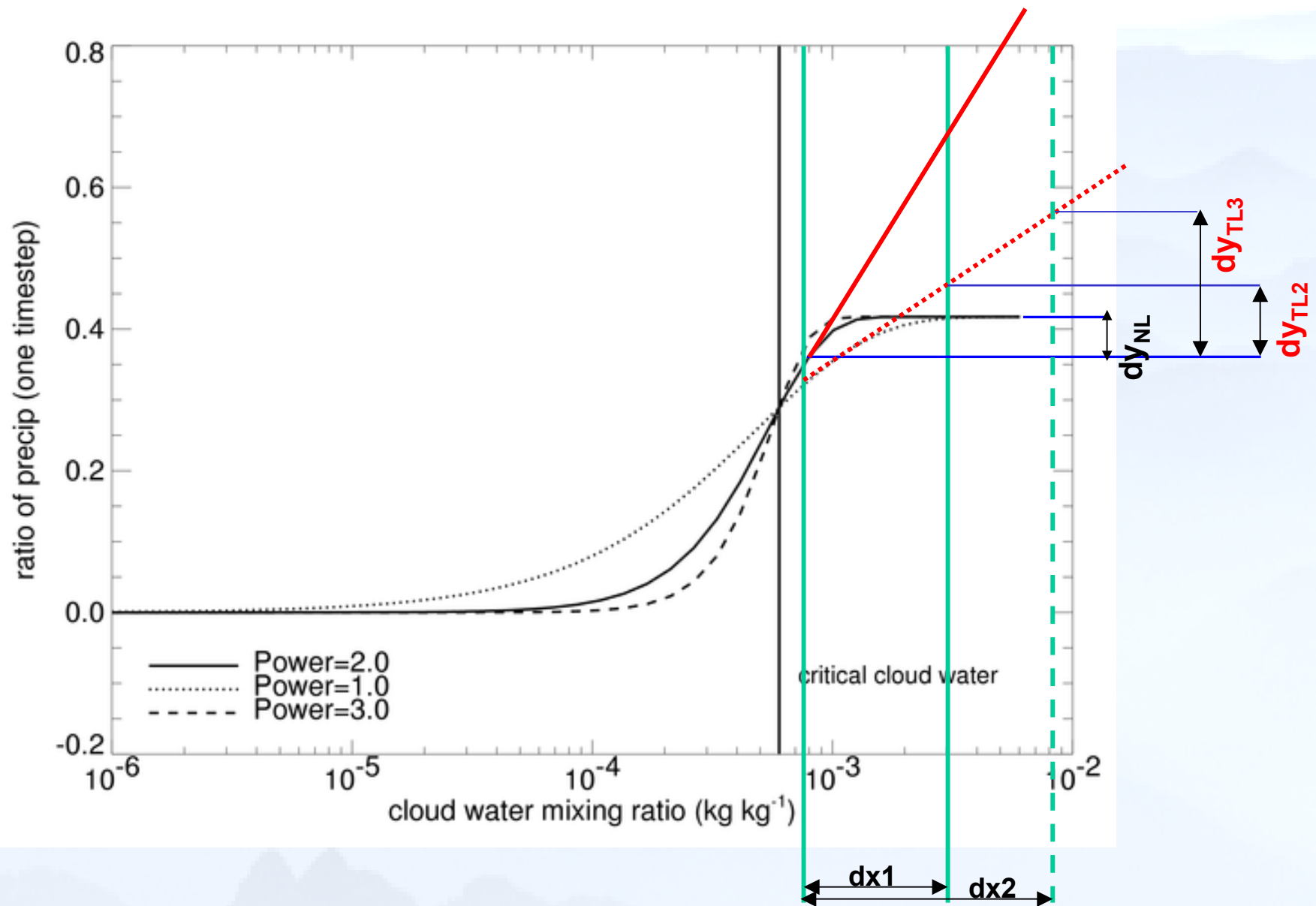
Potential source of problem



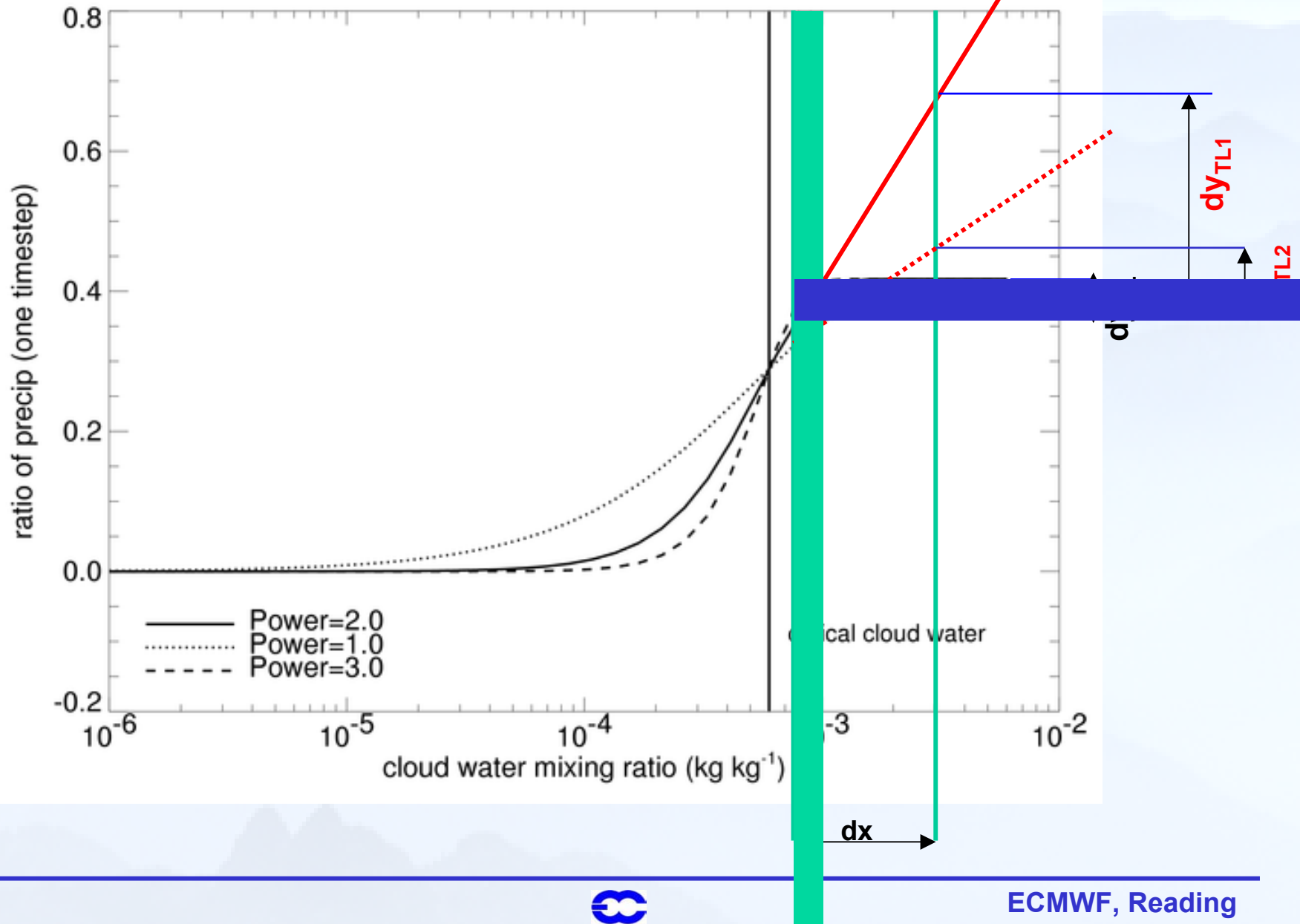
Possible solution, but ...



... may just postpone the problem and influence the performance of NL scheme



However, the better the model \rightarrow the smaller the increments



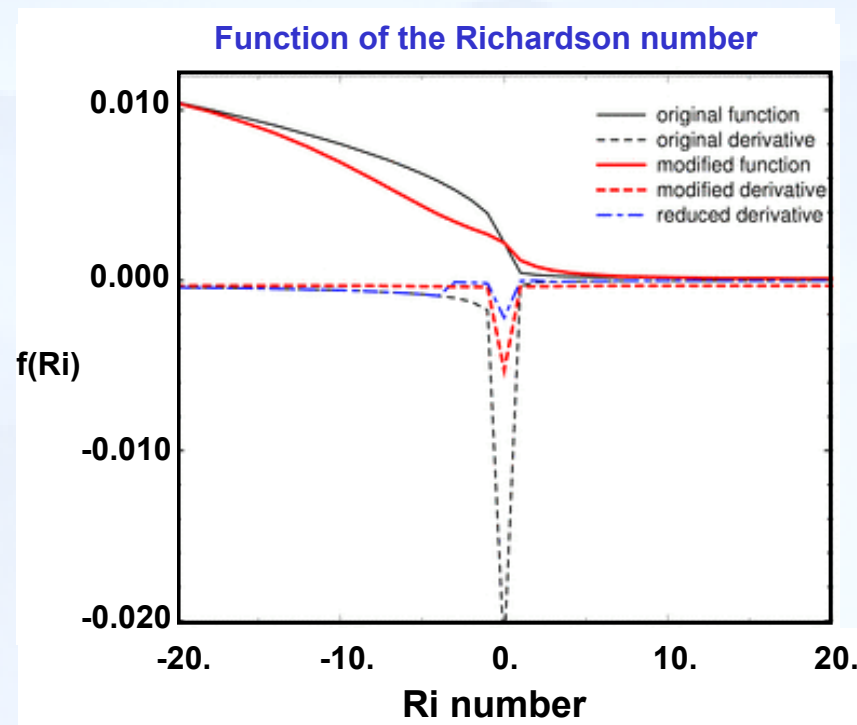
Examples of regularizations and simplifications (1)

Regularization of vertical diffusion scheme:

- **perturbation of the exchange coefficients** (which are function of the Richardson number Ri) **is neglected, $K' = 0$** (Mahfouf, 1999)

- **reduced perturbation of the exchange coefficients** (Janisková et al., 1999):

- original computation of Ri modified in order to modify/reduce $f'(Ri)$, or
- reducing a derivative, $f'(Ri)$, by factor 10 in the central part (around the point of singularity)

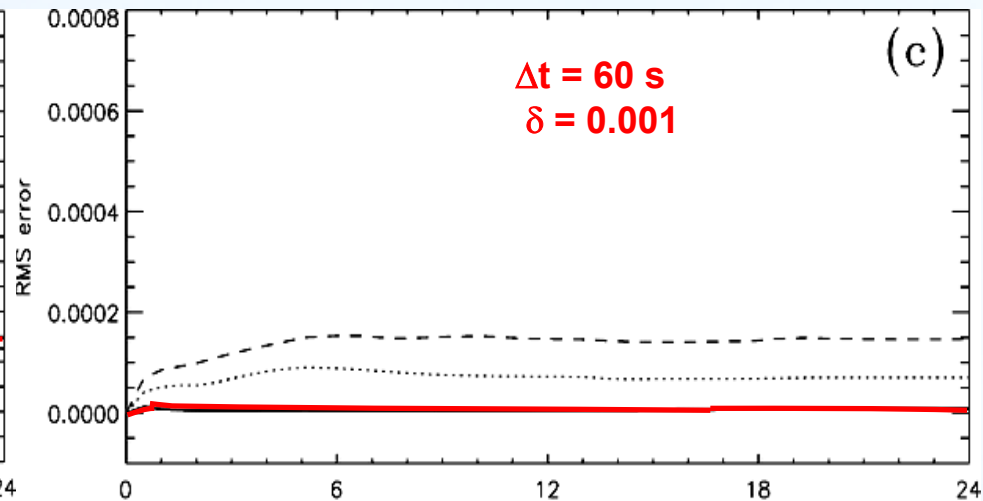
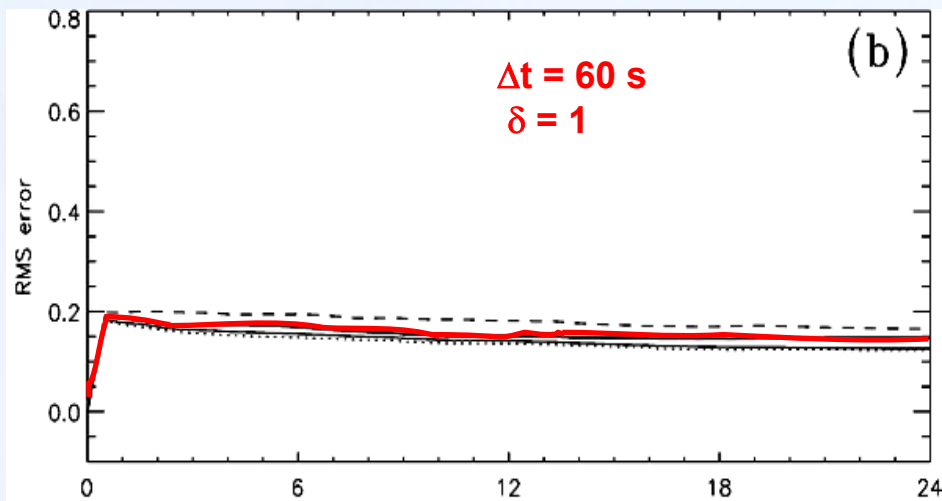
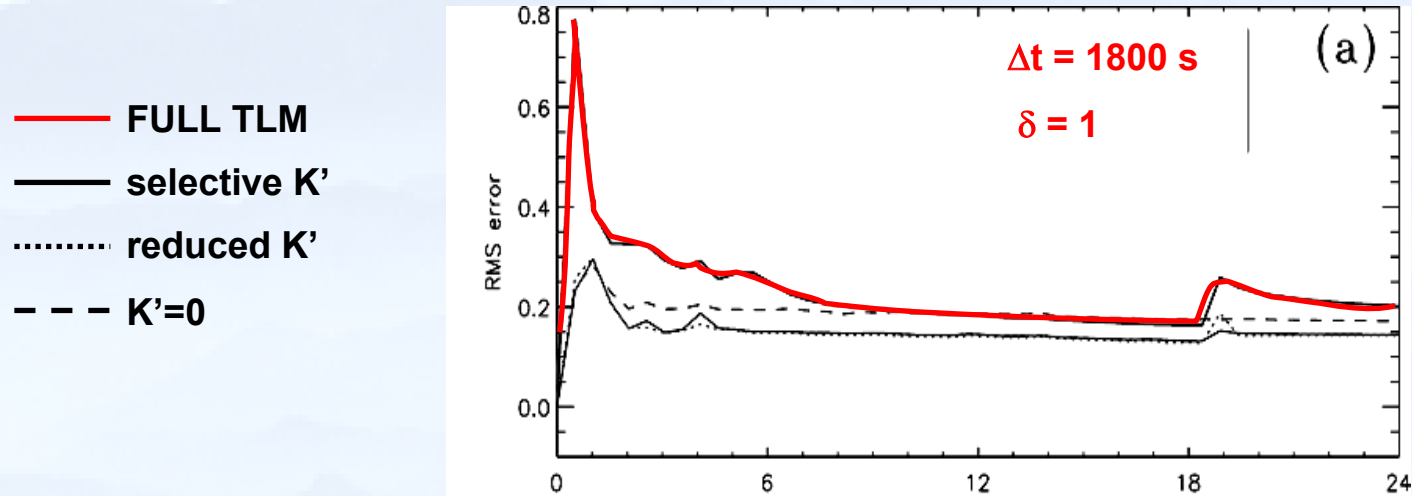


Examples of regularizations and simplifications (2)

- **reduction of the time step to 10 seconds** to guarantee stable time integrations of the associated TL model (*Zhu and Kamachi, 2000*)
- **selective regularization of the exchange coefficients K** based on the linearization error and a criterion for the numerical stability (*Laroche et al., 2002*)

Comparison: FULL TL – $K'=0$ – reduced K' – selective K' (Laroche et al.2002)

RMS linearization errors for the potential temperature perturbations at the 1st level above the surface



Importance of the regularization of TL model

- regularizations help to remove the most important threshold processes in physical parametrizations which can effect the range of validity of the tangent linear approximation
- after solving the threshold problems

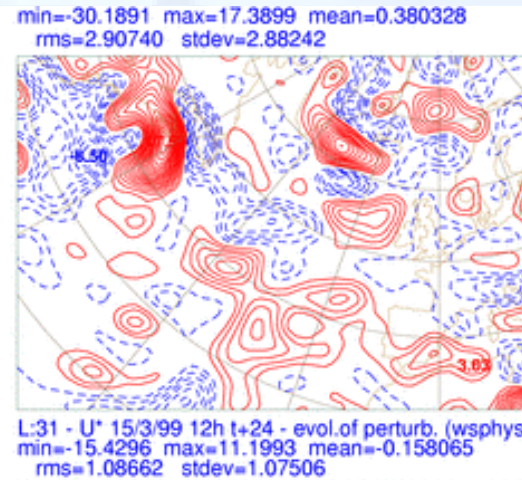
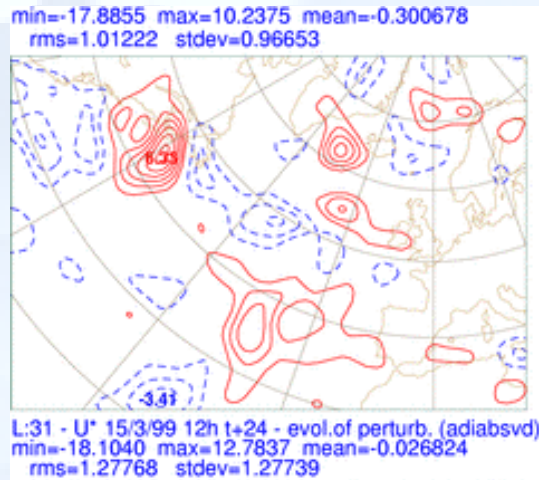


clear advantage of the diabatic TL evolution of errors compared to the adiabatic evolution

