

# **Assimilation of Radio Occultation Data at DWD**

Andreas Rhodin

*andreas.rhodin@dwd.de*

Deutscher Wetterdienst, Offenbach, Germany

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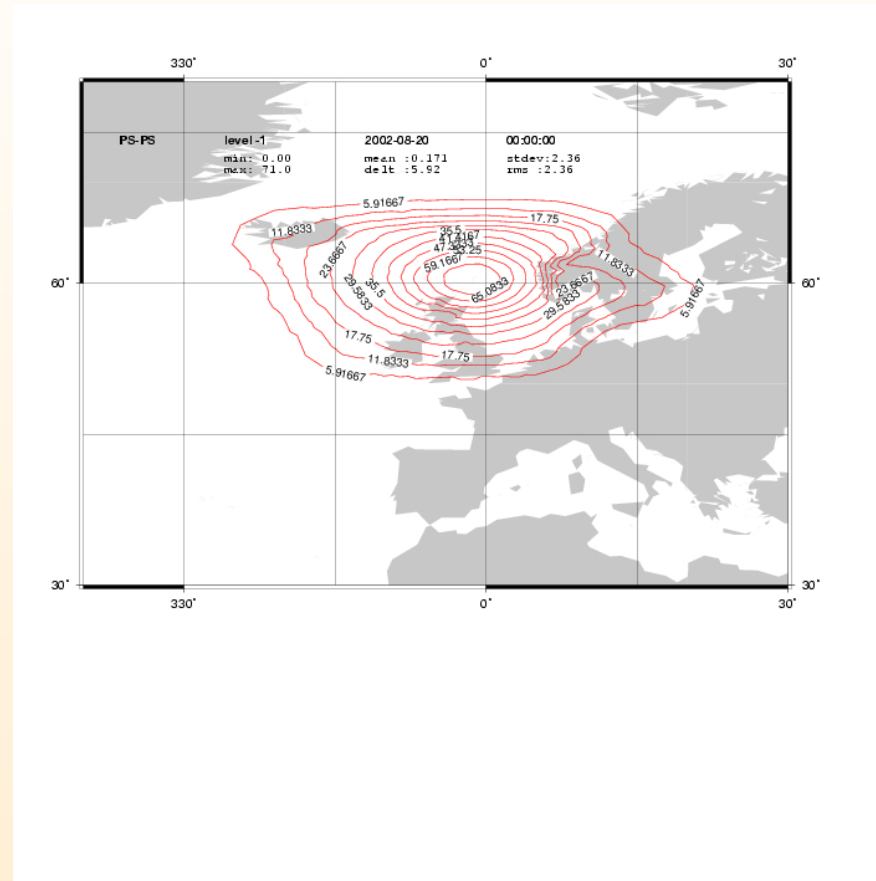
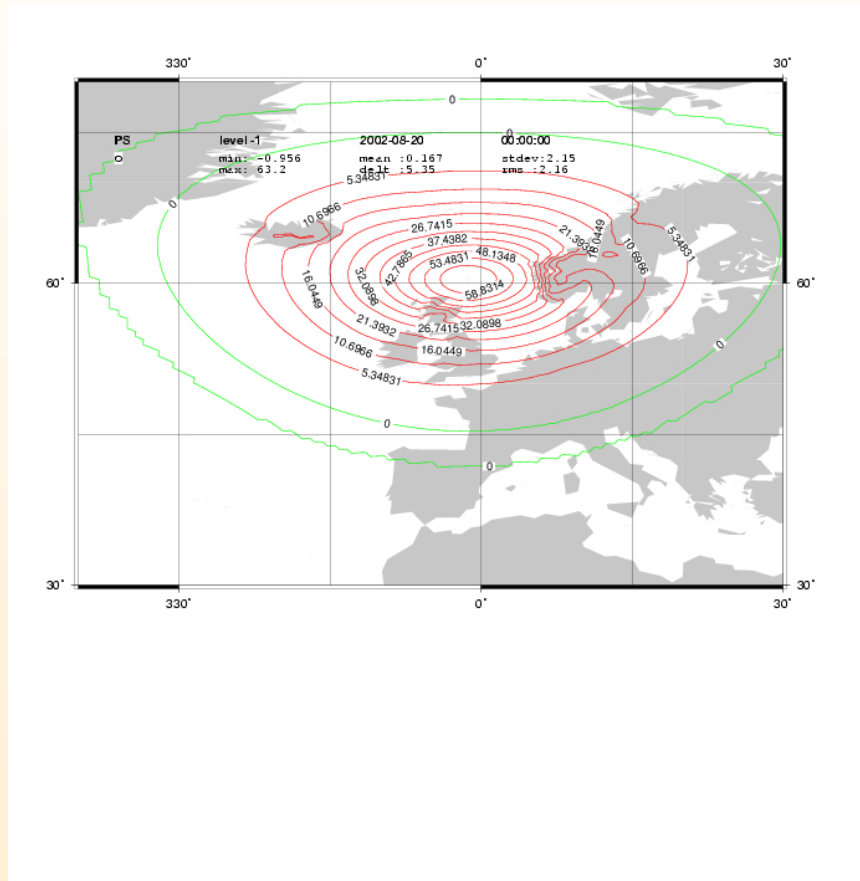
# DWD 3D-Var

