

Practical instructions for solving MWR forward and inverse problems

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Contributions from:

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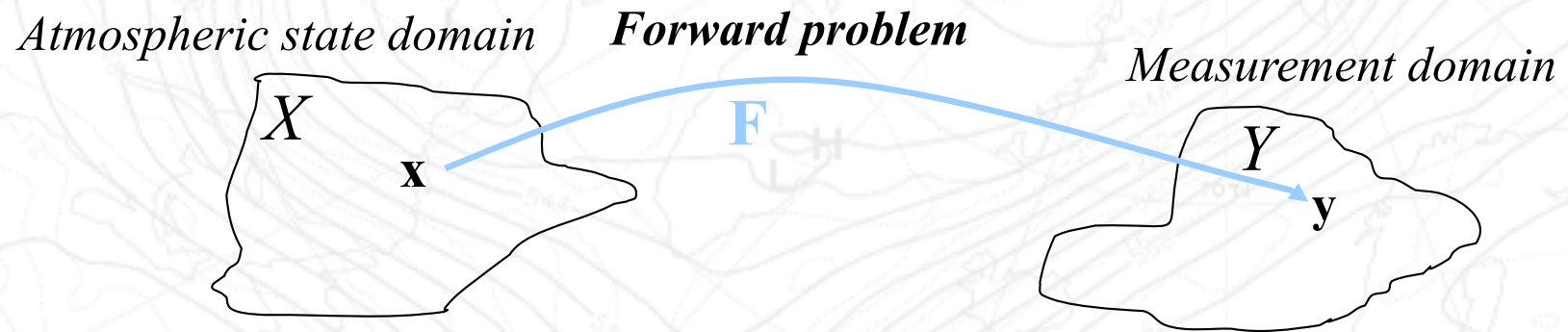
Practical instructions for MWR forward and inverse problems

- ❑ Introduction
 - Forward and inverse problems
- ❑ Software tools
 - RTTOV-gb
 - 1DVAR
 - Net1D
- ❑ Summary and practical information

Introduction

Forward and inverse problems

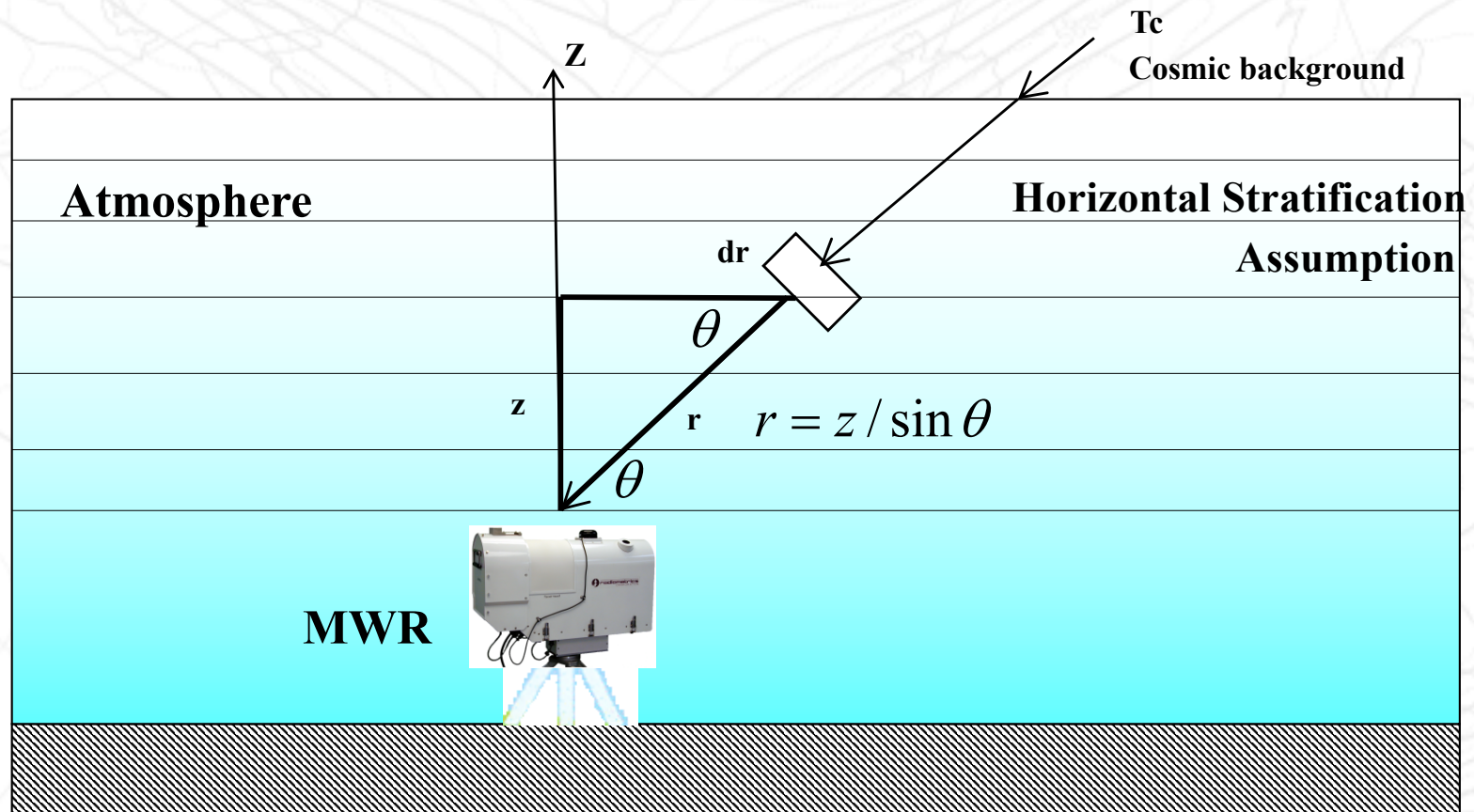
- Going from “atmospheric state \rightarrow Tb” is called **Forward Problem**