Zonal wind increments at model level ~ 1000 hPa

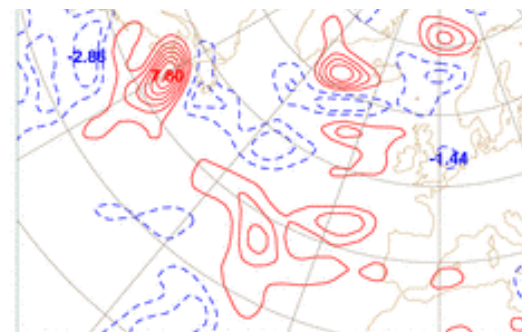
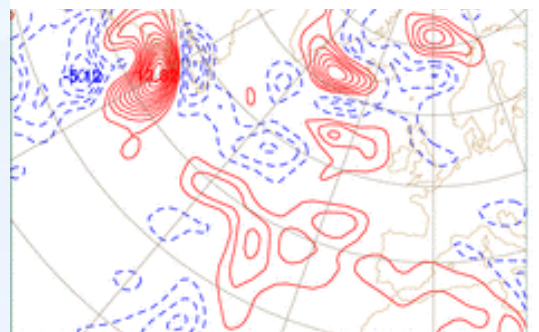
[24-hour integration]

FD



TL_{ADIAB}

TL_{ADIABSVD}



TL_{WSPHYS}

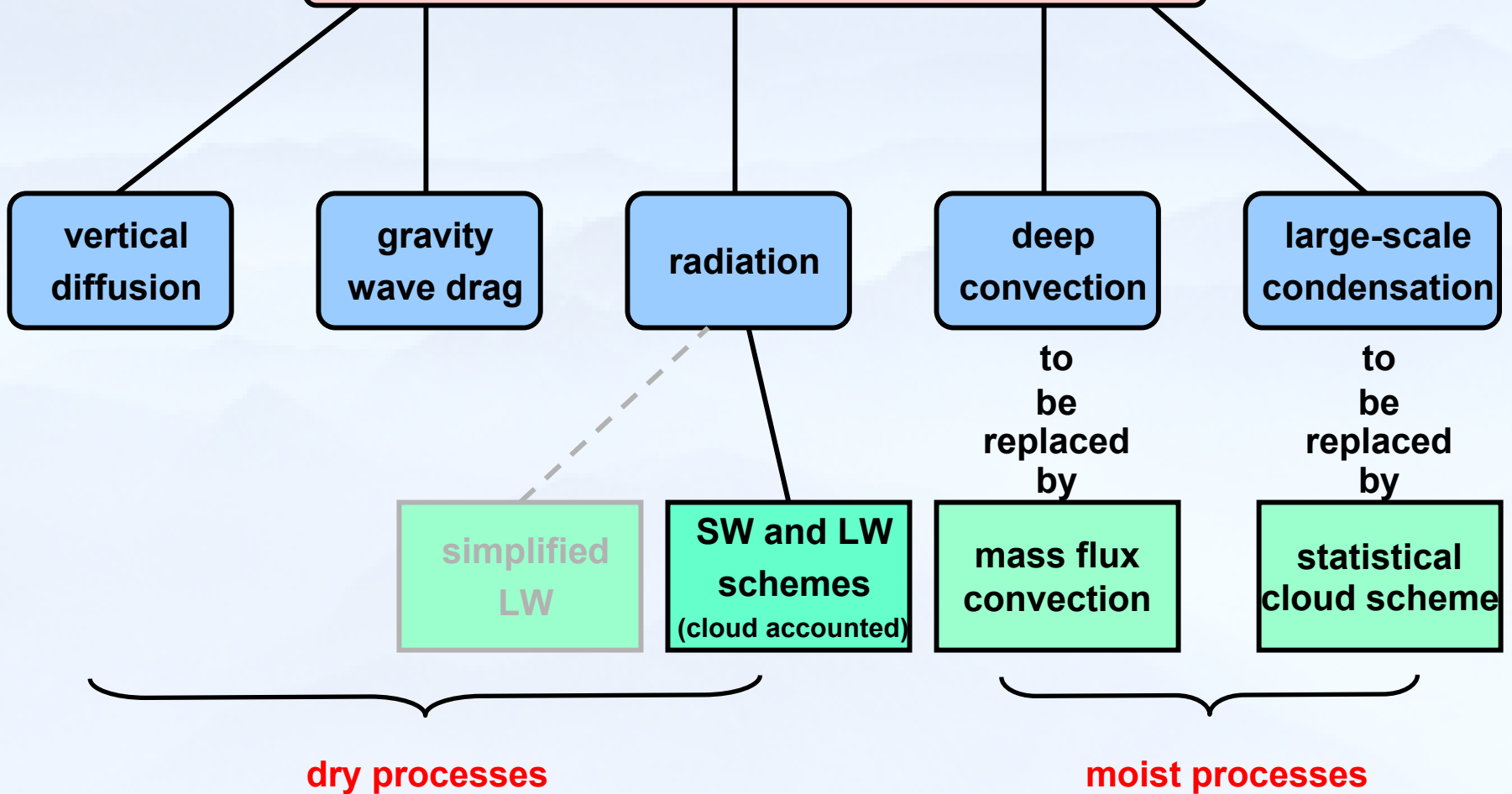
TL_{ADIAB} – adiabatic TL model

TL_{ADIABSVD} – TL model with very simple vertical diffusion (*Buizza 1994*)

TL_{WSPHYS} – TL model with the whole set of simplified physics (*Mahfouf 1999*)

Simplified physical parametrizations

ECMWF LINEARIZED PHYSICS



Tangent-linear diagnostics

Comparison:

finite differences (FD) \leftrightarrow tangent-linear (TL) integration

$$M(\mathbf{x}_{an}) - M(\mathbf{x}_{fg}) \leftrightarrow M'(\mathbf{x}_{an} - \mathbf{x}_{fg})$$

$(an = analysis, \quad fg = first\ guess)$

Diagnostics:

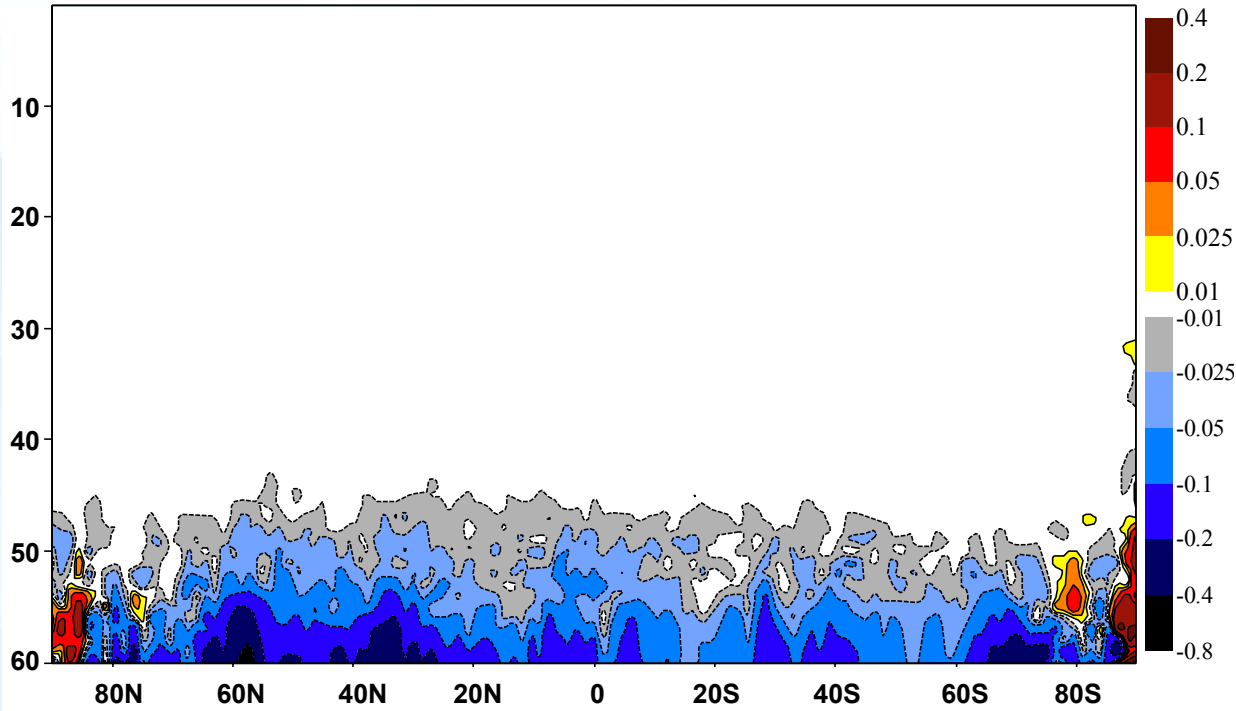
- mean absolute errors:

$$\varepsilon = \left| \left[M(\mathbf{x}_{an}) - M(\mathbf{x}_{fg}) \right] - M'(\mathbf{x}_{an} - \mathbf{x}_{fg}) \right|$$

- relative error

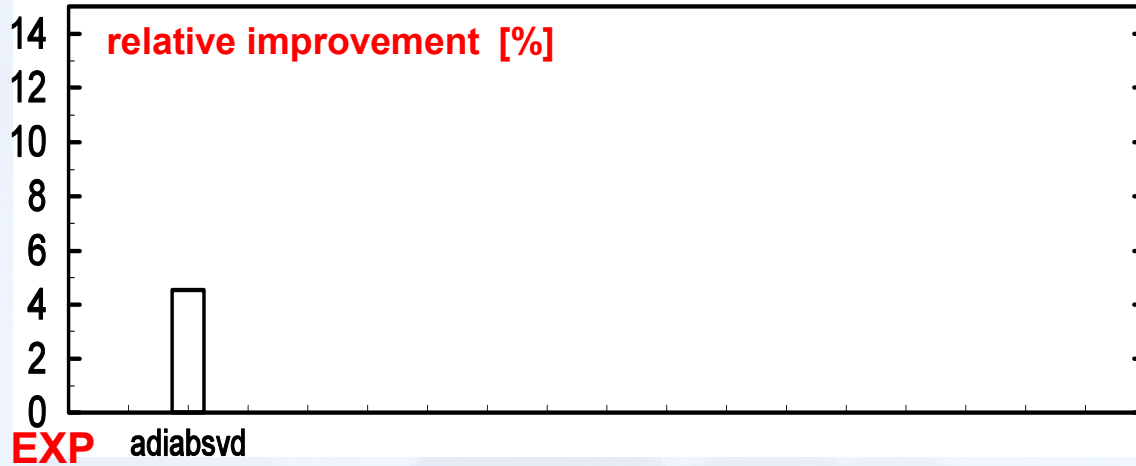
$$\frac{\varepsilon_{\text{EXP}} - \varepsilon_{\text{REF}}}{\varepsilon_{\text{REF}}} \cdot 100\%$$