- A 3D-Var system is build up in order to replace the operational OI. The new system is more appropriate for assimilation of remote sensing data.
- The control variables are still chosen in observation space (OSAS/PSAS). This approach provides full flexibility for the specification of the background error covariance matrices.
- The development builds up on activities started at MPIfM, including the implementation of a GPS Ray-tracing Operator (Michael Gorbunov, Luis Kornblueh).  
The Ray-tracing Operator will remain in the DWD 3D-Var System as a prototype for a limb-sounding operator.

## Status

- Up to now the background error model of the OI are used.
- In situ observation operators are implemented.
- (Re)implementation of the GPS ray-tracer operator is under way.
- Some of the deficiencies of the OI (due to local solution of the equations in observation boxes) are not present in the 3D-Var.

# Response to a single pressure observation



Surface pressure analysis increment for a single observation in the 3D-Var (left) and OI (right).

## 3D-Var Formulation

$$J(\mathbf{x}) = \frac{1}{2} \left( (\mathbf{x} - \mathbf{x}_b) \mathbf{P}_b^{-1} (\mathbf{x} - \mathbf{x}_b) + (H(\mathbf{x}) - \mathbf{o}) \mathbf{R}^{-1} (H(\mathbf{x}) - \mathbf{o}) \right)$$

**R:** Observation error covariance matrix:

Sparse, Only data within one observation are correlated.

**P<sub>b</sub>:** Forecast Error Covariance Matrix

Dense within an influence radius of some 1000 km, zero outside

- size of  $\mathbf{x}$  :  $n \approx 10^7 \dots 10^8$
- size of  $\mathbf{P}_b$  :  $n^2 \approx 10^{15}$

ca. 10% nonzero elements.

# OSAS - Observation Space Assimilation System

Cost function:

$$J(\mathbf{x}) = \frac{1}{2} \left( (\mathbf{x} - \mathbf{x}_b) \mathbf{P}_b^{-1} (\mathbf{x} - \mathbf{x}_b) + (H(\mathbf{x}) - \mathbf{o}) \mathbf{R}^{-1} (H(\mathbf{x}) - \mathbf{o}) \right)$$

Gradient of the cost function:

$$\frac{d}{d\mathbf{x}} J = \mathbf{H}^T \mathbf{R}^{-1} [H(\mathbf{x}) - \mathbf{o}] + \mathbf{P}_b^{-1} [\mathbf{x} - \mathbf{x}_b]$$

$\mathbf{H}$  is the Jacoby matrix of the operator  $H$ .

In the minimum of  $J$  its gradient is zero and the analyzed state  $\mathbf{x}_a$  is:

$$\mathbf{x}_a - \mathbf{x}_b = [\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H} + \mathbf{P}_b^{-1}]^{-1} \mathbf{H}^T \mathbf{R}^{-1} [\mathbf{o} - H(\mathbf{x}_b)]$$

Algebraic manipulation yields:

$$\mathbf{x}_a - \mathbf{x}_b = \mathbf{P}_b \mathbf{H}^T [\mathbf{H} \mathbf{P}_b \mathbf{H}^T + \mathbf{R}]^{-1} [\mathbf{o} - H(\mathbf{x}_b)]$$

For in situ observations of model variables  $\mathbf{H} \mathbf{P}_b \mathbf{H}^T$  is the background correlation matrix at observation points (denoted  $\mathbf{B}$  in the following).