Introduction

Forward and inverse problems

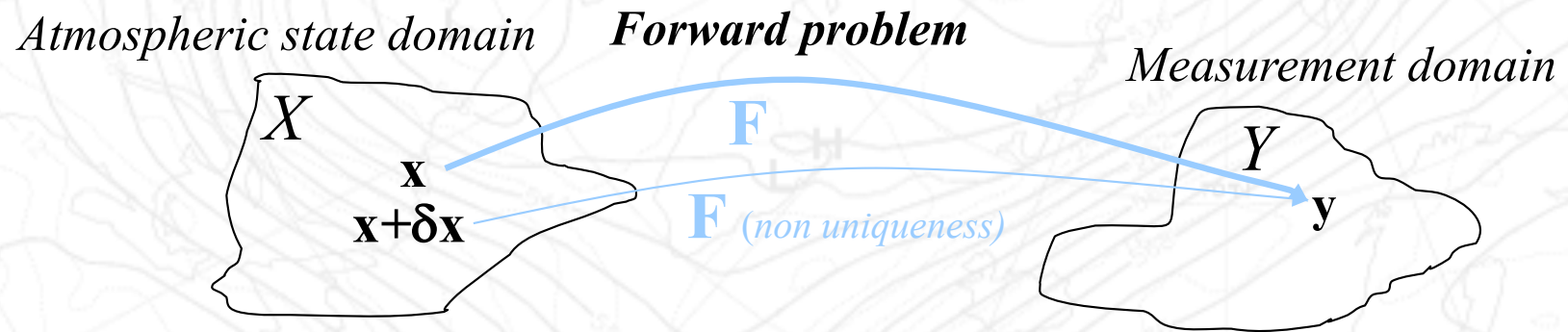
$$\underbrace{T_b}_{\text{Measured quantity}} = \underbrace{T_c \cdot e^{-\tau(0,\infty)}}_{\text{Cosmic background term (attenuated along the path)}} + \underbrace{\int_0^\infty K_E(r) \cdot T(r) \cdot e^{-\int_0^r K_E(r') dr'} \cdot dr}_{\text{Atmospheric term (emitted \& attenuated along the path)}}$$



Introduction

Forward and inverse problems

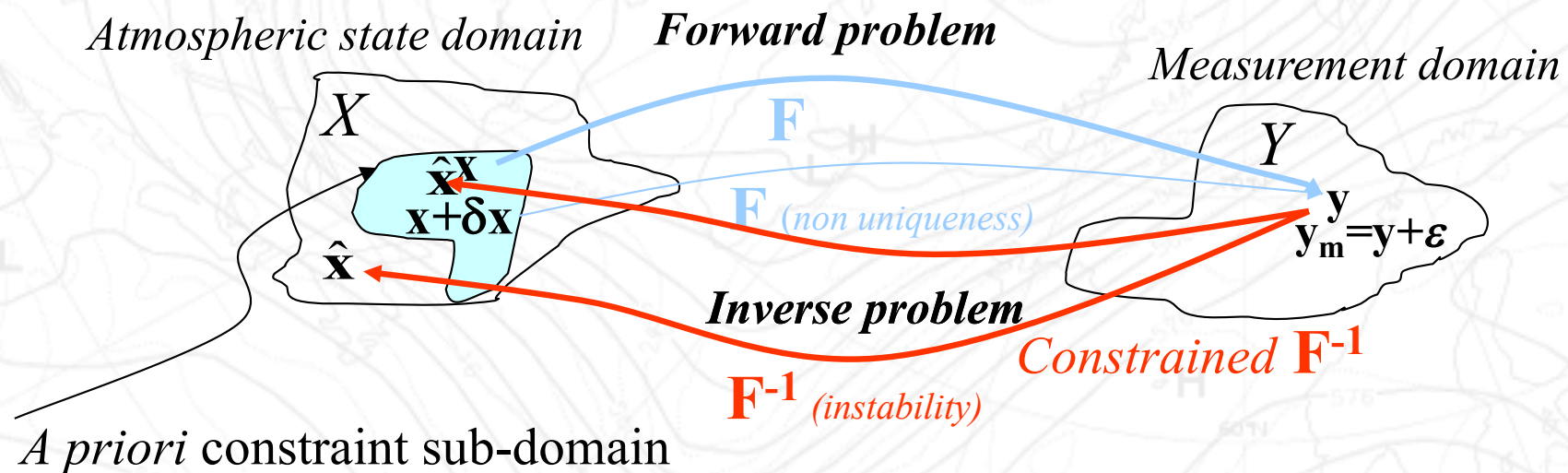
- Going from “atmospheric state \rightarrow Tb” is called **Forward Problem**