Temperature

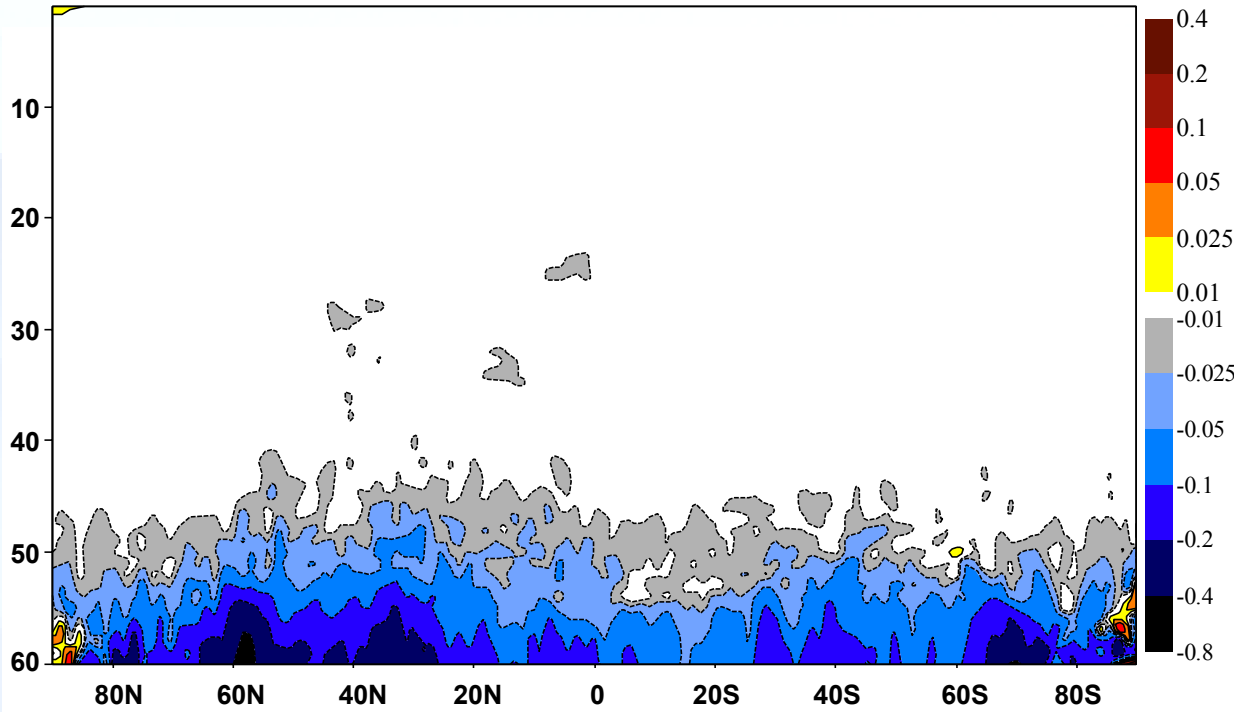


$$\epsilon_{\text{EXP}} - \epsilon_{\text{REF}}$$

REF = ADIAB

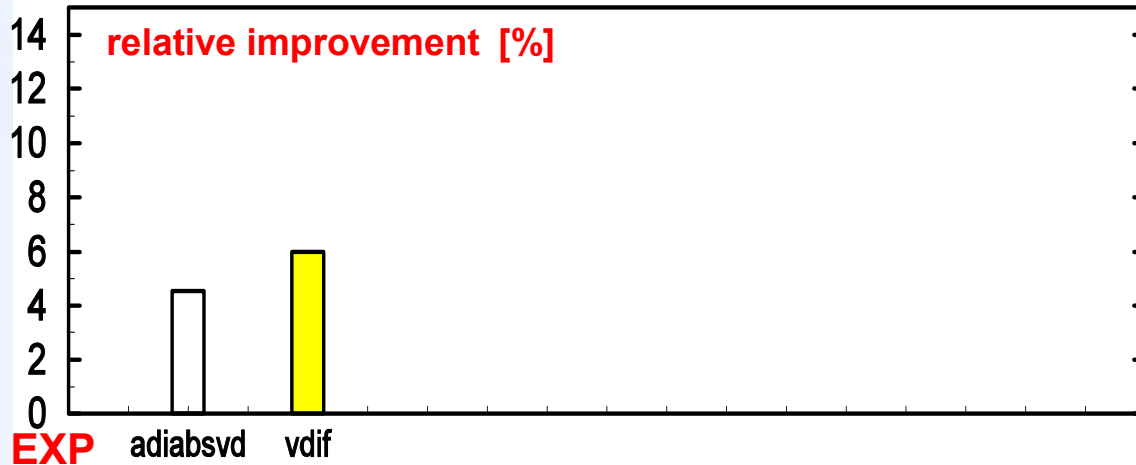


Temperature

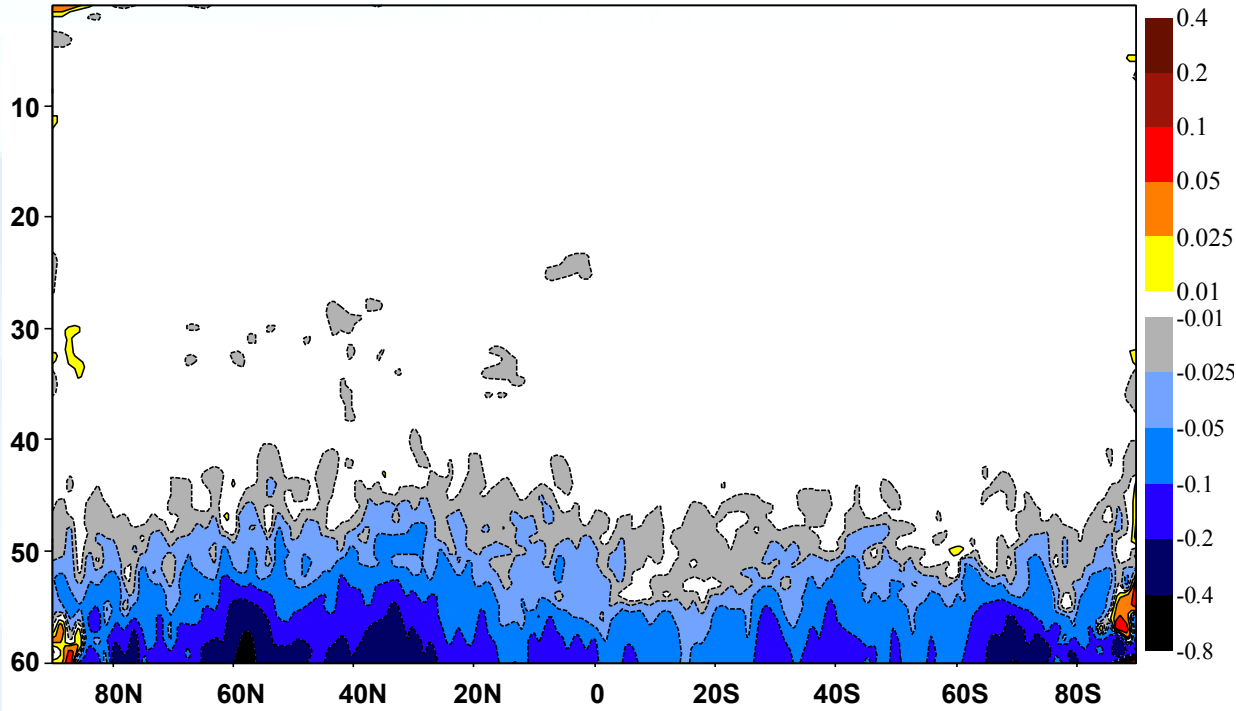


$$\epsilon_{\text{EXP}} - \epsilon_{\text{REF}}$$

REF = ADIAB

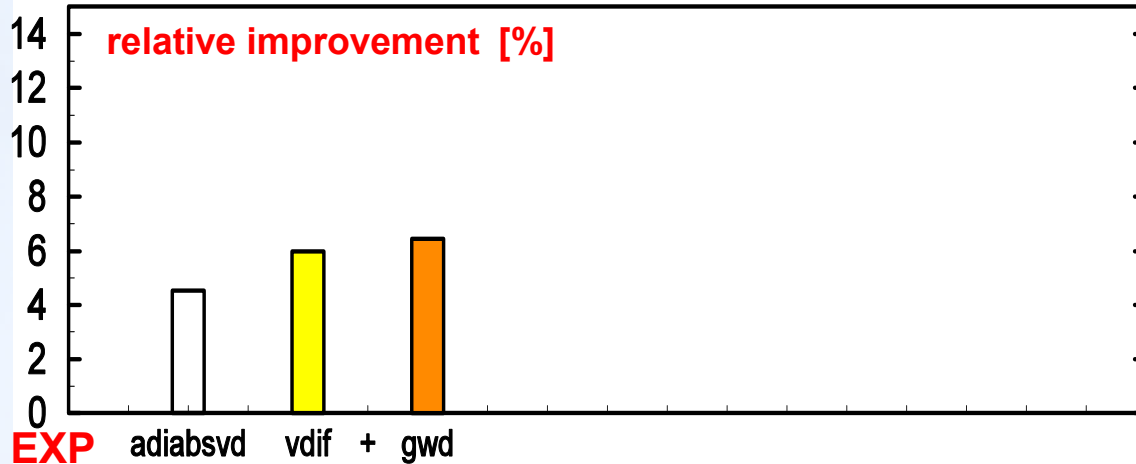


Temperature



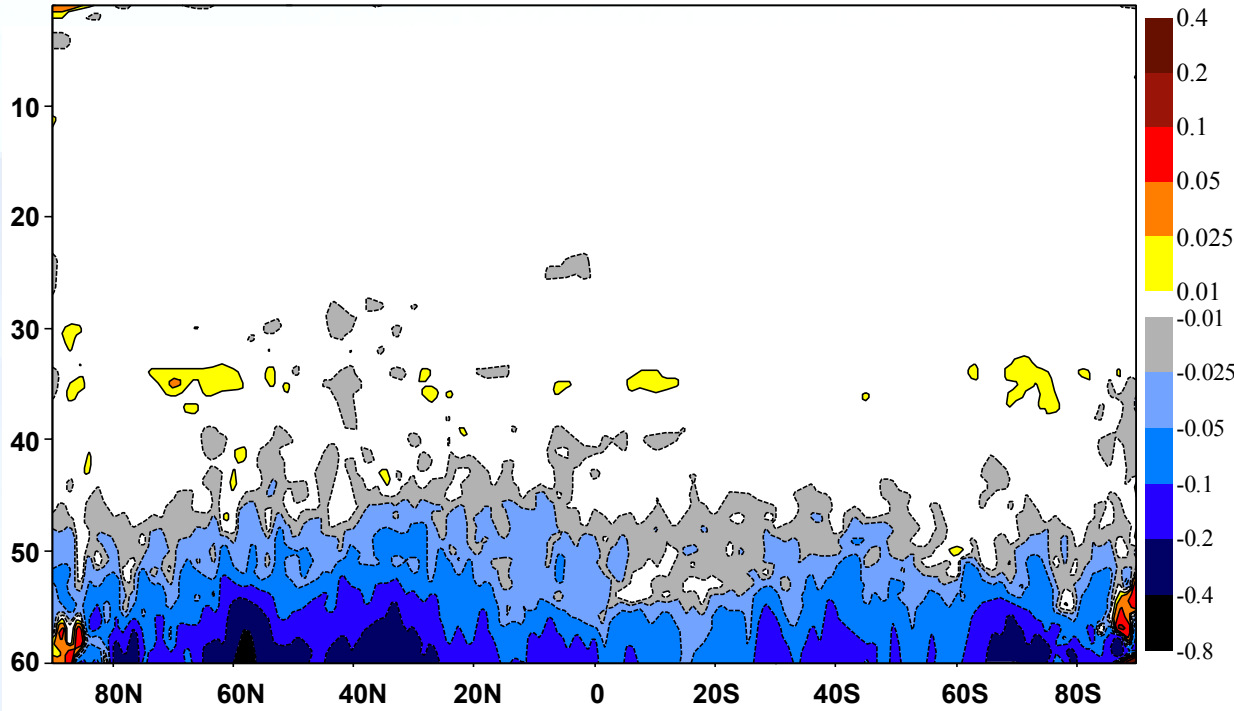
$$\epsilon_{\text{EXP}} - \epsilon_{\text{REF}}$$

REF = ADIAB



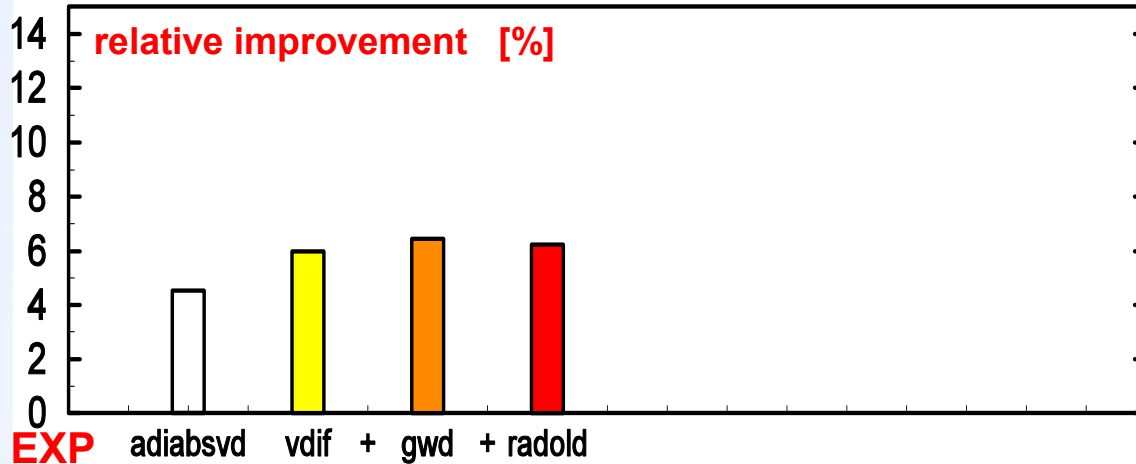
adiabsvd || vdif + gwd

Temperature

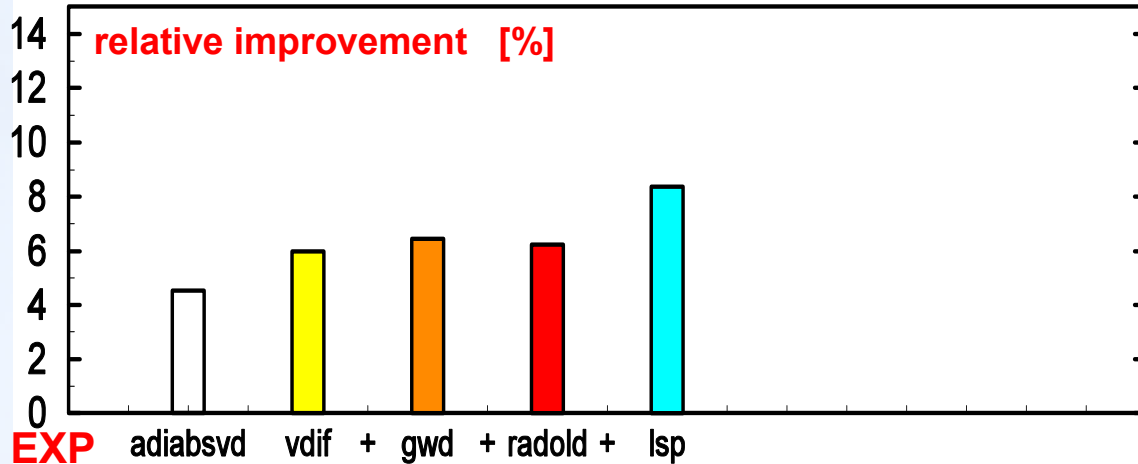
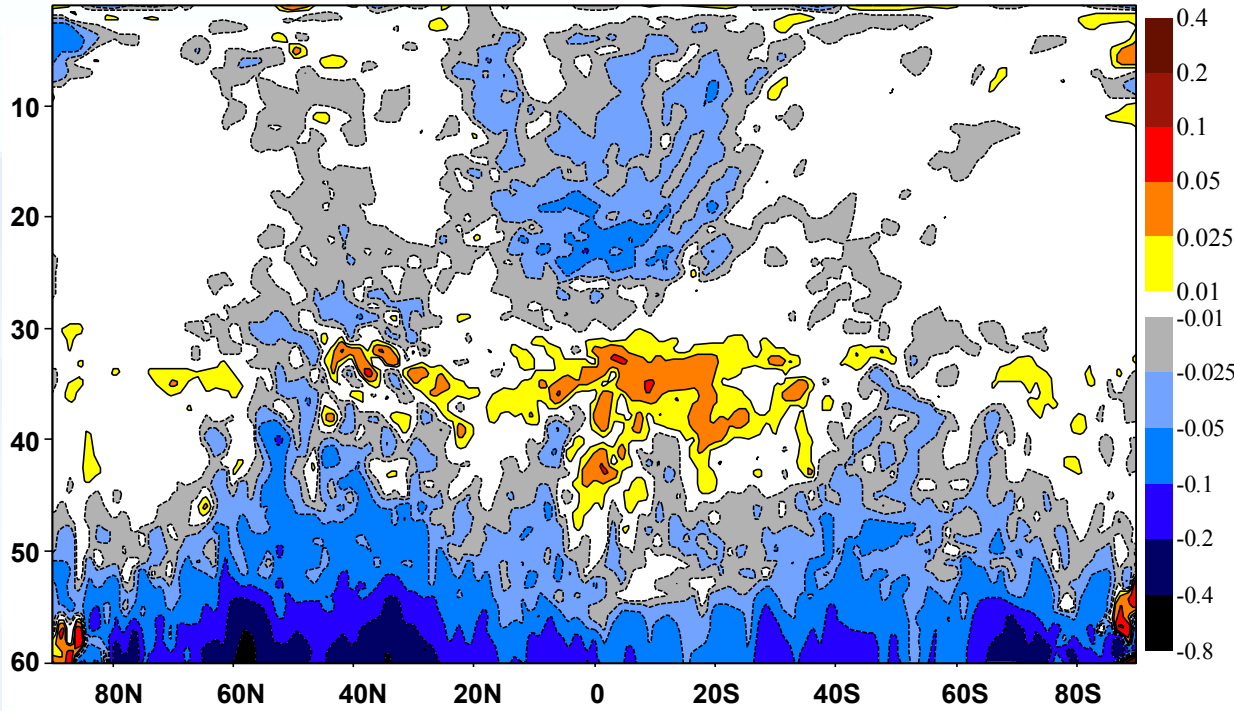