# Solution of the OSAS equation

$$\mathbf{x}_a - \mathbf{x}_b = \mathbf{P}_b \mathbf{H} (\mathbf{B} + \mathbf{R})^{-1} (\mathbf{o} - H(\mathbf{x}_b))$$

with  $\mathbf{B} = \mathbf{H}^T \mathbf{P}_b \mathbf{H}$ .

$\mathbf{B}$  is the forecast error covariance matrix for the observed quantities.

$(\mathbf{B} + \mathbf{R})$  is a symmetric and positive definite Matrix.

The set of equations is solved in two steps:

1. Solve the system of linear equations:

$$\mathbf{z} = (\mathbf{B} + \mathbf{R})^{-1} (\mathbf{o} - H(\mathbf{x}_b))$$

2. Perform the matrix vector multiplication:

$$\mathbf{x}_a - \mathbf{x}_b = (\mathbf{P}_b \mathbf{H}) \mathbf{z}$$

# CG algorithm - Preconditioning - Parallelization

In order to solve the system of linear equations

$$\mathbf{z} = (\mathbf{B} + \mathbf{R})^{-1}(\mathbf{o} - H(\mathbf{x}_b))$$

a preconditioned Conjugate Gradient (CG) algorithm is used.

For preconditioning, an approximation to  $(\mathbf{B} + \mathbf{R})^{-1}$  is required:

- The matrix  $\mathbf{B}$  is decomposed into blocks with size  $\approx 500 \times 500$ . Matrix elements are sorted, so that strongly correlated elements (nearby observations) are located within the same block.
- The block-diagonal matrix  $\tilde{\mathbf{B}}$  is used for the preconditioner  $(\tilde{\mathbf{B}} + \mathbf{R})^{-1}$ . Block diagonals are inverted by Cholesky decomposition.
- The block structure is the basis of parallelization. Matrix blocks are distributed over different processor elements.
- The block structure can also be used for a single observation mode (similar to 1D-Var) by taking only one observation per box.