Introduction

Forward and inverse problems

- Going from “atmospheric state \rightarrow Tb” is called **Forward Problem**

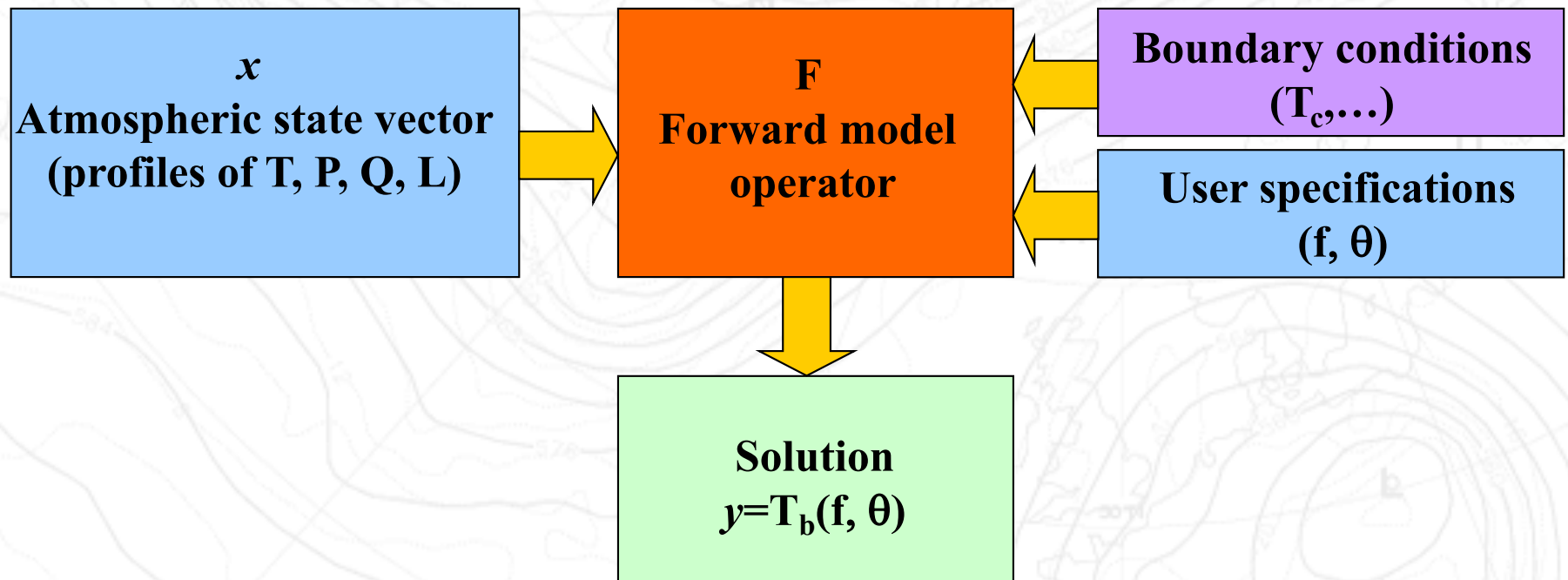


- Going from “Tb \rightarrow Atmospheric state” is called **Inverse Problem**
 - Ill-posed problem
 - solution is not unique/stable
 - Need for a priori knowledge to constrain the solution

Introduction

Forward and inverse problems

- ❑ The Forward Problem can be modeled in a straightforward manner
- ❑ Forward model (FM) operator scheme



- ❑ Errors
 - Discretization errors
 - Absorption model uncertainties
 - Other assumptions (scattering, plane parallel,...)

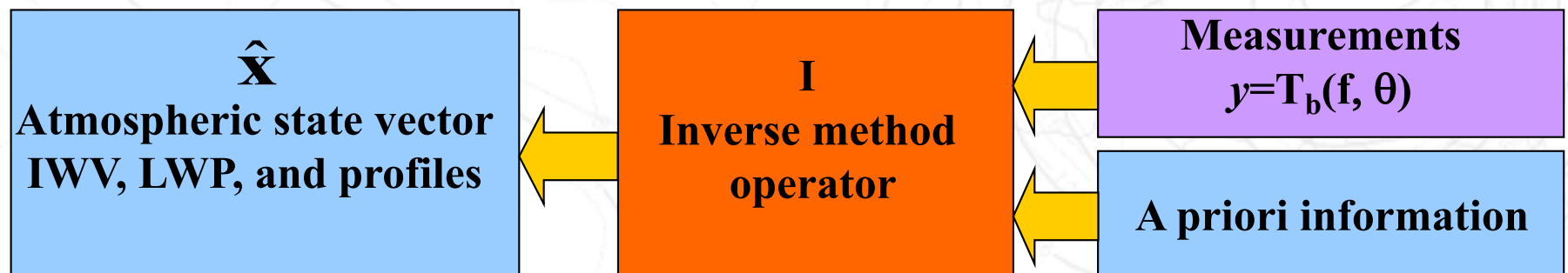
Introduction

Forward and inverse problems

- ❑ The Inverse Problem may be very tricky
 - Tb are processed to estimate atmospheric variables
- ❑ Examples of inverse methods:
 - Multivariate regression
 - Neural-network methods
 - Optimal estimation method