$$\epsilon_{EXP} - \epsilon_{REF}$$

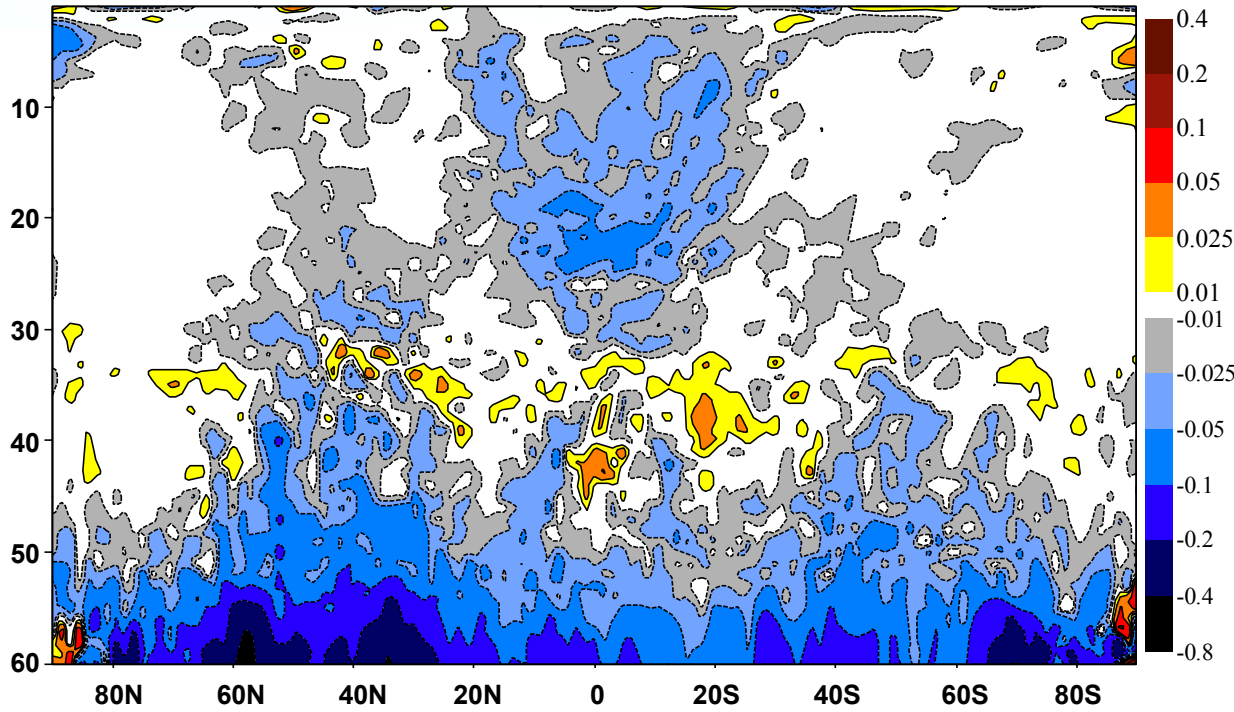
REF = ADIAB



Temperature

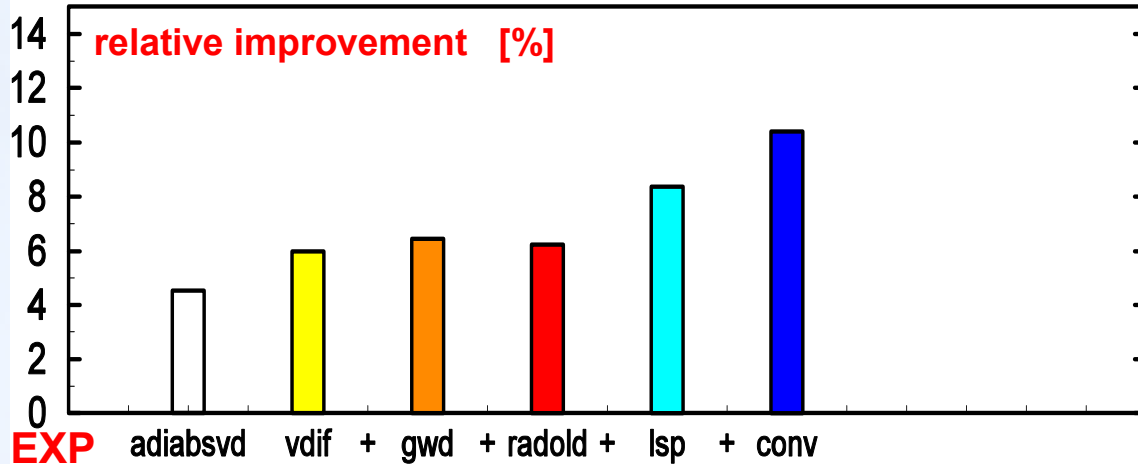


Temperature

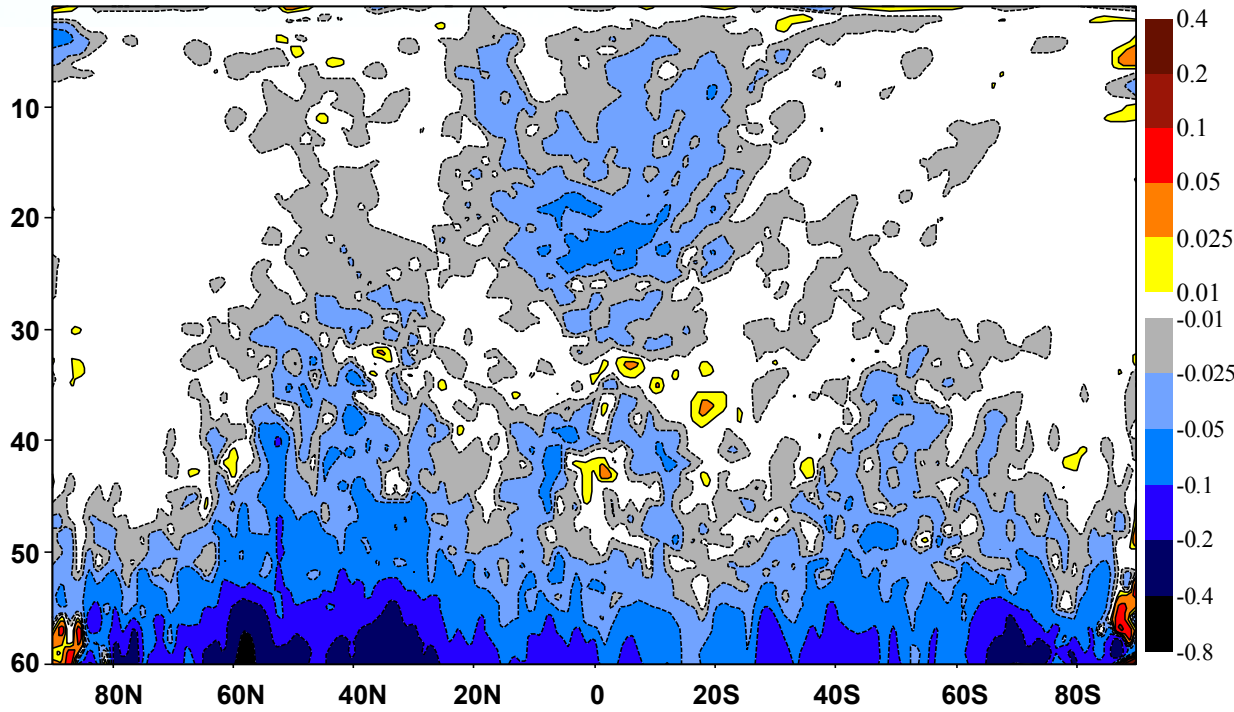


$$\epsilon_{EXP} - \epsilon_{REF}$$

REF = ADIAB

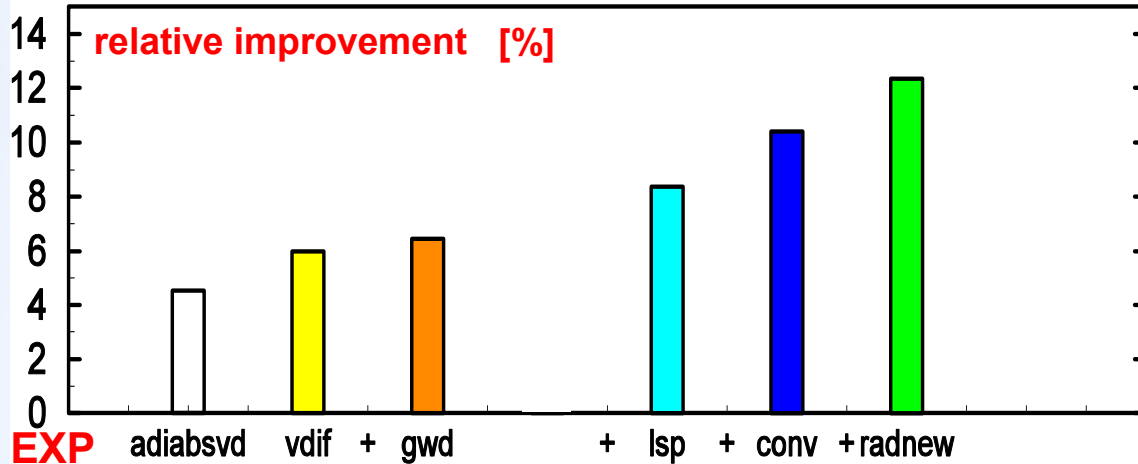


Temperature



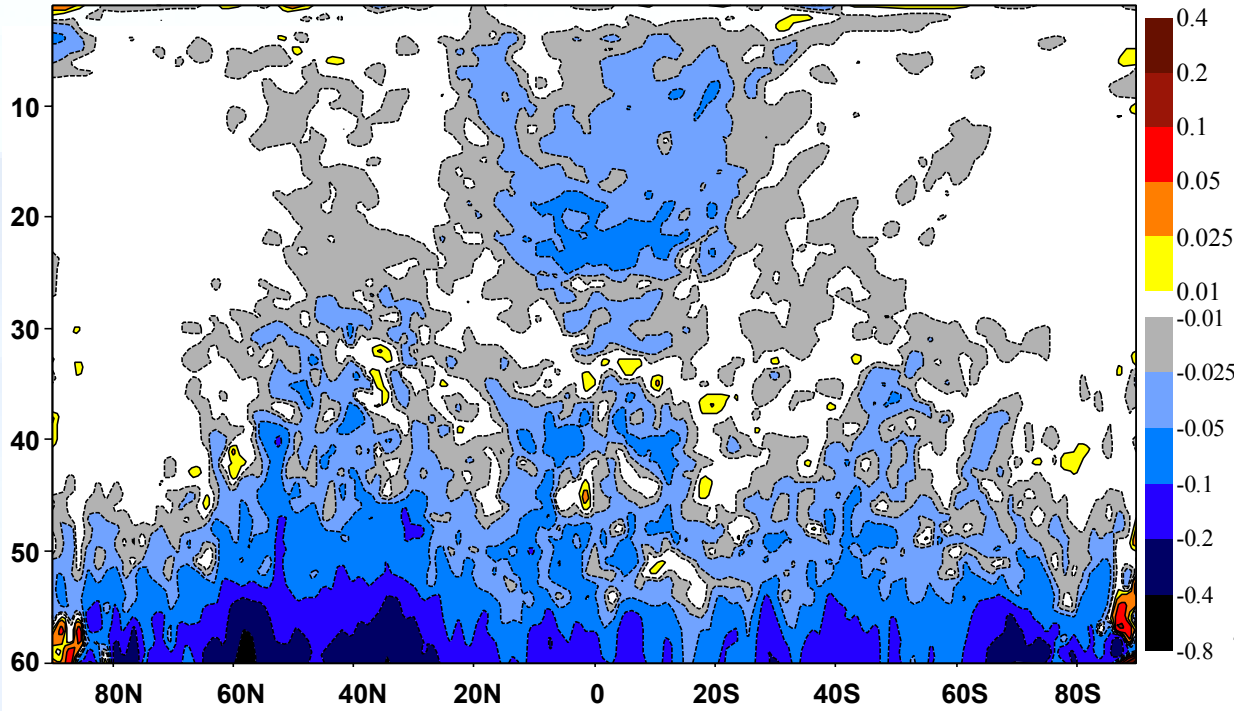
$$\epsilon_{EXP} - \epsilon_{REF}$$

REF = ADIAB



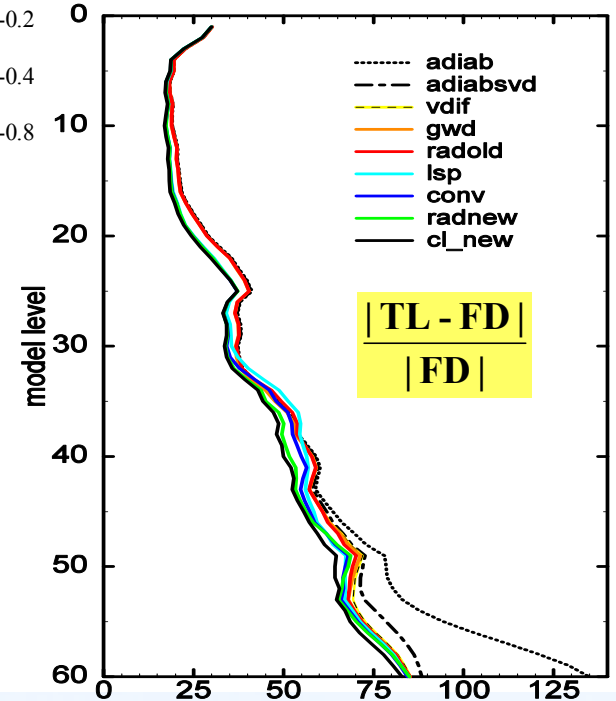
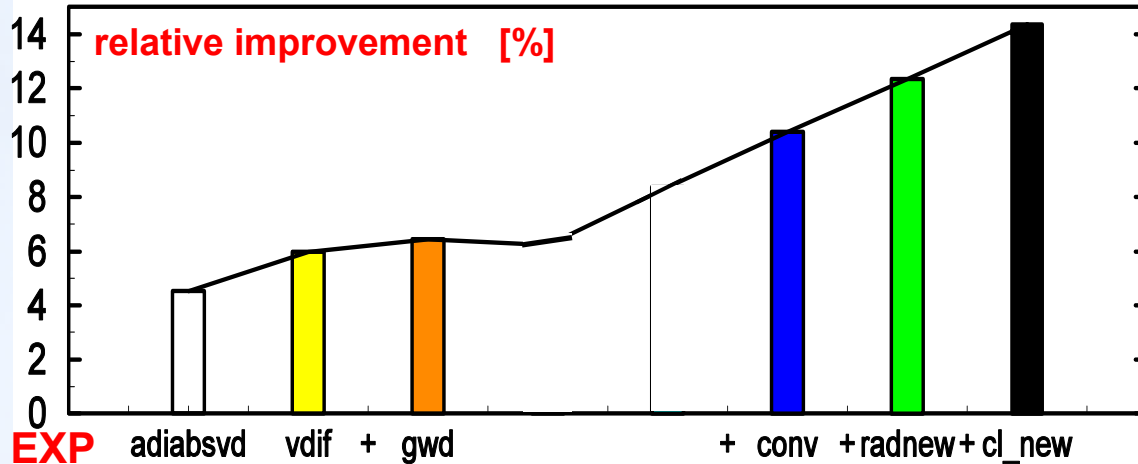
adiabsvd || vdif + gwd + lsp + conv + radnew

Temperature



$$\epsilon_{\text{EXP}} - \epsilon_{\text{REF}}$$

REF = ADIAB



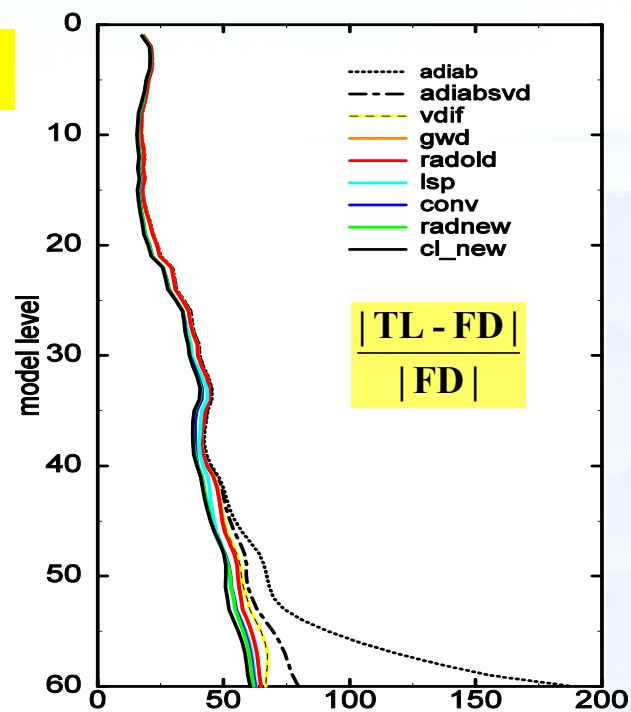
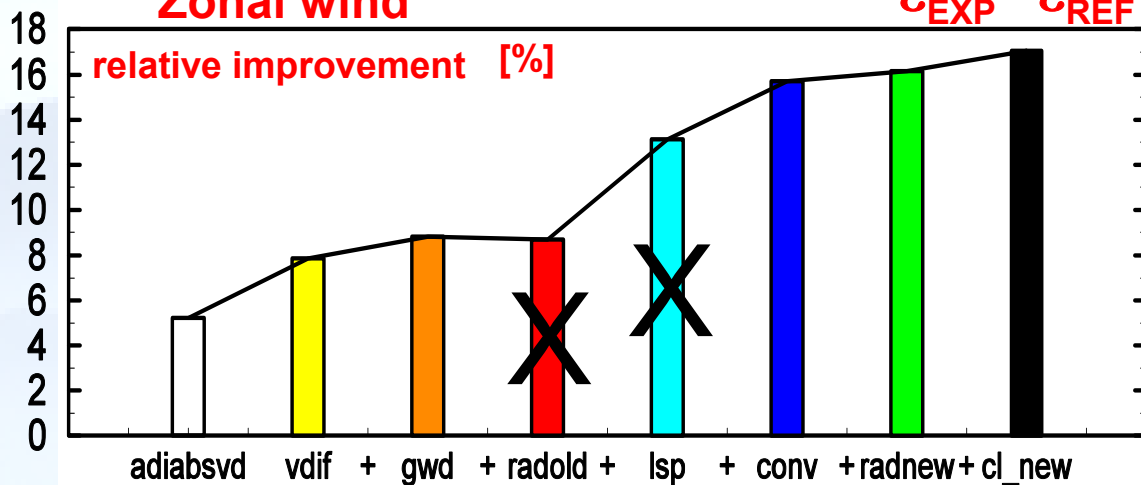
adiabsvd || vdif + gwd

+ conv + radnew + cl_new

Impact of physical processes

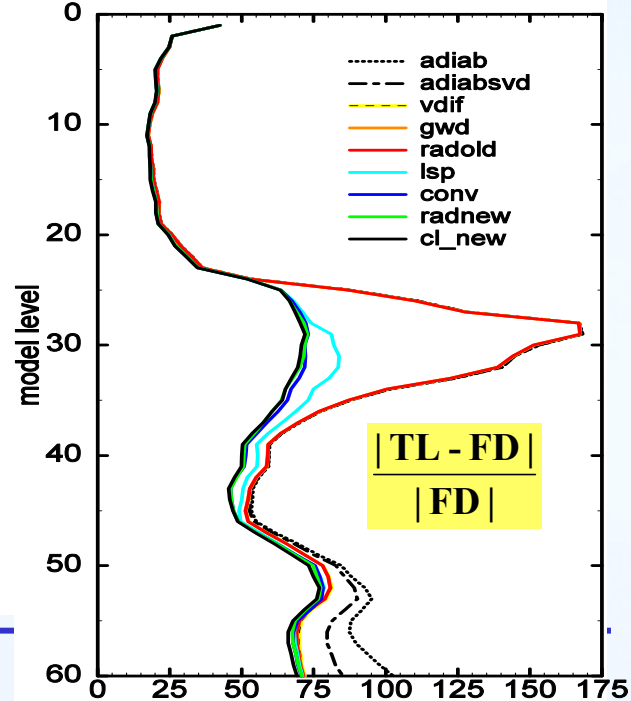
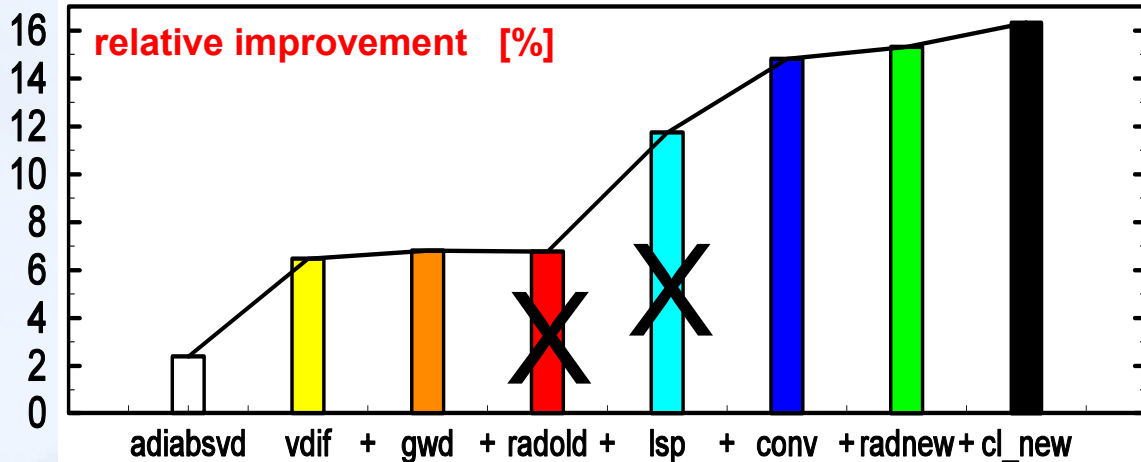
Zonal wind