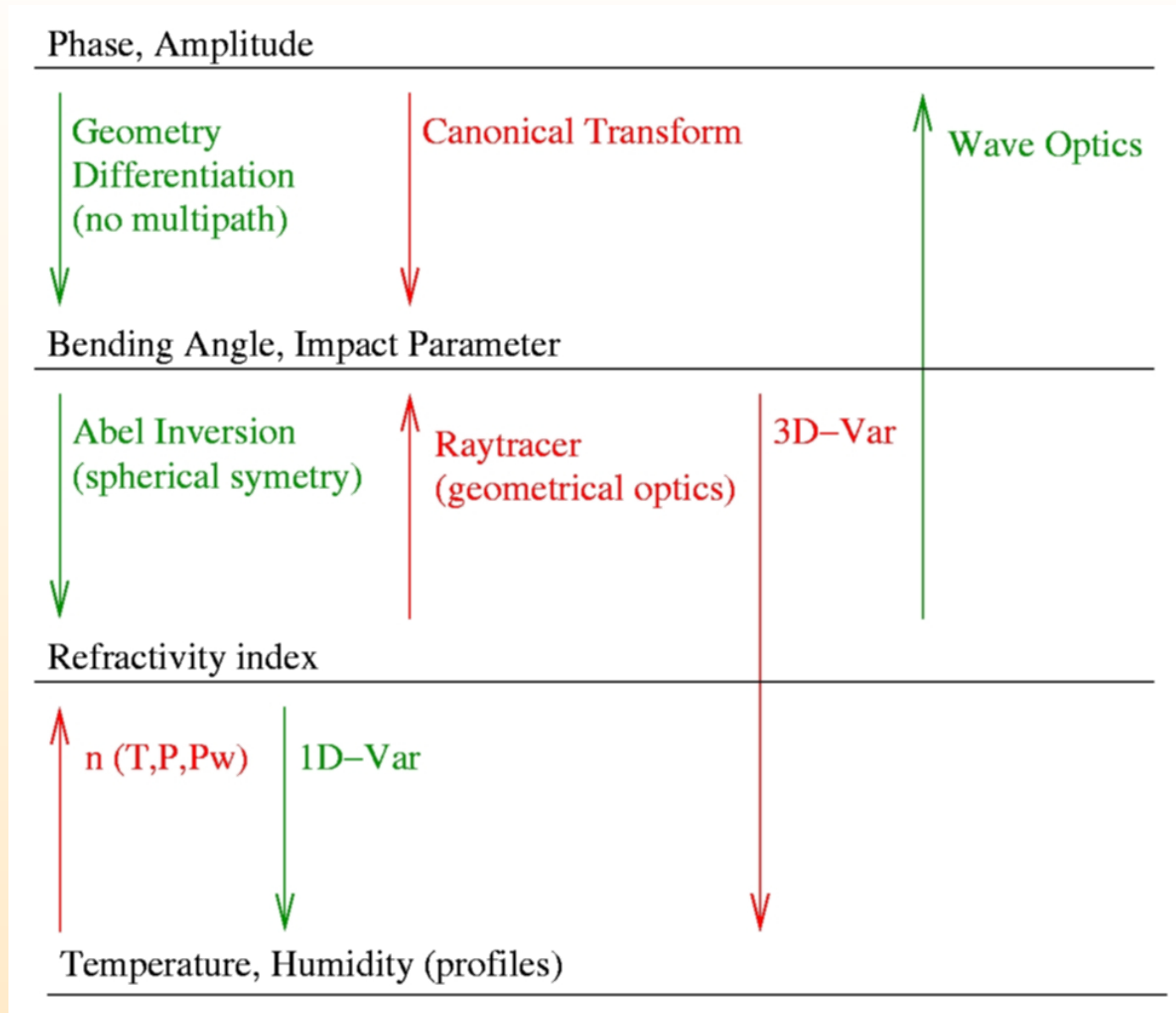
# Nonlinear Observation Operators

$$\mathbf{z} = (\mathbf{B} + \mathbf{R})^{-1}(\mathbf{o} - H(\mathbf{x}_b))$$

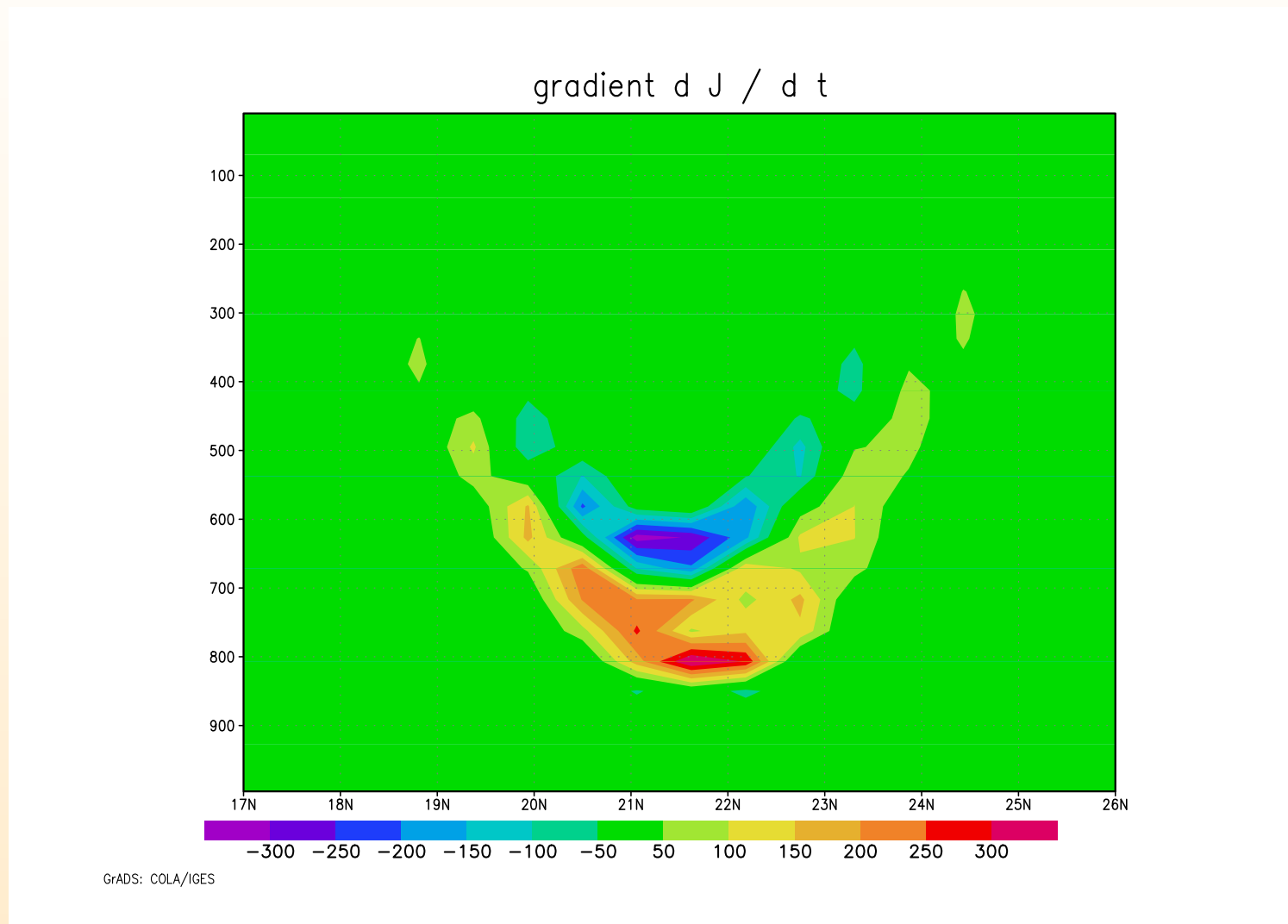
- For nonlinear observation operators  $H$  and non-Gaussian error distributions (variational quality control) the linear OSAS equation only holds approximately.
- The CG algorithm serves as a Newton step within an outer loop.
- Due to nonlinearities in  $H$  and  $\mathbf{R}$ , the matrices  $\mathbf{R}$  and  $\mathbf{B} = \mathbf{H}\mathbf{P}_b\mathbf{H}^T$  must be updated in each outer iteration.

Evaluation of the nonlinear terms in general is expensive.