Provided by proprietary software
of commercial MWR

More research-oriented

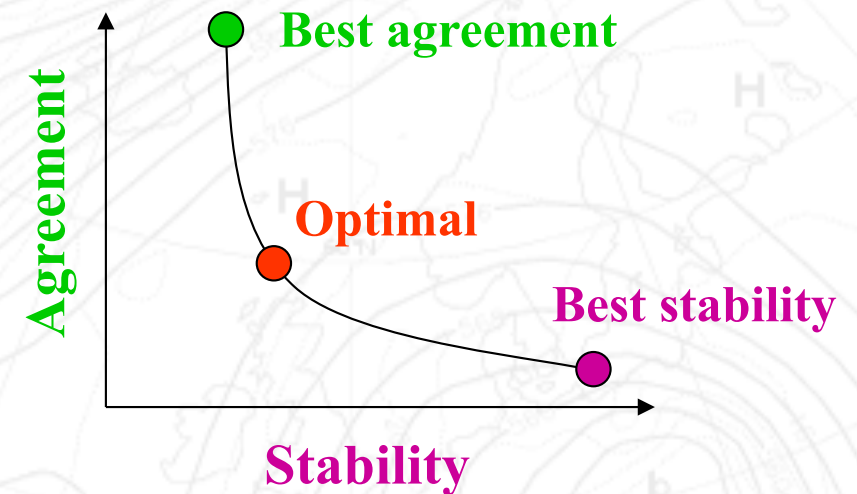


Introduction

Forward and inverse problems

Optimal Estimation Method

- ❑ Method to optimally couple observations with a priori background knowledge (considering the uncertainty of both)



- ❑ Assumptions:
 - Moderately non-linear problem
 - Unbiased and Gaussian-distributed errors

Introduction

Forward and inverse problems

Minimization of a cost function J

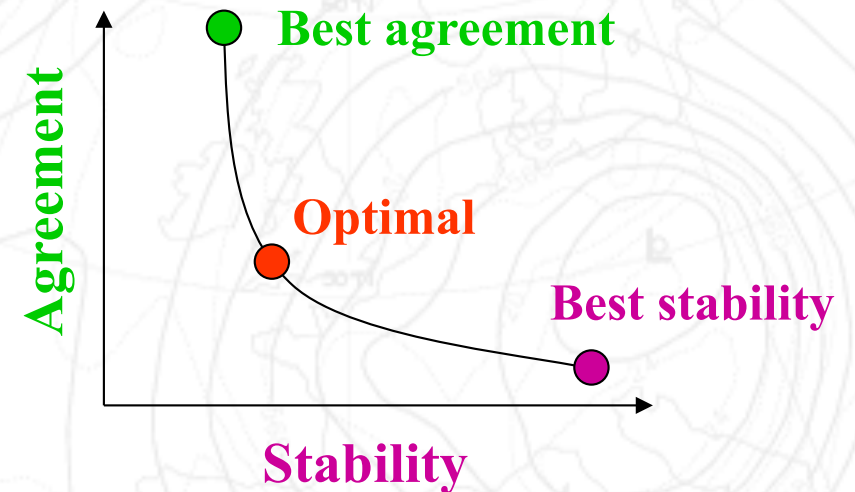
$$J = \underbrace{\left[y - F(x) \right]^T R^{-1} \left[y - F(x) \right]}_{\text{Agreement (wrt the Obs)}} + \underbrace{\left[x - x_b \right]^T B^{-1} \left[x - x_b \right]}_{\text{Stability (wrt the Bkg)}}$$

Agreement
(wrt the Obs)

Stability
(wrt the Bkg)

□ Where:

- forward model $F(x)$
- error covariance matrices of Bkg (B) and Obs (R)



Introduction

Forward and inverse problems

Iterative solution

$$x_{i+1} = x_i + \underbrace{\left(B^{-1} + K_i^T R^{-1} K_i \right)^{-1}}_{\text{Solution increment}} \left[\underbrace{K_i^T R^{-1} (y - F(x_i))}_{\text{This pulls the solution towards obs; how strongly depends on } R^{-1}} - \underbrace{B^{-1} (x_i - x_b)}_{\text{This pulls the solution towards bkg; how strongly depends on } B^{-1}} \right]$$

Updated solution

Solution at previous iteration

K^T transforms from y to x space

y : Observation vector (T_B)

F : Forward Model (e.g. RTTOV-gb)

K : Jacobian ($\partial F / \partial x$)

x_b : a priori profile

R : Obs error covariance matrix

B : Bkg error covariance matrix

A : Solution error covariance matrix

The solution increment is weighted with the sum of obs and bkg uncertainty

$$A_{i+1} = \left(B^{-1} + K_i^T R^{-1} K_i \right)^{-1}$$

If B and R are estimated correctly, the diagonal of A gives the expected random error at each height

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Software tools

RTTOV-gb

Forward model (FM)

- ❑ There are several forward model available for ground-based MWR (ARTS, MonoRTM, PAMTRA,...)
- ❑ But none was specifically designed for NWP Data Assimilation
 - Fast parameterized radiative transfer
- ❑ RTTOV is a fast forward model developed at EUMETSAT NWP SAF
 - Widely used in the NWP community for DA of satellite observations
 - But limited to satellite observing geometry
- ❑ RTTOV was modified within TOPROF to work for ground-based observations, hence RTTOV-gb
 - RTTOV-gb computes the forward + Jacobians (K) wrt to x
- ❑ RTTOV-gb currently only works for MW observations