$\epsilon_{EXP} - \epsilon_{REF}$



Specific humidity

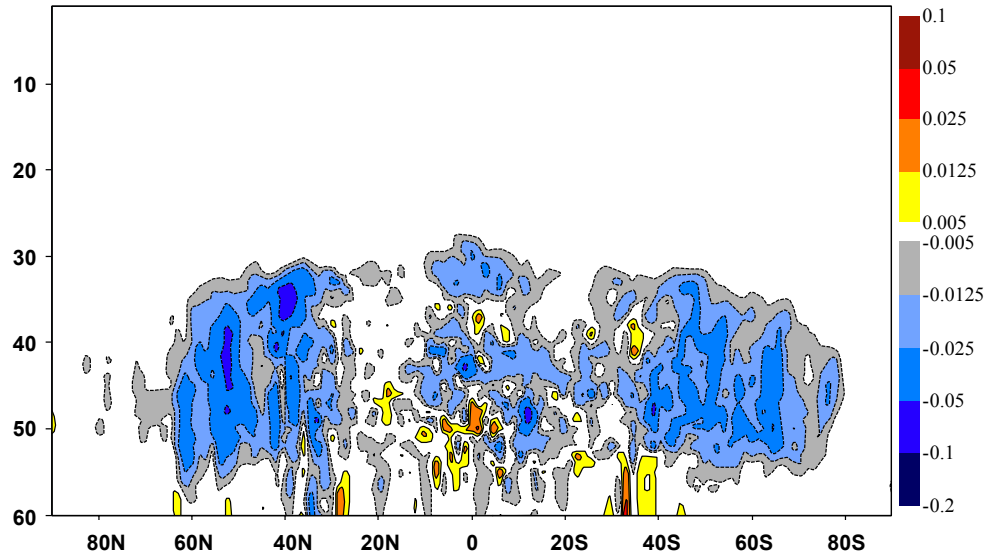
$\epsilon_{EXP} - \epsilon_{REF}$



Impact of moist processes (lsp + conv)

$$\epsilon_{\text{EXP}} - \epsilon_{\text{REF}}$$

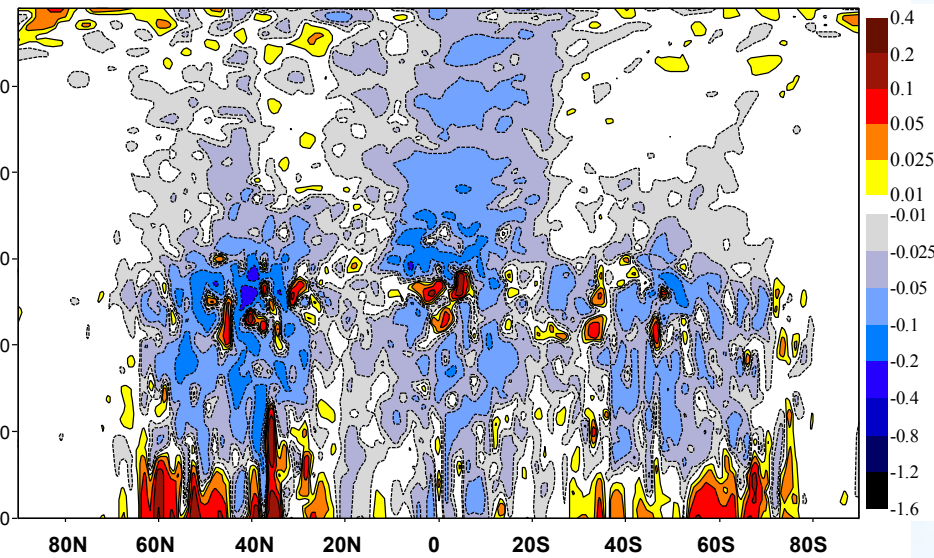
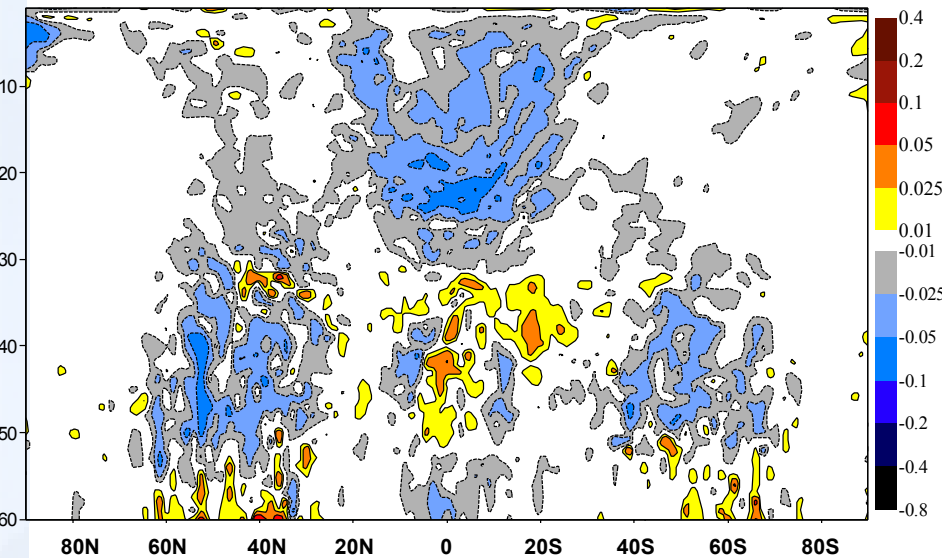
REF = ADIAB



Specific humidity: 7.7 %

Temperature: 3.8 %

Zonal wind: 6.1 %



Physics in 4D-Var

- In incremental 4D-Var, the objective function is minimized in terms of increments:

$$\delta \mathbf{x}_i = \mathbf{M}'(t_i, t_0) \delta \mathbf{x}_0 \quad \leftarrow \text{tangent linear model}$$

with the model state defined at any time t_i as: $\mathbf{x}_i = \mathbf{x}_i^b + \delta \mathbf{x}_i$, $\mathbf{x}_i^b = \mathbf{M}(t_i, t_0) \mathbf{x}_0^b$

- 4D-Var can be then approximated to the first order as minimizing:

$$\mathcal{J}(\delta \mathbf{x}_0) = \frac{1}{2} \delta \mathbf{x}_0^T \mathbf{B}^{-1} \delta \mathbf{x}_0 + \frac{1}{2} \sum_{i=0}^n \left(H'_i(\delta \mathbf{x}_i) - \mathbf{d}_i \right)^T \mathbf{R}_i^{-1} \left(H'_i(\delta \mathbf{x}_i) - \mathbf{d}_i \right)$$

where $\mathbf{d}_i = y_i^o - H_i(\mathbf{x}_i^b)$ is the innovation vector

Gradient of the objective function to be minimized:

$$\nabla_{\delta \mathbf{x}_0} \mathcal{J} = \mathbf{B}^{-1} \delta \mathbf{x}_0 + \frac{1}{2} \sum_{i=0}^n \mathbf{M}^T(t_i, t_0) \mathbf{H}_i^T \mathbf{R}_i^{-1} \left(H'_i(\delta \mathbf{x}_i) - \mathbf{d}_i \right)$$

\mathbf{d}_i \leftarrow computed with the non-linear model at high resolution using full physics $\leftarrow M$

$\delta \mathbf{x}_i$ \leftarrow computed with the tangent-linear model at low resolution using simplified physics $\leftarrow M'$

$\nabla_{\delta \mathbf{x}_0} \mathcal{J}$ \leftarrow computed with a low resolution adjoint model using simplified physics $\leftarrow \mathbf{M}^T$

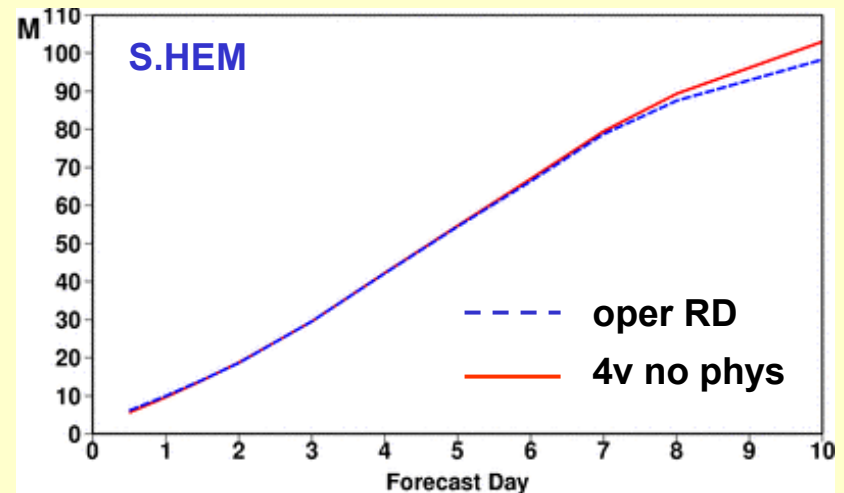
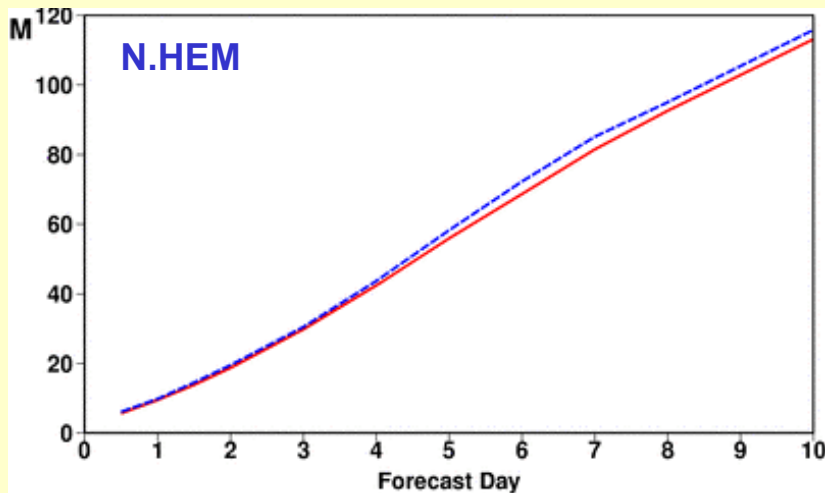
Impact of the linearized physics in 4D-Var (1)

- comparisons of the operational version of 4D-Var against the version without linearized physics included shows:
 - *positive impact on analysis and forecast*

FORECAST VERIFICATION – 500 hPa GEOPOTENTIAL

period: 15/11/2000 – 13/12/2000

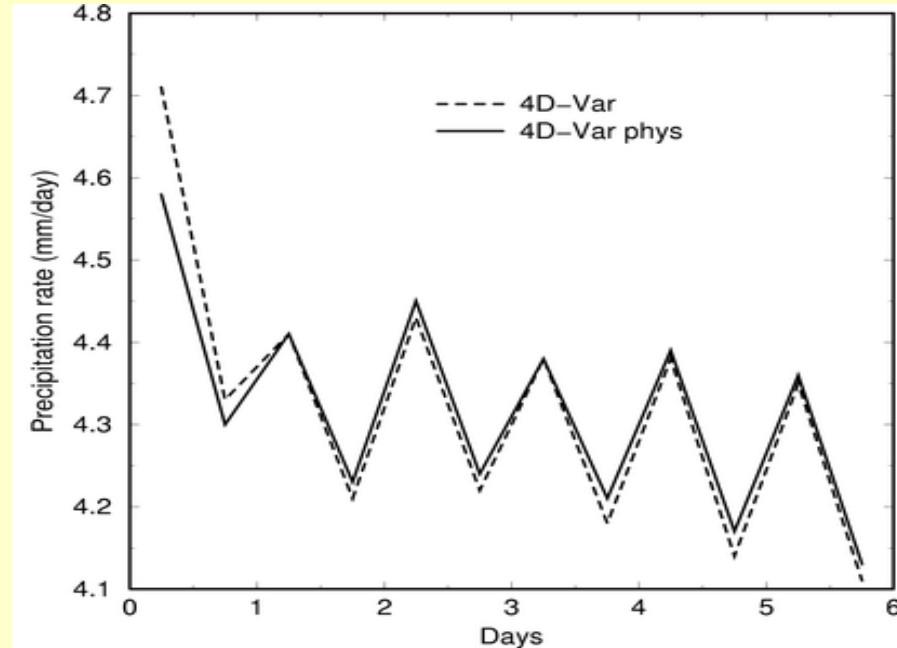
root mean square error – 29 cases



Impact of the linearized physics in 4D-Var (2)

– *reducing spin-up problem when using physical processes*

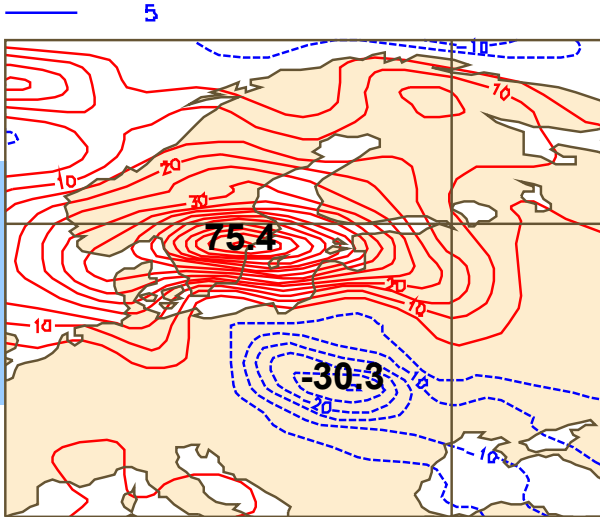
Time evolution of global precipitation in the tropical belt [30S, 30N] averaged over 14 forecasts issued from 4D-Var assimilation



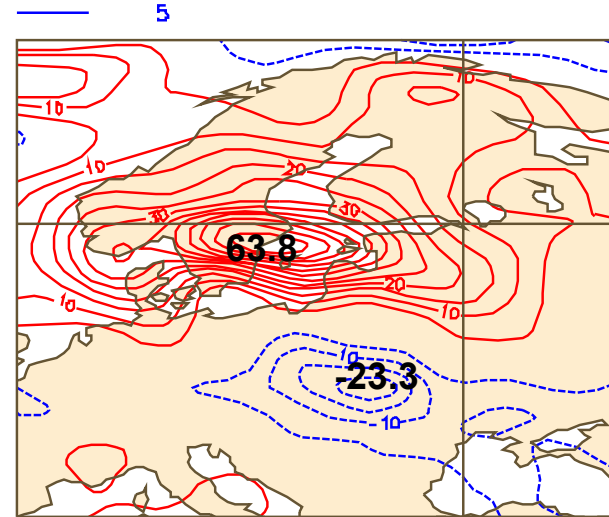
Impact of the linearized physics in 4D-Var (3)

1-DAY FORECAST ERROR OF 500 hPa GEOPOTENTIAL HEIGHT
OPER vs. NEWRAD (27/08/2001 t+24)

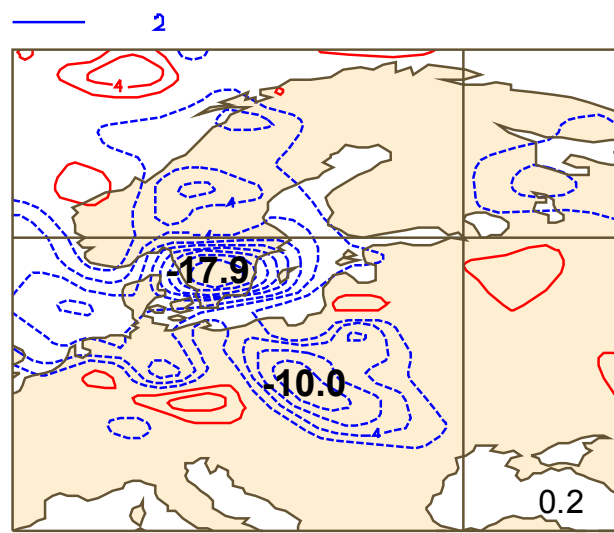
A1:
FC_OPER
-
ANAL_OPER



A2:
FC_NEWRAD
-
ANAL_OPER



A2 - A1



1D-Var assimilation of observations related to the physical processes

- For a given observation \mathbf{y}^o , 1D-Var searches for the model state $\mathbf{x}=(T, q_v)$ that minimizes the **objective function**:

$$\mathcal{J}(\mathbf{x}) = \underbrace{\frac{1}{2}(\mathbf{x} - \mathbf{x}^b)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}^b)}_{\text{Background term}} + \underbrace{\frac{1}{2}(H(\mathbf{x}) - \mathbf{y}^o)^T \mathbf{R}^{-1}(H(\mathbf{x}) - \mathbf{y}^o)}_{\text{Observation term}}$$

\mathbf{B} = background error covariance matrix

\mathbf{R} = observation and representativeness error covariance matrix

H = nonlinear observation operator (model space \rightarrow observation space)
(*physical parametrization schemes, microwave radiative transfer model, reflectivity model, ...*)

- The minimization requires an estimation of the **gradient of the objective function**:

$$\nabla \mathcal{J}(\mathbf{x}) = \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}^b) + \mathbf{H}^T \mathbf{R}^{-1}(H(\mathbf{x}) - \mathbf{y}^o)$$

- The operator \mathbf{H}^T can be obtained:
 - explicitly (Jacobian matrix)
 - using the adjoint technique

Precipitation assimilation at ECMWF

Goal: To assimilate observations related to precipitation and clouds in ECMWF's 4D-Var system including parameterizations of atmospheric moist processes.

More recent developments:

- New simplified convection scheme (*Lopez 2003*)
 - New simplified cloud scheme (*Tompkins & Janisková 2003*)
 - Microwave Radiative Transfer Model (*Bauer, Moreau 2002*)
- } used in 1D-Var
- Assimilation experiments of direct measurements from TRMM and SSM/I (TB or Z) instead of indirect retrievals of rainfall rates, in a '1D-Var + 4D-Var' framework.