# GPS Radio Occultations - Observables

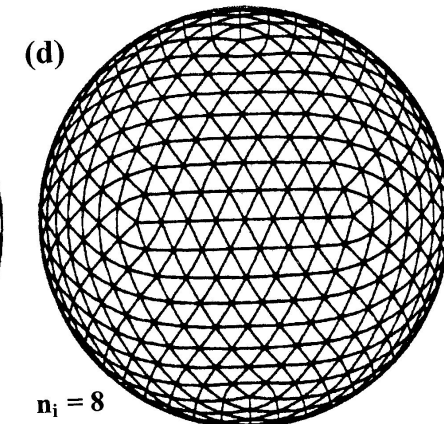
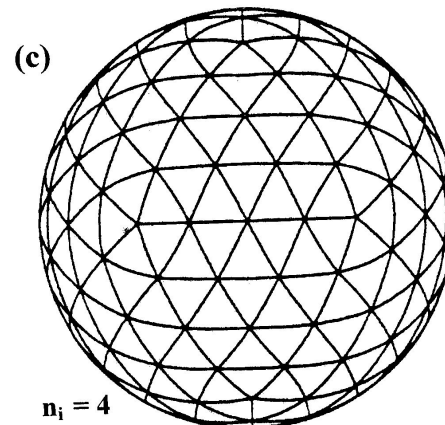
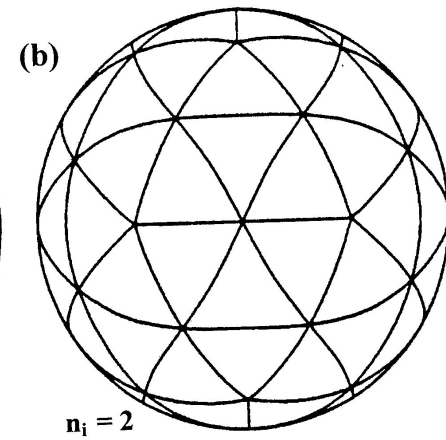
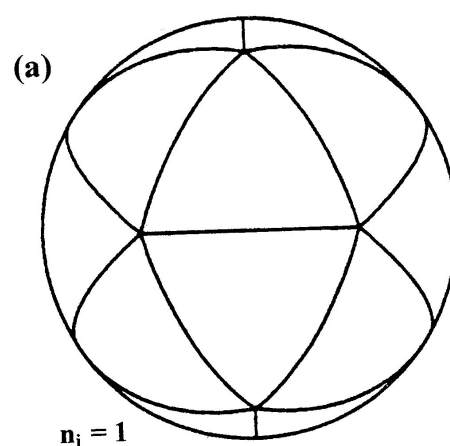
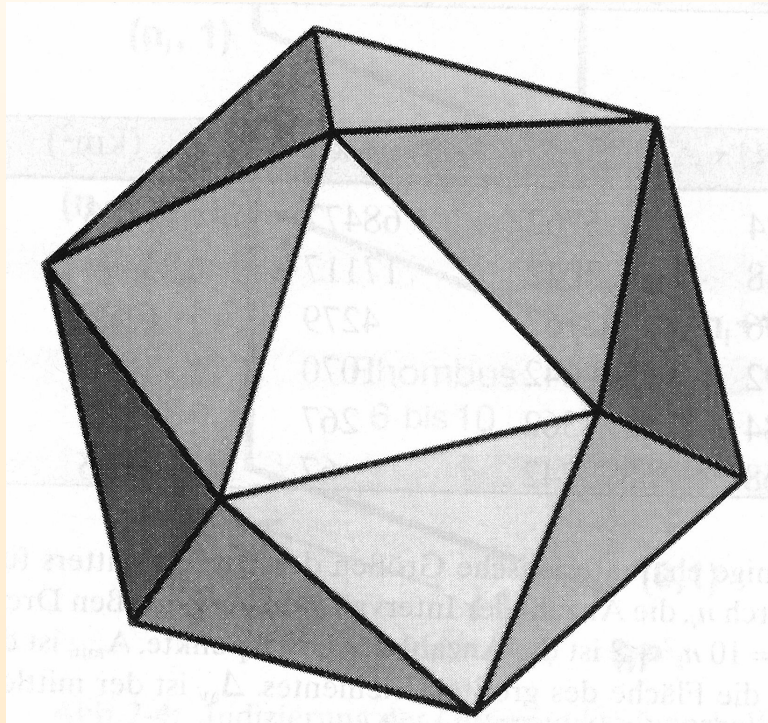