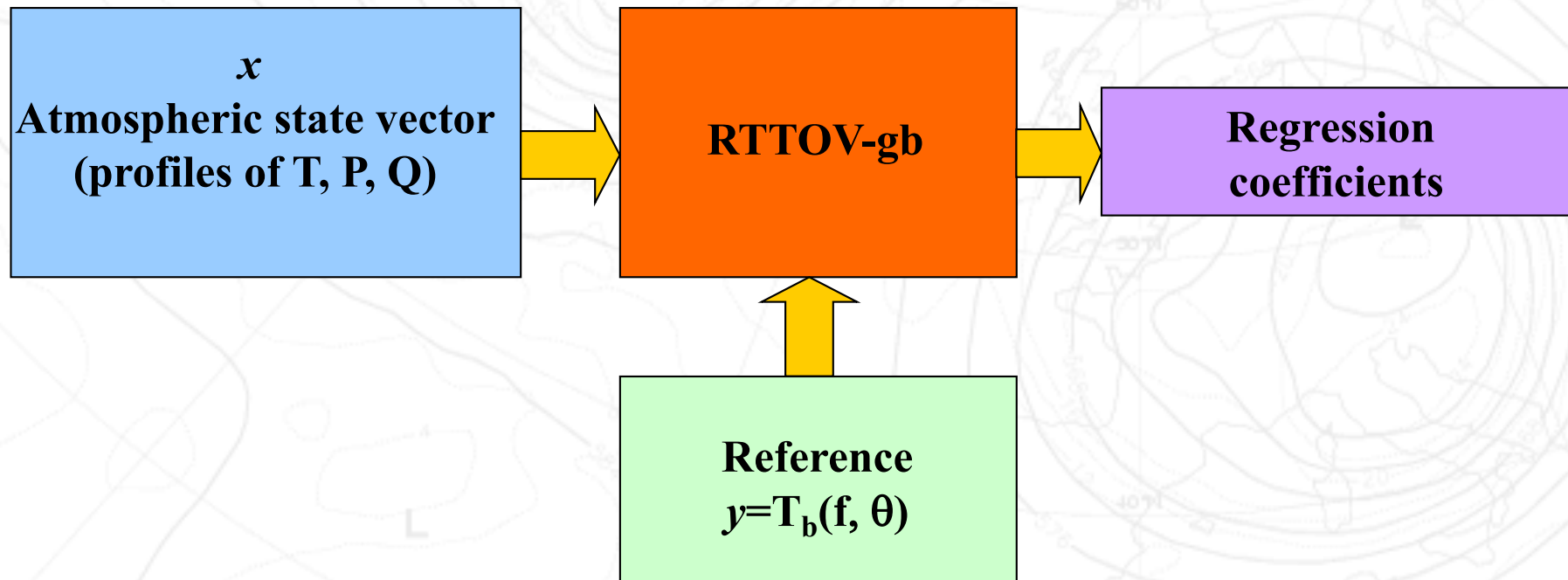
Software tools

RTTOV-gb

How RTTOV-gb works? (similarly to the parent RTTOV)

□ TRAINING:

- Input:
 - Climatological P, T, H profiles on N vertical levels;
 - Atmospheric transmittance calculated with a reference FM
- Output: **Regression coefficients**