"1D-Var+4D-Var" assimilation of observations related to precipitation

1D-Var on TBs or reflectivities

TMI TBs
or
TRMM-PR reflectivities

1D-Var on TMI or PR rain rates

Retrieval algorithm (2A12, 2A25)

"Observed" rainfall rates

Observations interpolated on model's T511 Gaussian grid

moist physics
+ radiative transfer

1D-Var

moist physics

background T, q_v

$$\text{"TCWV}_{obs} = \text{TCWV}_{bg} + m_r \delta q_v$$

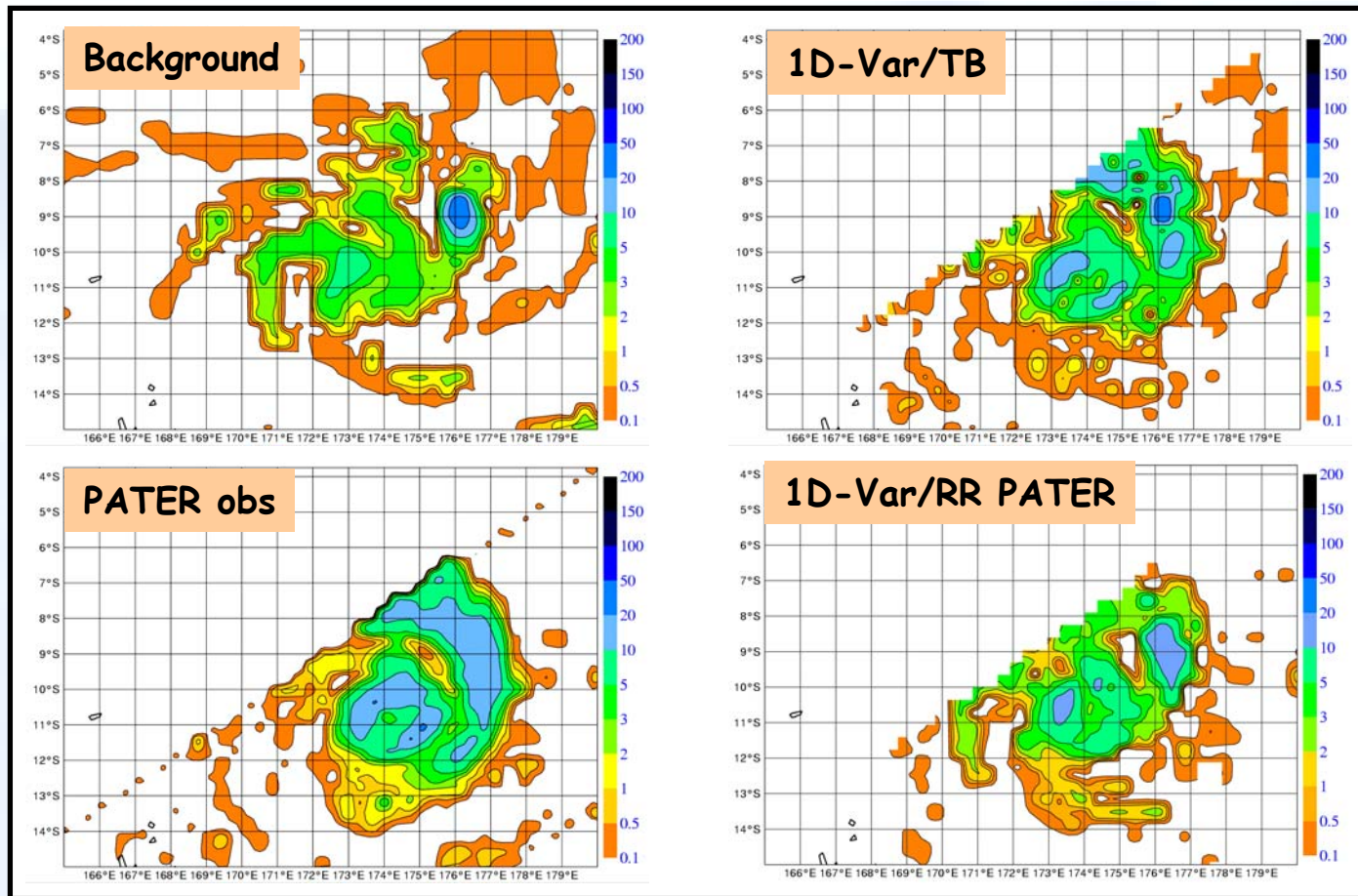
background T, q_v

4D-Var



1D-Var on TMI data

(Lopez and Moreau, 2003)

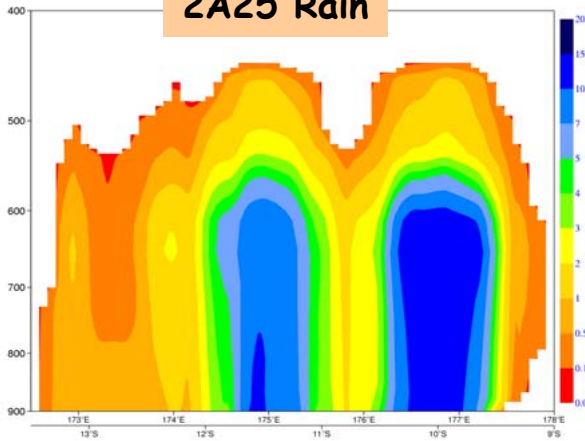


Tropical Cyclone Zoe (26 December 2002 @1200 UTC)

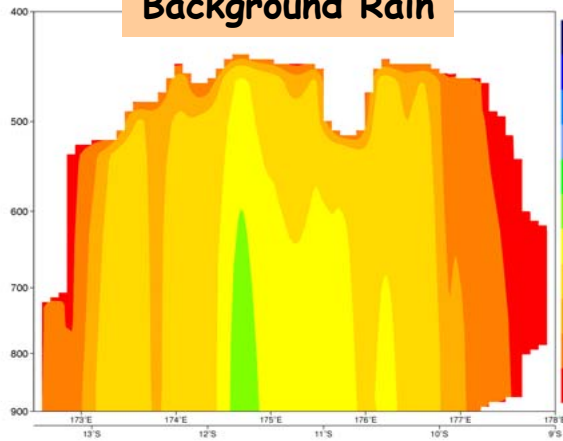
1D-Var on TMI Rain Rates / Brightness Temperatures

Surface rainfall rates (mm h⁻¹)

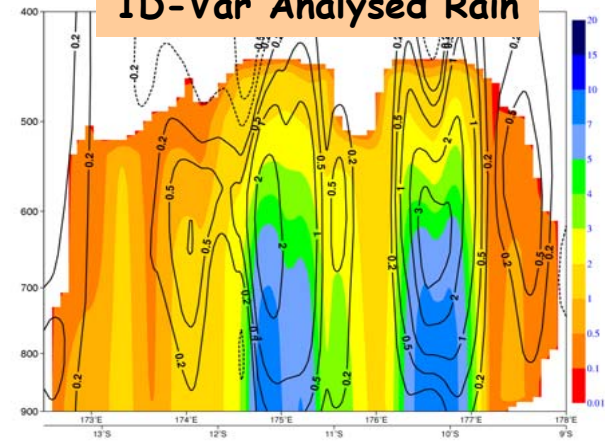
2A25 Rain



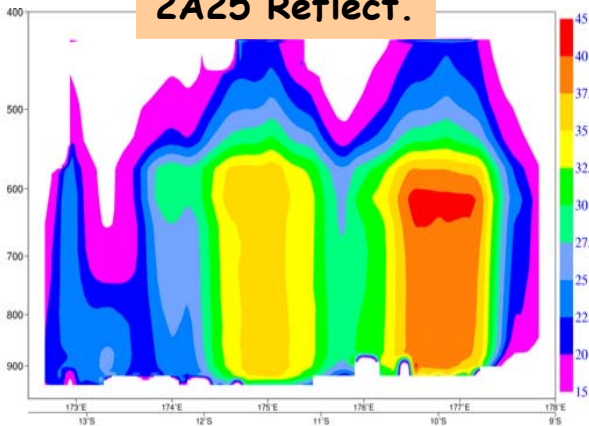
Background Rain



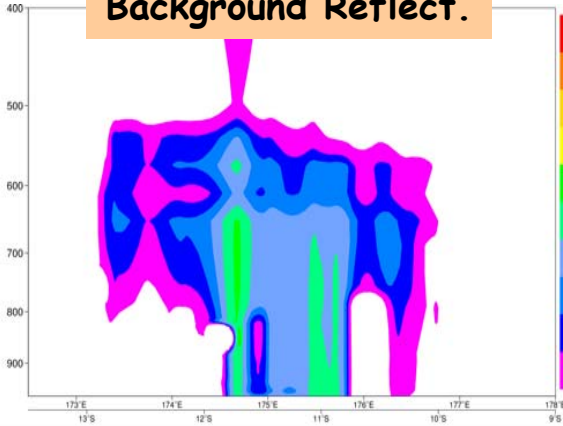
1D-Var Analysed Rain



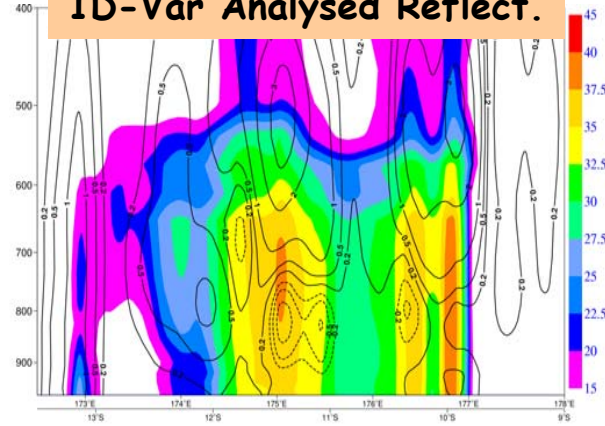
2A25 Reflect.



Background Reflect.



1D-Var Analysed Reflect.



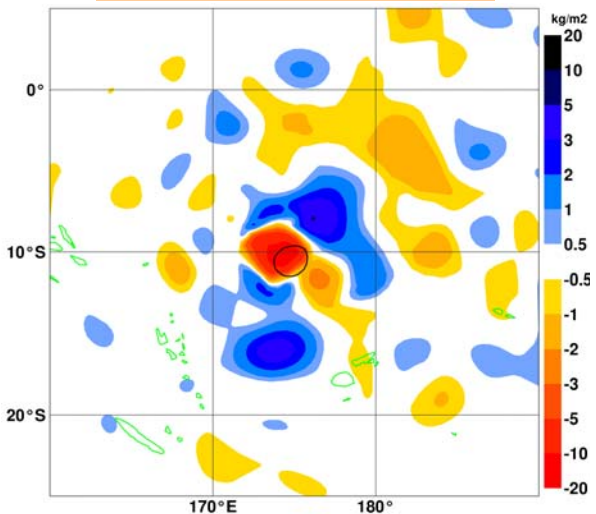
Tropical Cyclone Zoe (26 December 2002 @1200 UTC)

Vertical cross-section of rain rates (top, mm h^{-1}) and reflectivities (bottom, dBZ):
 observed (left), background (middle), and analysed (right).
 Black isolines on right panels = 1D-Var specific humidity increments.

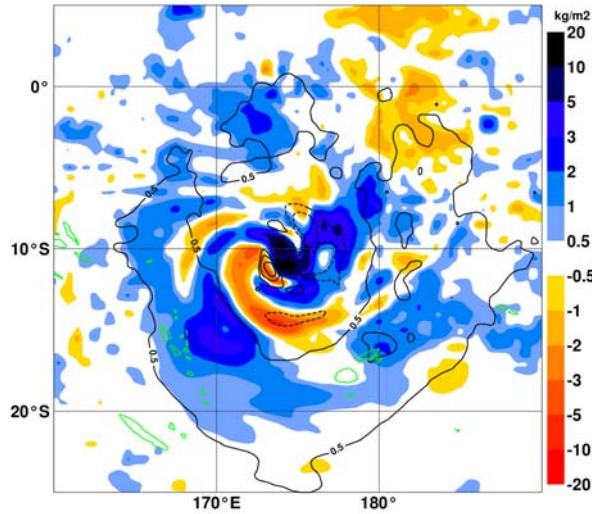
**“1D-Var+4D-Var” assimilation of TRMM-PR rain rates/reflectivities:
Impact on analysed and forecast TCWV and MSLP (Experiment – Control)
(Tropical Cyclone Zoe, 26-28 December 2002)**

**Analysis
26/12/2002 03UTC**

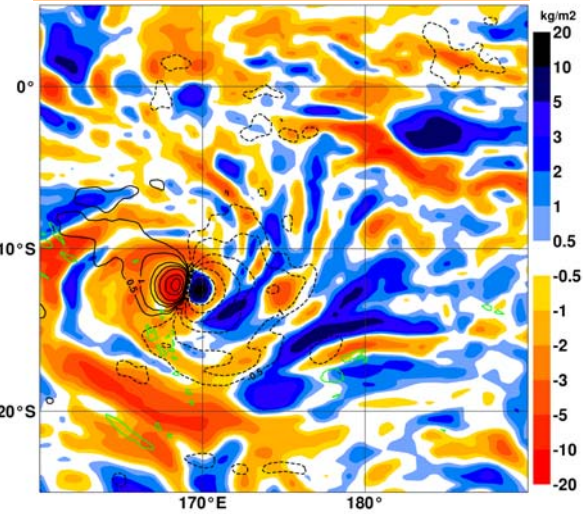
with PR rain rates



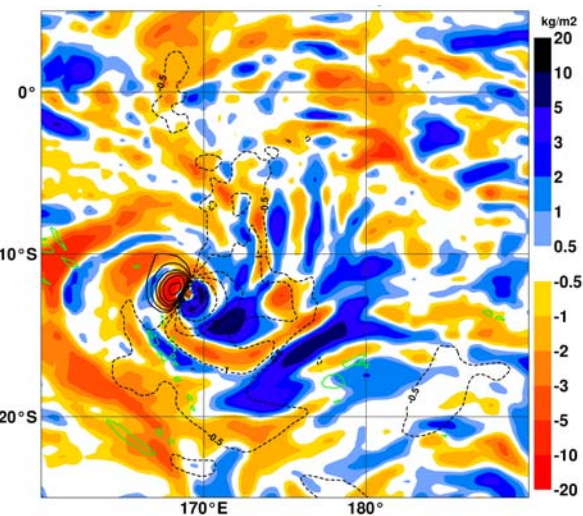
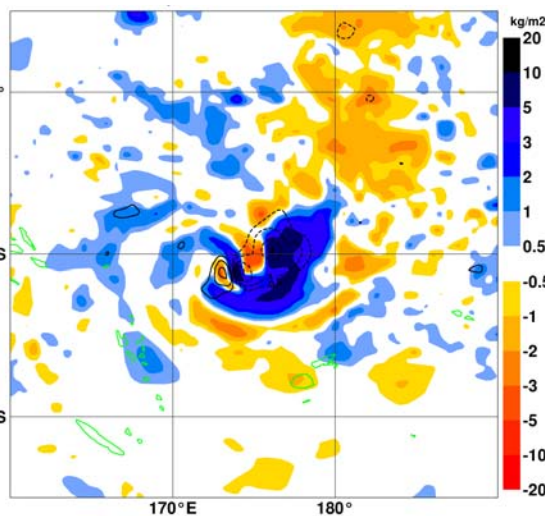
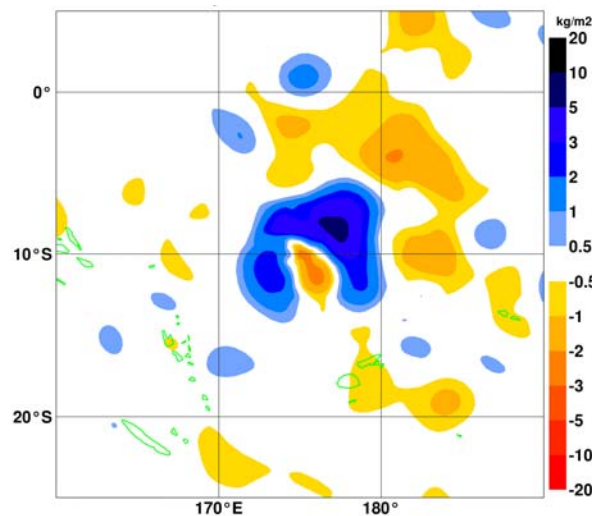
**Forecast
26/12/2002 03UTC t+9**



**Forecast
26/12/2002 03UTC t+57**



with PR reflectivities



1D-Var assimilation of ARM observations (1)