# GPS Raytracer - Sensitivity

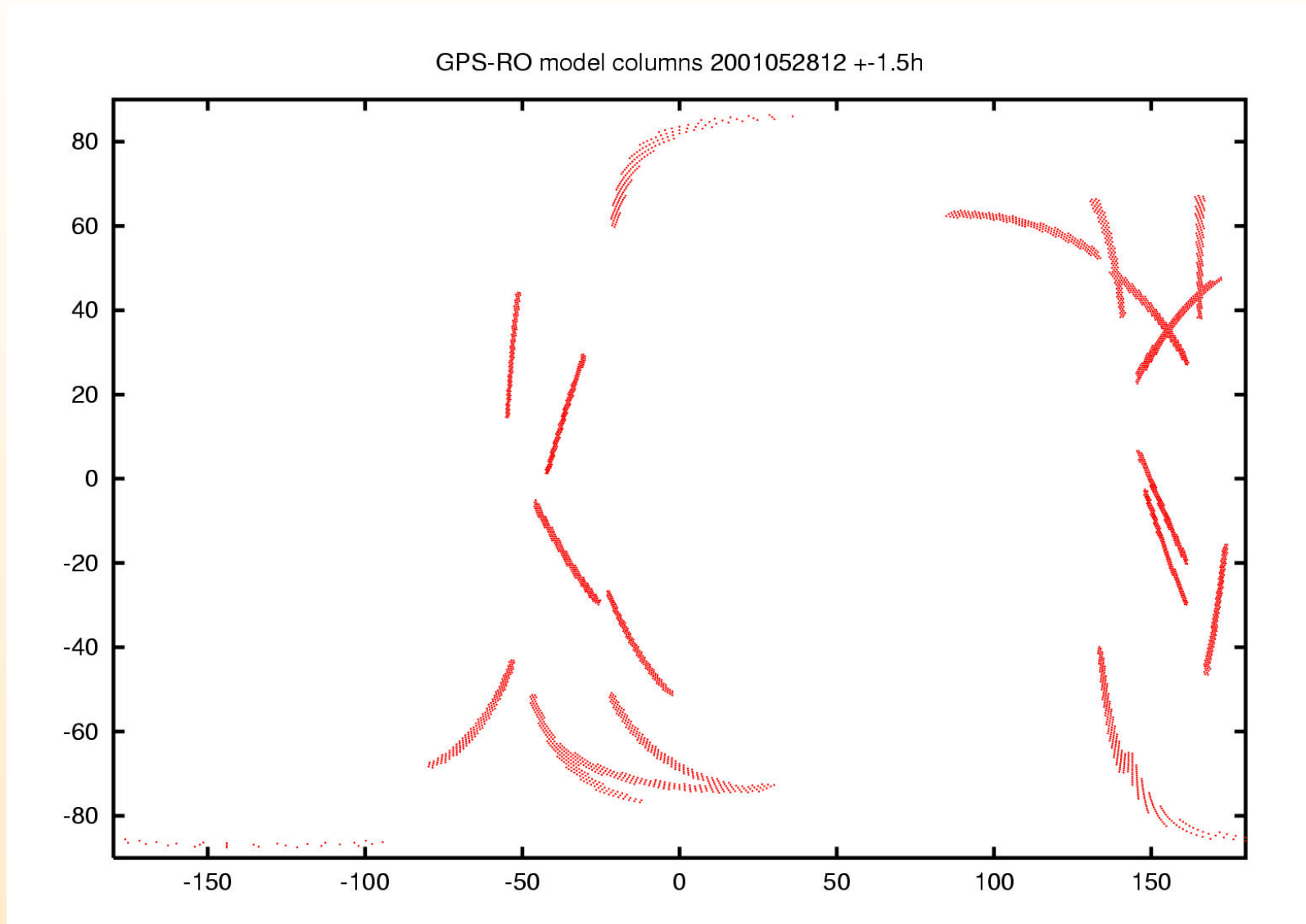


Gradient of the cost function with respect to the atmospheric temperature for a ray-tracer observation operator

# Global Model GME - Icosahedral Grid



# Selected model columns



Model columns (3277 out of 166410) required to run the GPS-ray-tracer operator for 19 occultations observed by CHAMP within an assimilation window of 3 hours.

# Ray-tracing operator - Implementation

Preprocessor:

Derive  $\epsilon(p)$  by the CT method

Observation Operator:

$$H(\mathbf{x}) = \epsilon(\mathbf{x}, p)$$

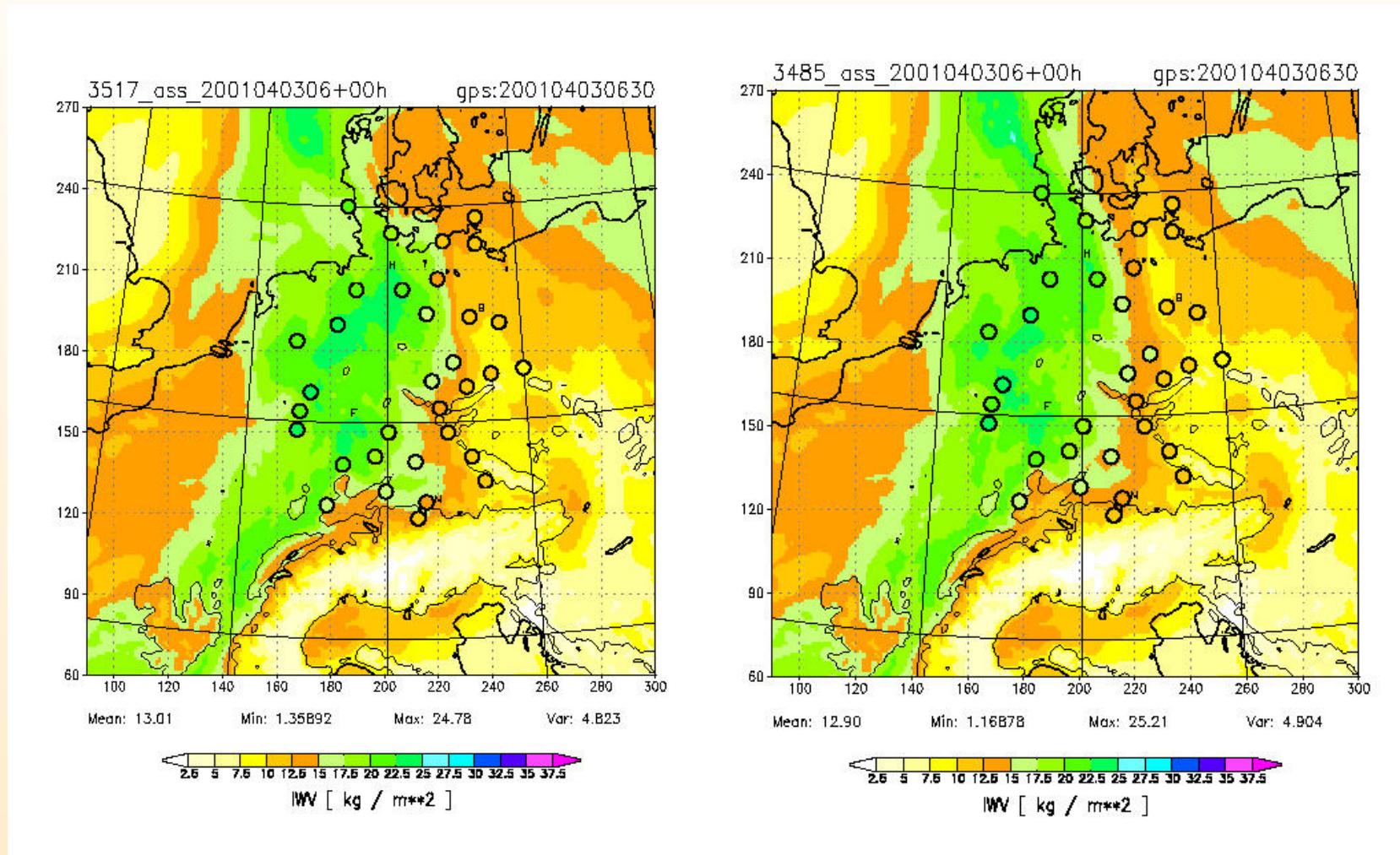
Implementation:

1. Identify model columns the observation operator  $H$  depends on.
2. Gather model columns required by  $H$  on one processor element.
3. Run the observation operator and its adjoint.
4. Solve the linearized problem  $\mathbf{z} = (\mathbf{B} + \mathbf{R})^{-1}(\mathbf{o} - H(\mathbf{x}_b))$ .
5. Post multiplication: project guess to the model columns required.  
goto 3 .
6. Post multiplication: project guess to the whole model grid.

# Additional Staff for DWD 3D-Var

1 scientist	5 years	Assimilation of GPS data: integrated water vapour radio occultations
3 scientists	3 years	3D-Var: observation operators numerical solvers forecast error estimation

# Impact of additional GPS integrated water vapour data



Integrated water vapour in the local model LM without (left) and with (right) assimilation of GPS data



# Expectations to the GRAS SAF

- Test of less sophisticated observation operators.
  - ▷ *Is running a ray-tracing observation operator worth the effort?*
- What kind of observational errors shall be assigned to the bending angle observations?  
What are the errors of the bending angles provided by the CT method?
- Quality control: Can outliers be identified a priori?
- Data in real time.
  - ▷ *The time window for observations currently is  $2\frac{1}{2}$  hours at 0 and 12 UTC.*