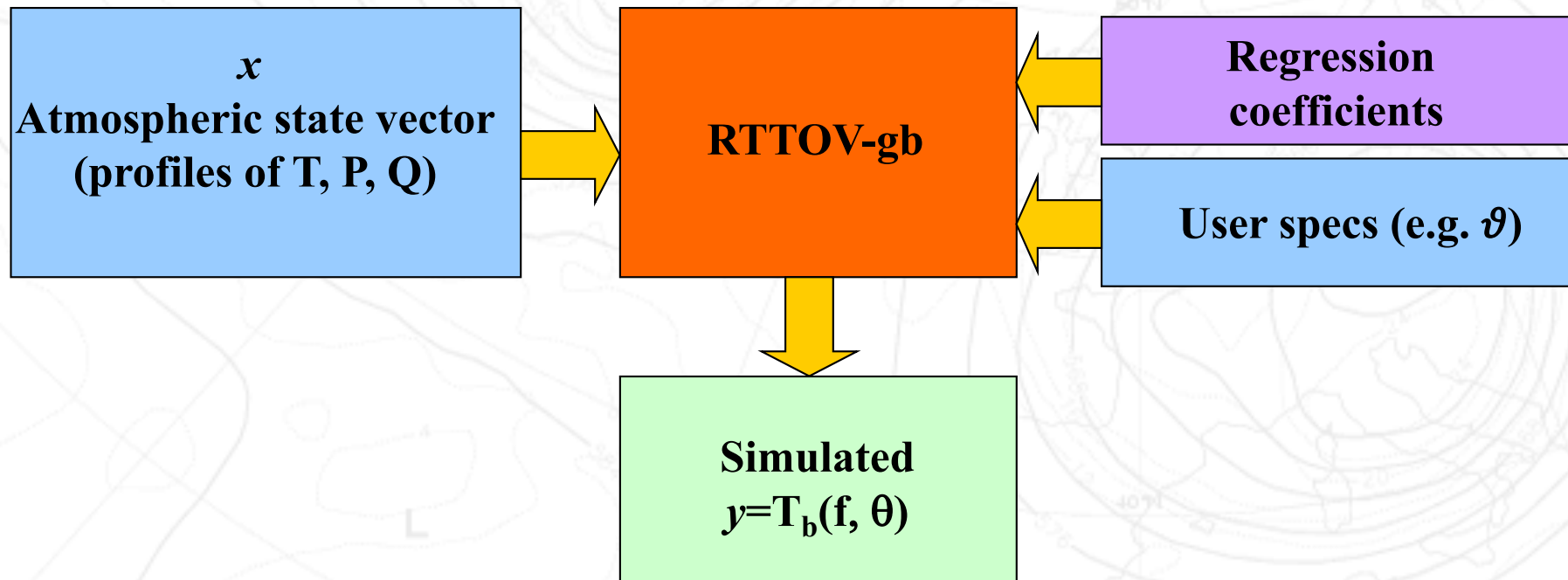
Software tools

RTTOV-gb

How RTTOV-gb works? (similarly to the parent RTTOV)

□ FORWARD COMPUTING:

- Input:
 - P, T, H profiles on N vertical levels;
 - Regression coefficients;
 - User's specs (elevation angle, instrument)
- Output: **Simulated Tb**



Software tools

RTTOV-gb

Statistical evaluation against reference model

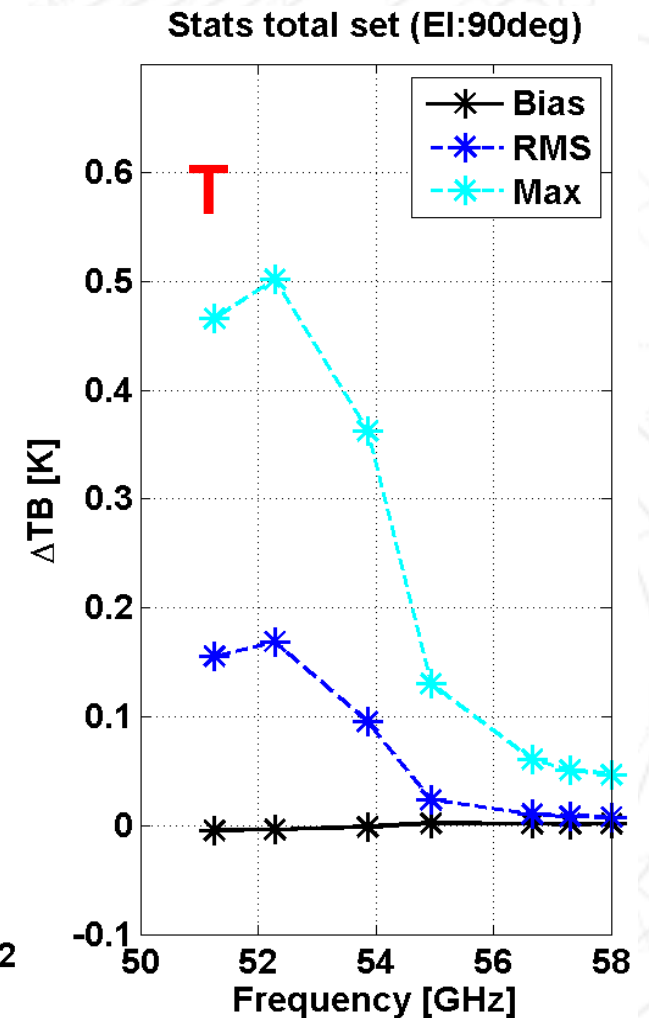
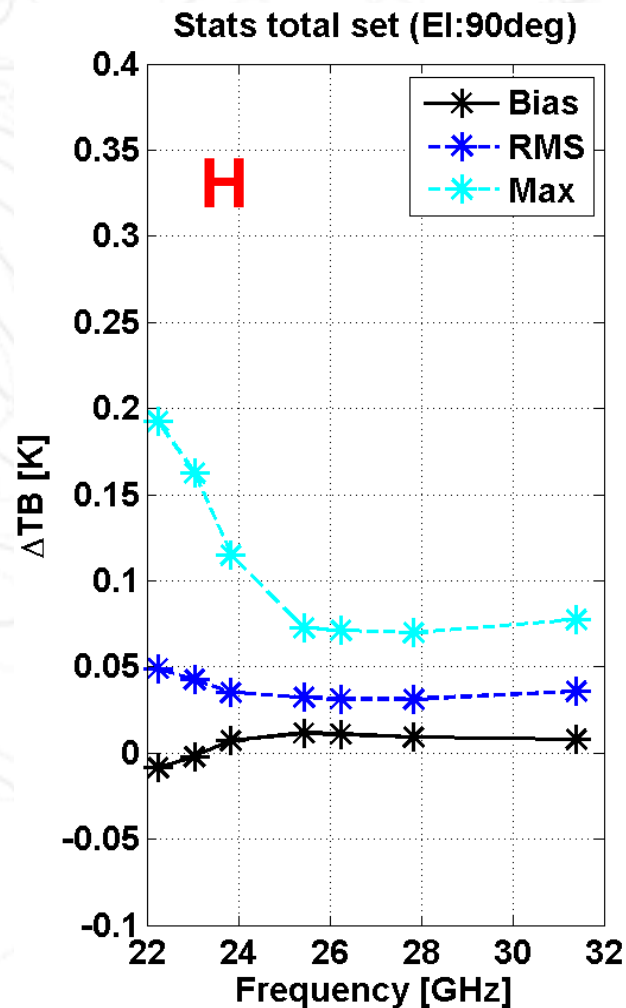
Specs:

Training: LBL Rosenkranz 1998 (R98) model

independent profiles

HATPRO Channels

90° elev. angle



Software tools

RTTOV-gb

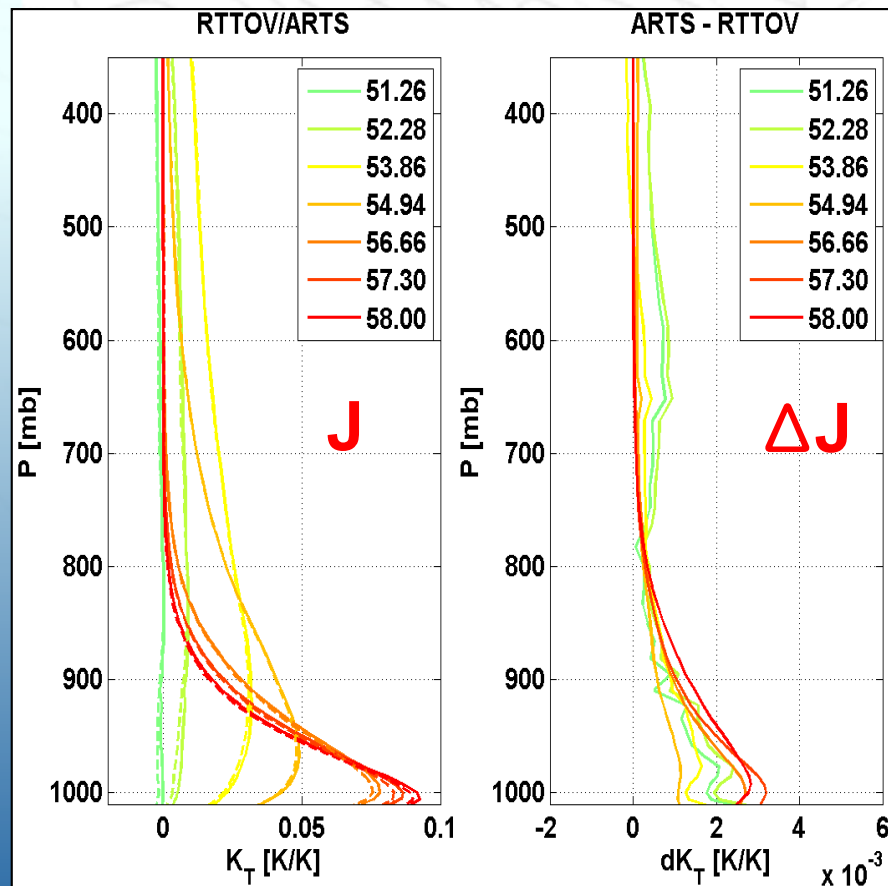
RTTOV and RTTOV-gb also compute the Jacobian $K = \partial F / \partial x$

Jacobian Evaluation (with respect to reference model ARTS)