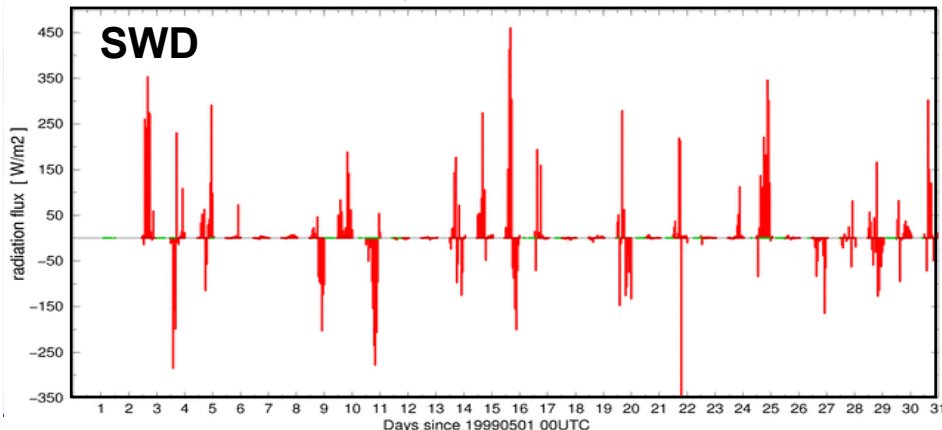
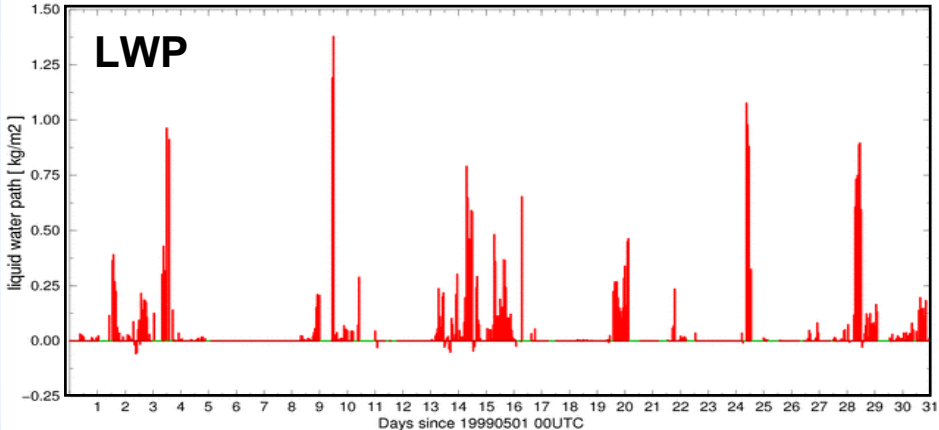
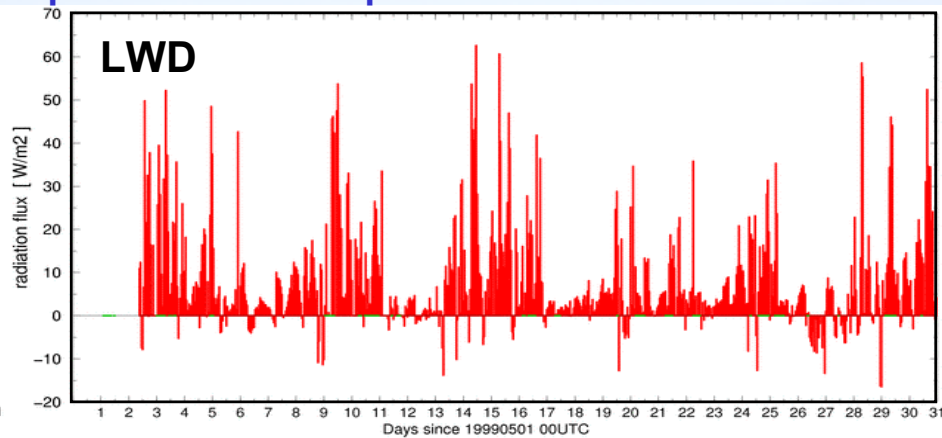
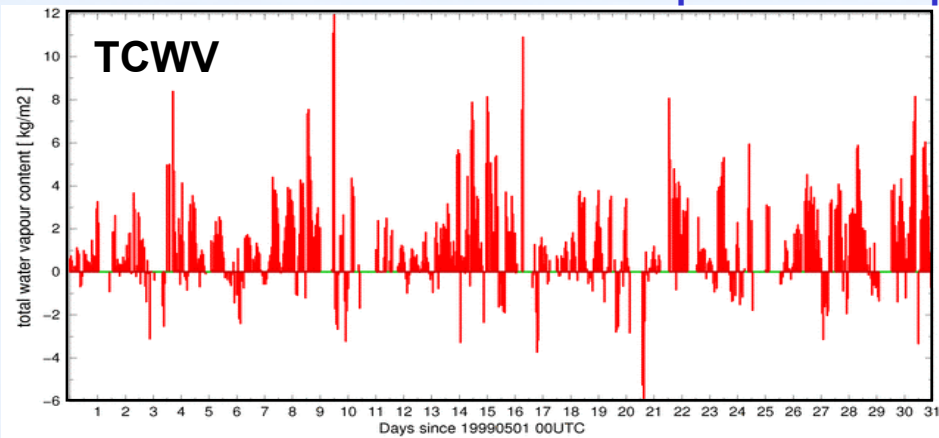
ARM SGP, May 1999 - observations:

- surface downward longwave radiation (LWD),
- total column water vapour (TCWV)
- cloud liquid water path (LWP)

Observation operator includes:

- shortwave and longwave radiation schemes
- diagnostic cloud scheme

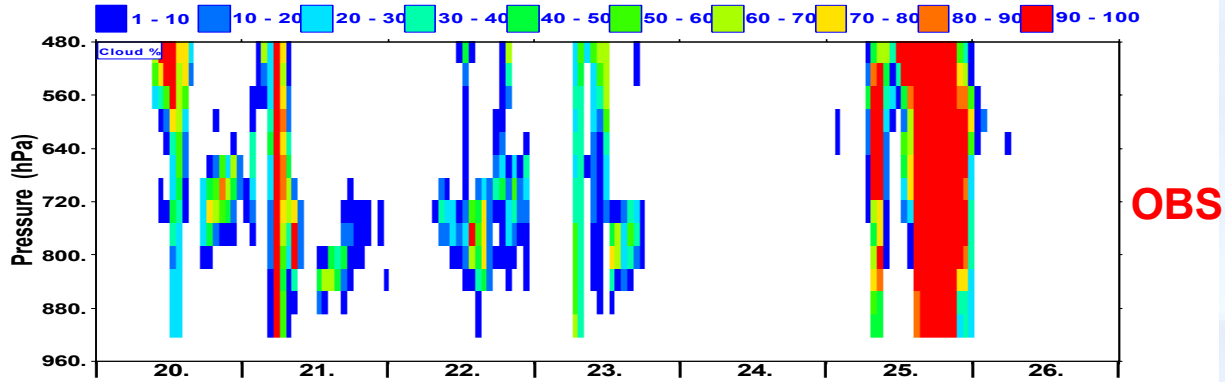
| FG - OBS | - | ANAL - OBS |



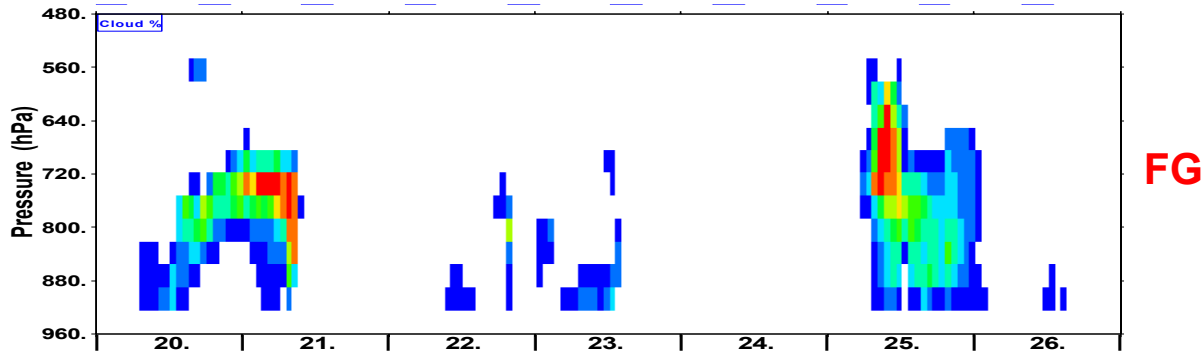
positive values = improvement

1D-Var assimilation of ARM observations (2)

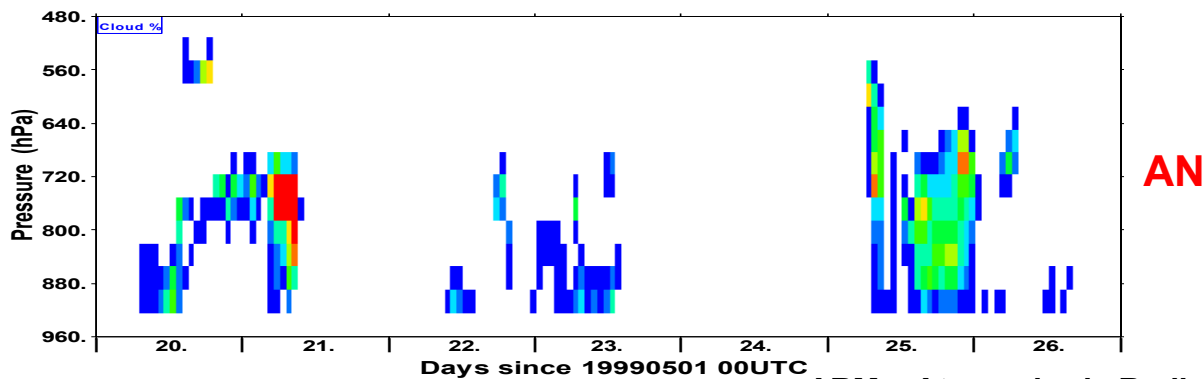
Time series of the cloud fraction (%) for the period 20-26 May 1999.



OBS

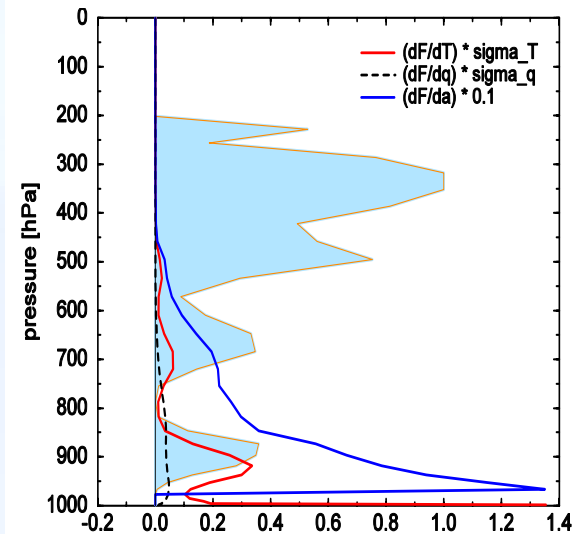


FG



AN

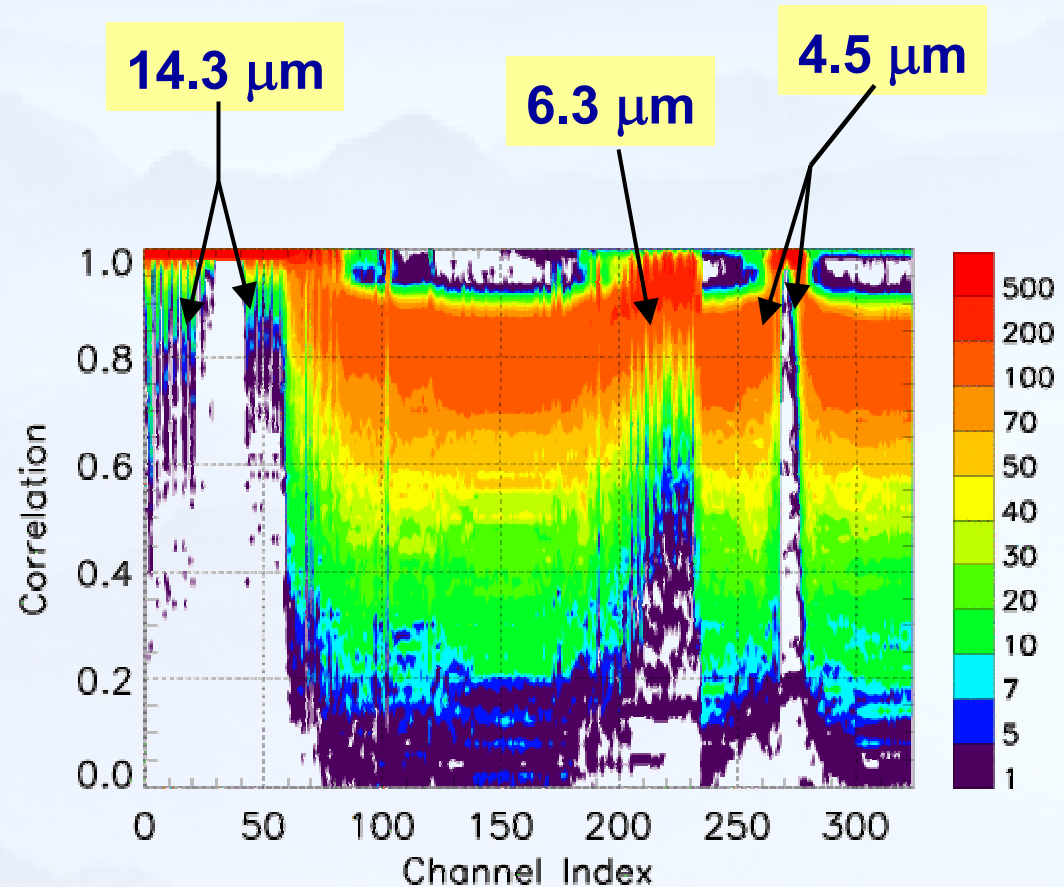
Sensitivity of LWD to T, q and cloud fraction



- 4D-Var assumes that the forward operator is linear in the vicinity of the background
 - ✓ Fairly true for cloudy upper tropospheric channels

PDF of correlations between $\mathbf{H} \cdot \delta \mathbf{x}$ and $H[\mathbf{x} + \delta \mathbf{x}] - H[\mathbf{x}]$
Hemispheric data

- $\mathbf{x} = T, q$ profile
- $H =$ cloud scheme + RTTOVCLD
- $\delta \mathbf{x} =$ perturbation

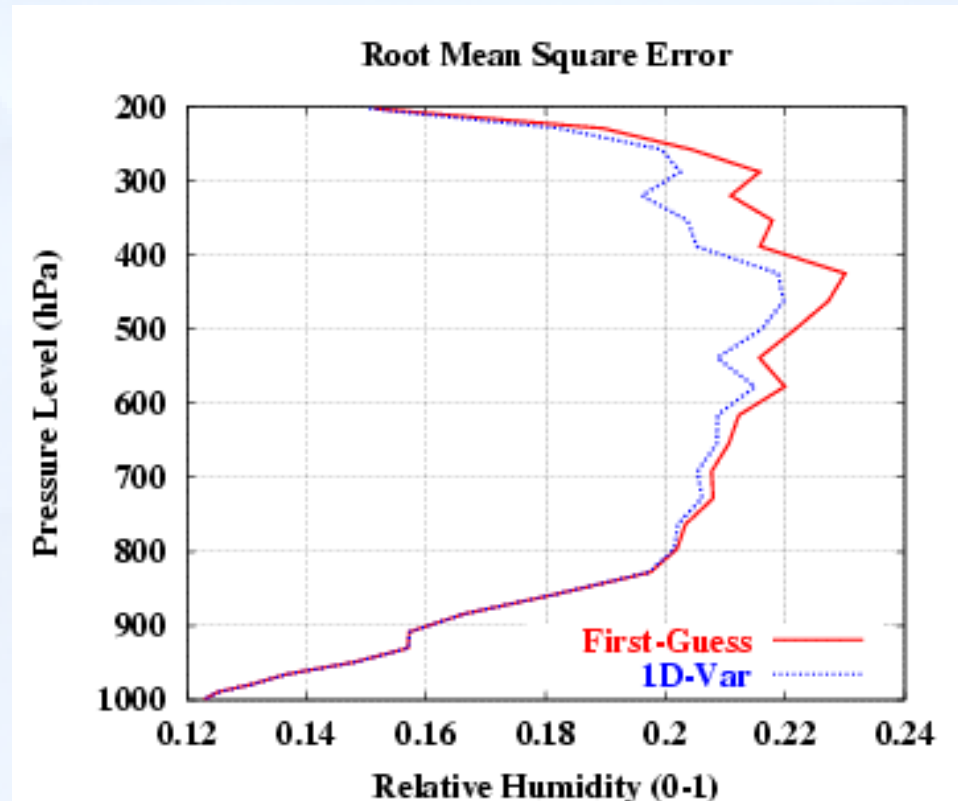


- Linear 1D-Var retrievals
 - ✓ observations = 35 upper tropospheric AIRS channels
 - ✓ performed only if clouds are detected in more than 13 channels

Validation:

1D-Var vs European radiosondes Nov 2002 and Feb 2003

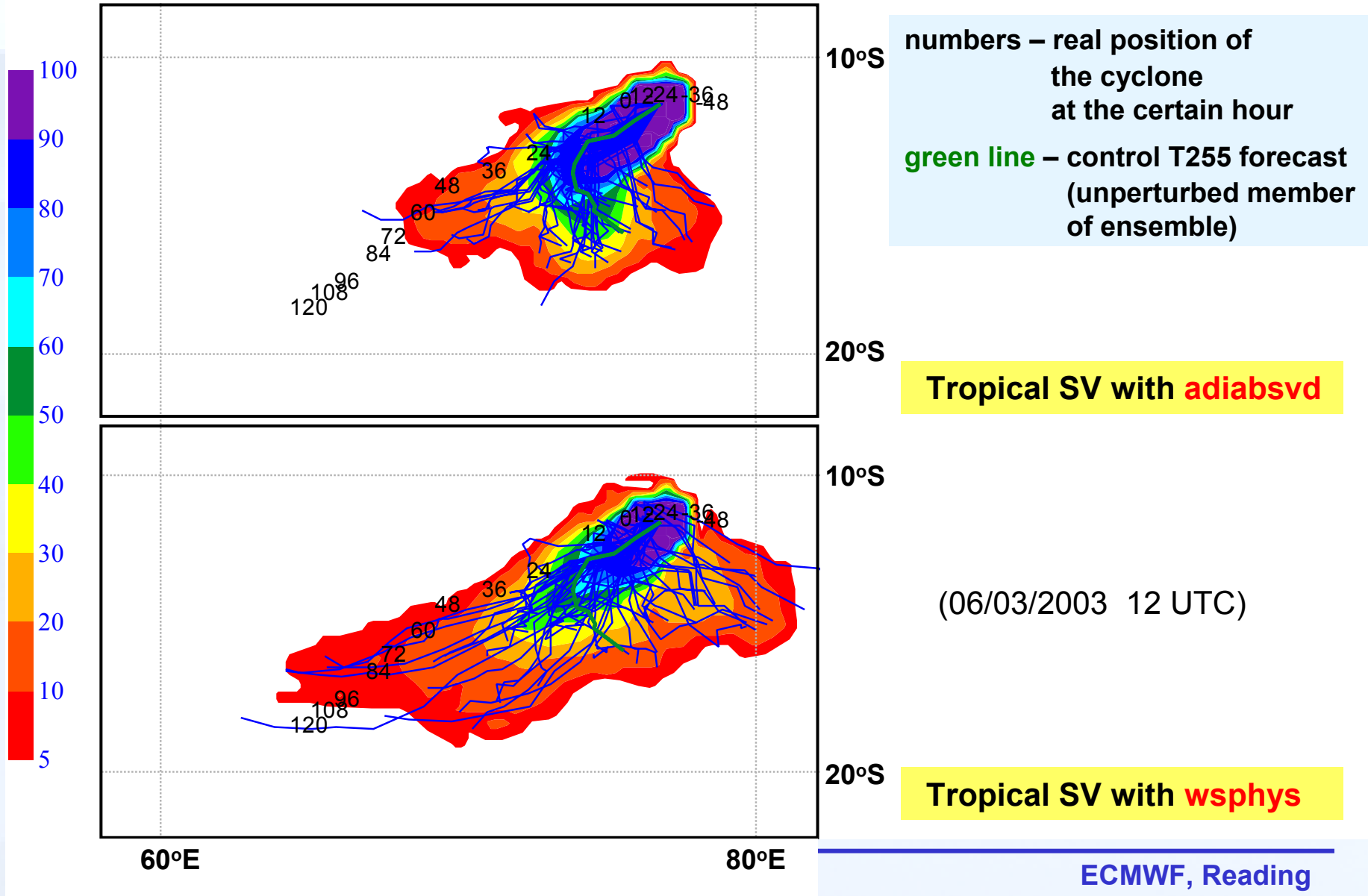
- If $T < 243\text{K}$ use Vaisala RS90 only
- ~ 250 matches in upper troposphere
- ~ 2300 matches in lower troposphere



Tropical singular vectors

(Leutbecher and Van Der Grijn, 2003)

Probability that the cyclone KALUNDE will pass within 120 km radius during the next 120 hours



Sensitivity of the parametrization scheme to input variables

- using the adjoint technique
- adjoint \mathbf{F}^T of the linear operator \mathbf{F} provides the gradient of an objective function \mathcal{J} with respect to \mathbf{x} (*input variables*) given the gradient of \mathcal{J} with respect to \mathbf{y} (*output variables*):

$$\frac{\partial \mathcal{J}}{\partial \mathbf{x}} = \mathbf{F}_x^T \frac{\partial \mathcal{J}}{\partial \mathbf{x}} \quad \text{or} \quad \nabla_{\mathbf{x}} \mathcal{J} = \mathbf{F}_x^T \nabla_{\mathbf{y}} \mathcal{J}$$

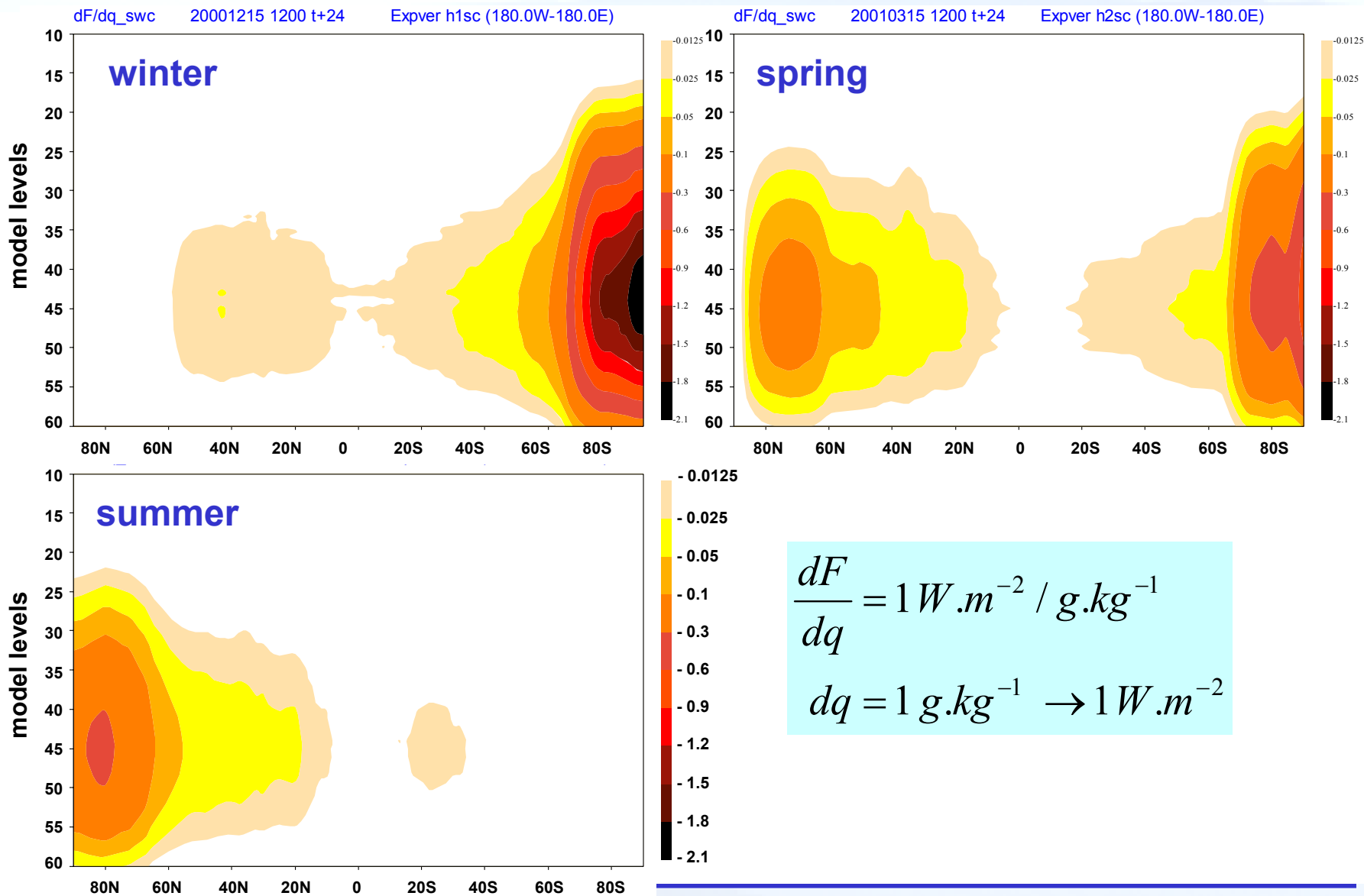
EXAMPLE: sensitivity of radiation scheme - the gradient with respect to \mathbf{y} of unity size (i.e., *perturbation of some of the radiation fluxes with $\pm 1 \text{ W.m}^{-2}$*)

$$\nabla_{\mathbf{x}} \mathcal{J} = \frac{\partial F}{\partial \mathbf{x}} \left\{ \begin{array}{l} \partial F / \partial T \quad \text{sensitivity to: temperature} \\ \partial F / \partial q \quad \text{spec. humidity} \\ \partial F / \partial a \quad \text{cloud cover} \\ \partial F / \partial q_{lw} \quad \text{cloud lwc} \\ \partial F / \partial q_{iw} \quad \text{cloud iwc} \end{array} \right.$$

- experiments done in the global model:
 - potential for a thorough evaluation of the relative importance of different variables for parametrization scheme
 - investigation of spatial and temporal patterns of sensitivity variations

Sensitivity of the shortwave upward radiation flux at the TOA with respect to specific humidity [W.m⁻²/g.kg⁻¹]

CLEAR SKY



$$\frac{dF}{dq} = 1 \text{ W.m}^{-2} / \text{g.kg}^{-1}$$

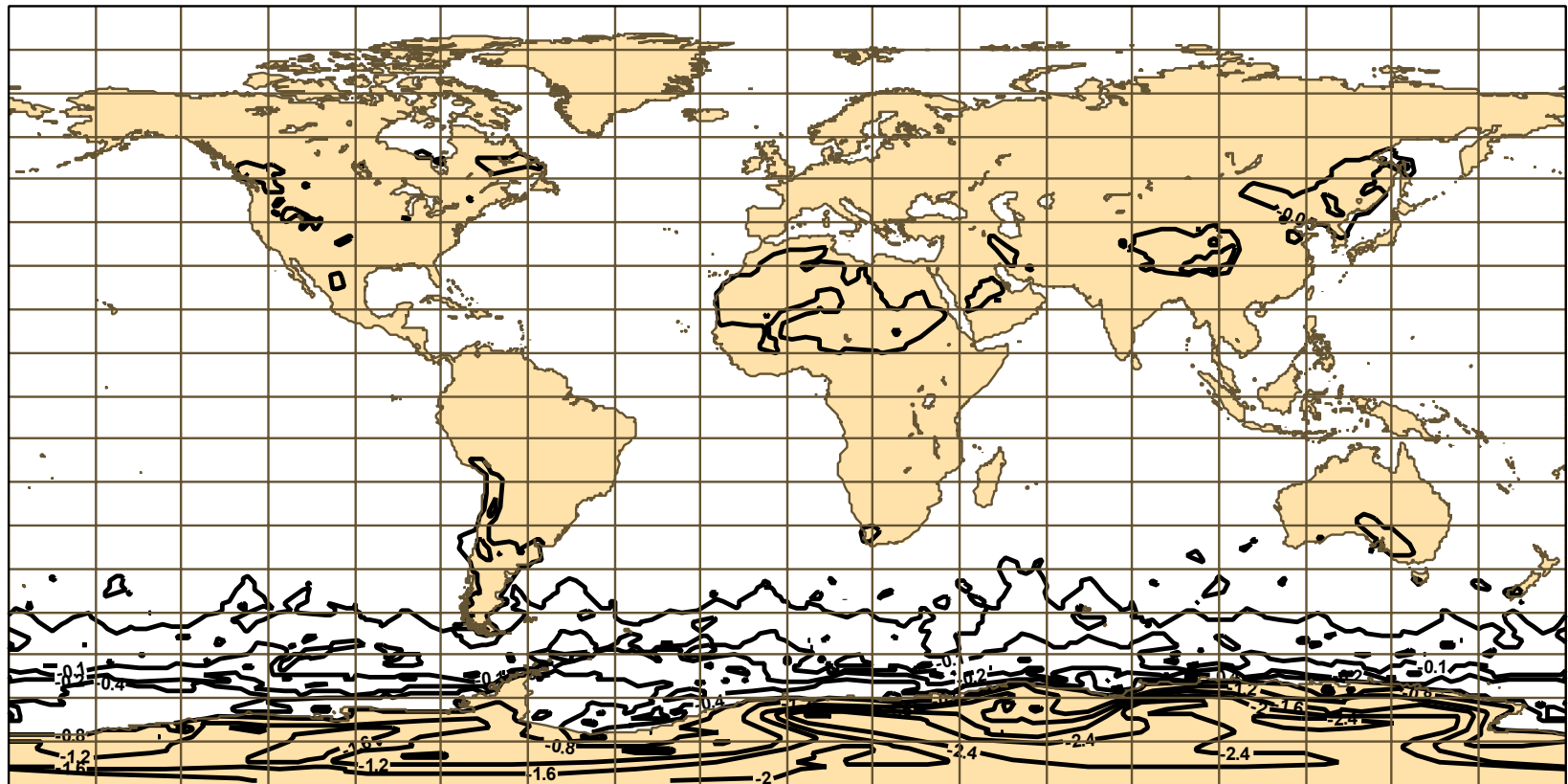
$$dq = 1 \text{ g.kg}^{-1} \rightarrow 1 \text{ W.m}^{-2}$$

Sensitivity of the shortwave upward radiation flux at the TOA with respect to specific humidity [W.m⁻²/g.kg⁻¹]
CLEAR SKY

Level 44 ~ 700 hPa

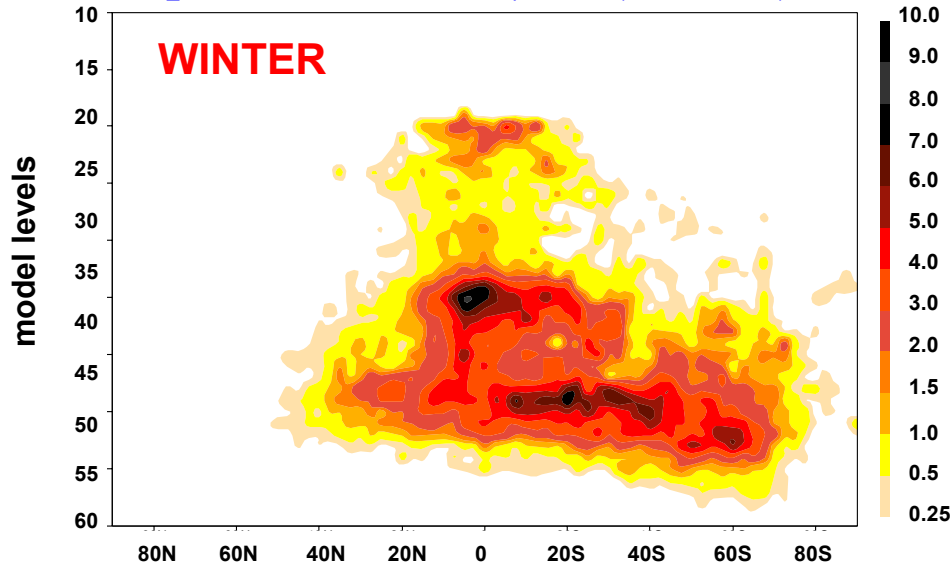
WINTER

dF/dq_{swc} 15 December 2000 12UTC ECMWF t+24 Level 44



Sensitivity of the shortwave/longwave upward radiation flux at the TOA with respect to cloud fraction [W.m⁻²/cloudfr]

dF/dc_sw 20001215 1200 t+24 Expver h1sw (180.0W-180.0E)

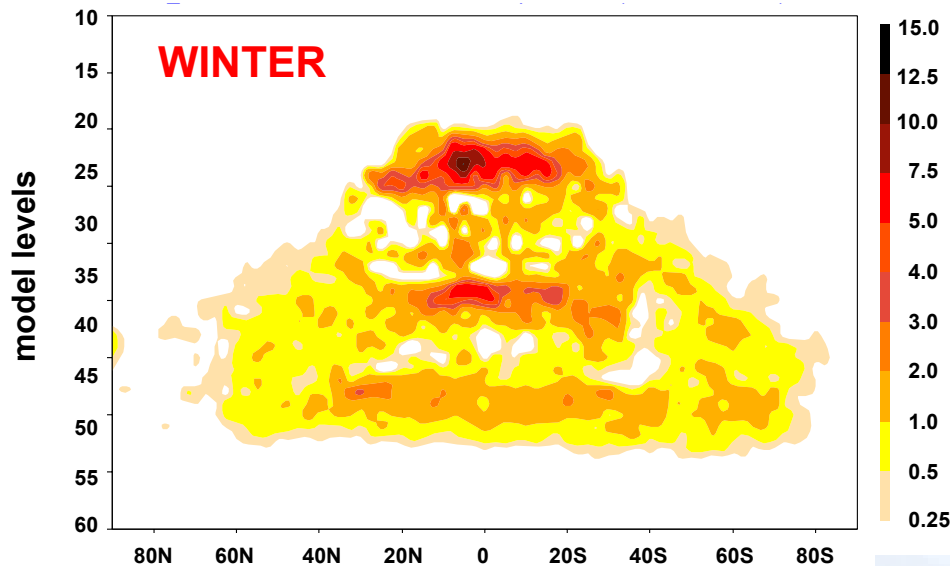


shortwave radiation

$$\frac{dF}{dc} = 5 \text{ W.m}^{-2} / \text{cloudfr}$$

$$dc = 1 \rightarrow 5 \text{ W.m}^{-2}$$

$$dc = 0.2 \rightarrow 1 \text{ W.m}^{-2}$$



longwave radiation

Summary

- **Positive impact from including linearized physical parametrization schemes**
(into the assimilating model, singular vector computations used in EPS)
has been demonstrated in experimental and operational runs.
- **Adjoint of physical processes can also be used for sensitivity studies and model parameter estimation.**
- **Physical parametrizations become important components in recent variational data assimilation systems.**
- **Some care must be taken when deriving the linearized parametrization schemes**
(regularizations/simplifications).
- **This is particularly true for the assimilation of observations related to precipitation, clouds and soil moisture, to which a lot of effort is currently devoted.**
- **One cannot also forget technical difficulties and time-consuming adjoint development** → *reliable and efficient automatic tool for adjoint coding would be useful.*

ZERO BENEFIT



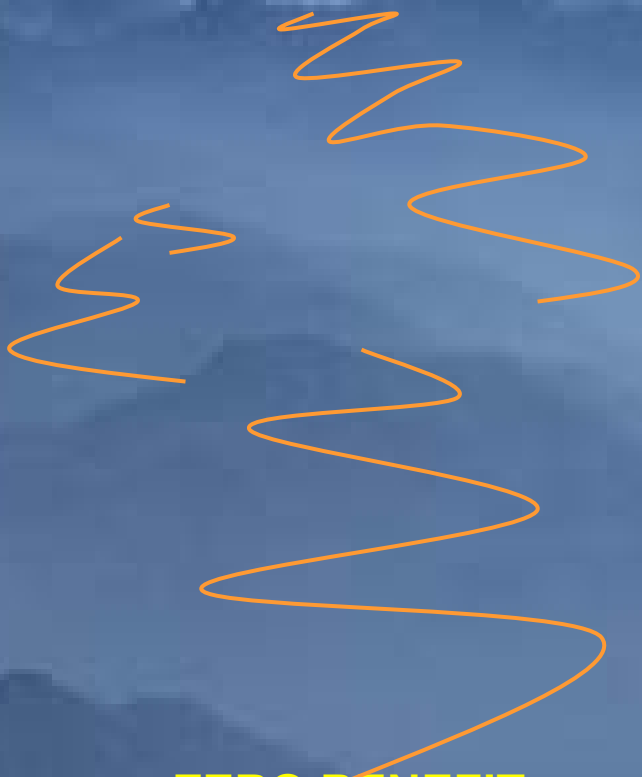
ZERO BENEFIT



ZERO BENEFIT



**WE ARE
HERE**



ZERO BENEFIT

**MAXIMUM
POTENTIAL BENEFIT**

**WE ARE
HERE**

ZERO BENEFIT



FORECAST VERIFICATION – 500 hPa GEOPOTENTIAL

period: 11/05/2001 – 26/05/2001

(4D- Var experiments with the new linearized radiation: **lin_rad**)

root mean square error – 16 cases

Northern Hemisphere

North America