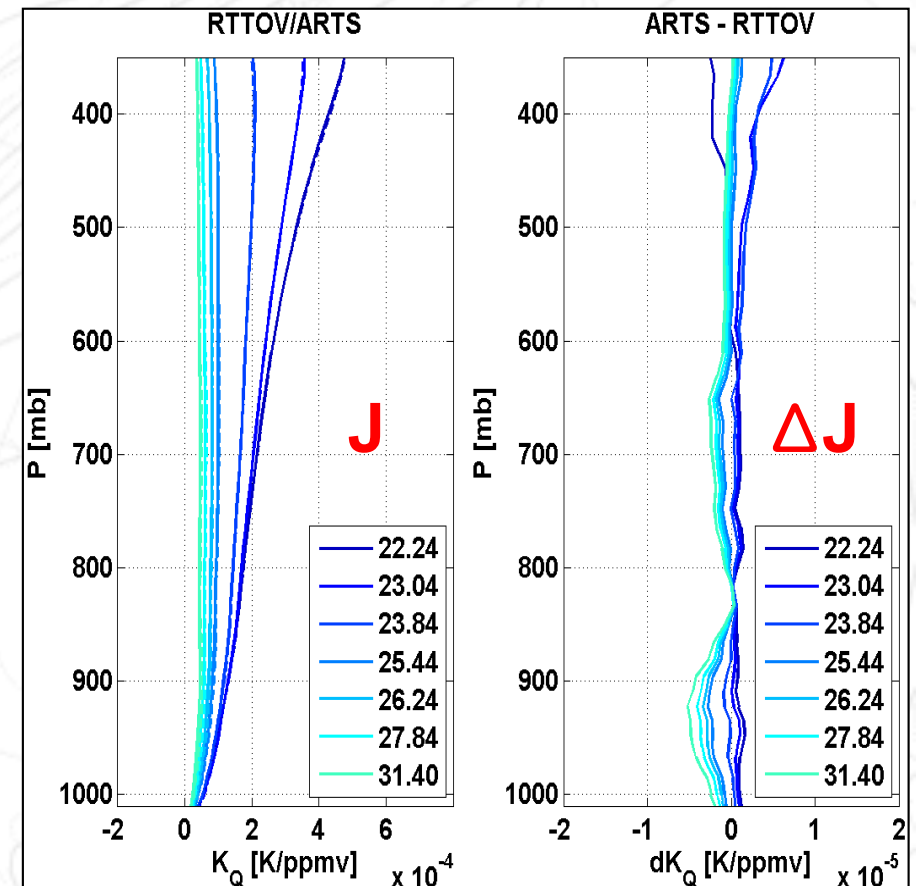
T Jacobian differences < 3%

Q Jacobian differences < 5%

TEMPERATURE



HUMIDITY



Software tools

RTTOV-gb

RTTOV-gb testing

- ❑ Comparison wrt reference model*
 - T_B differences less than typical MWR uncertainties
 - Jacobians nearly identical to reference model
 - RTTOV-gb faster than reference model
- ❑ RTTOV-gb well suited for serving as FM for the direct DA of MWR T_b observations.
- ❑ RTTOV-gb is now in experimental use at some European institutions:
 - University: Geophysics and Meteorology Inst. (Univ. Köln)
 - National Weather Services: Meteo France, DWD



* De Angelis et al., RTTOV-gb, Geosci. Model Develop., 2016

Software tools

RTTOV-gb

What do we need for network NWP Data Assimilation?

- ❑ Test the consistency NWP-FM-Obs → O-B analysis
- ❑ O-B: Observations minus Background
 - Observations are from the instruments
 - Background is from the NWP model
- ❑ O-B analysis is needed for
 - checking the consistency of the Obs with the Bkg
 - estimating the potential innovation in the Obs

Software tools

RTTOV-gb

MWR O-B analysis with RTTOV-gb

- ❑ **Period:** 1 year – Jan-Dec 2014
- ❑ **Format:** NetCDF, CF-compliant, after HD(CP)2
- ❑ **Observations:** 6 MWR (JOYCE, LACROS, CESAR, SIRTAR, PAYERNE, RAO)
- ❑ **Background NWP model:**
 - AROME NWP system (3h-forecasts + Analyses)
- ❑ **Background Tb:**
 - Computed with RTTOV-gb from closest AROME profiles

Software tools

RTTOV-gb

MWR O-B analysis with RTTOV-gb

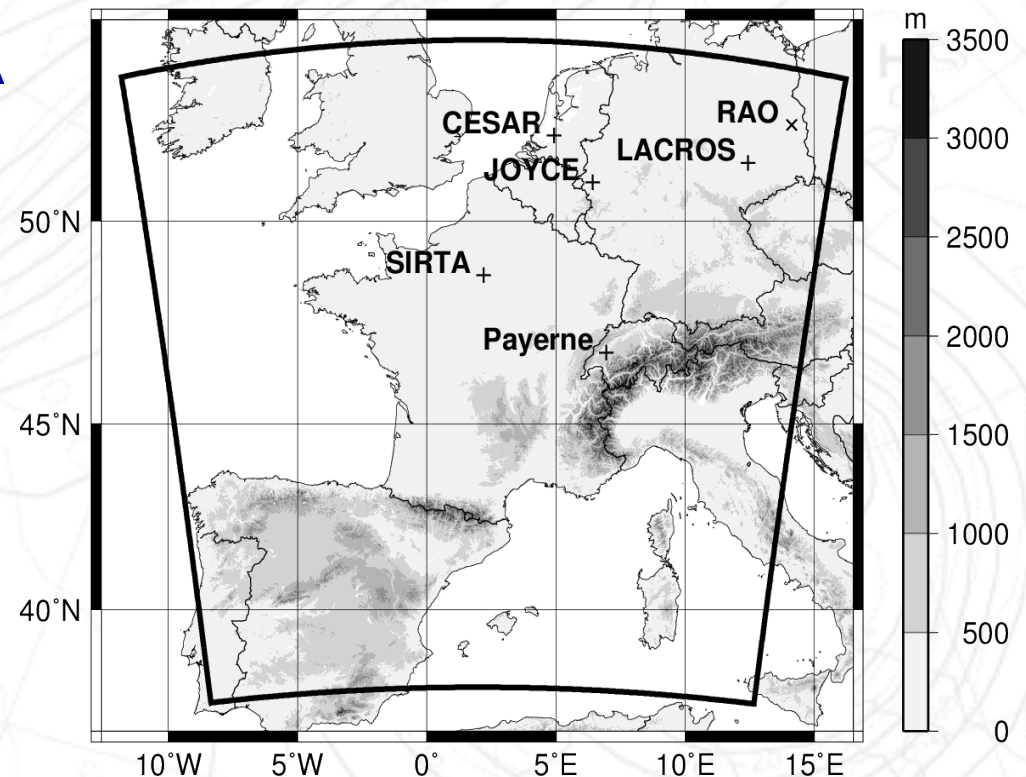
□ 6 stations

○ CESAR	(NL)	HATPRO
○ JOYCE	(DE)	HATPRO
○ LACROS	(DE)	HATPRO
○ Payerne	(CH)	HATPRO
○ RAO	(DE)	MP3000A
○ SIRTÀ	(FR)	HATPRO

HATPRO: 14 chans, 7 angles

MP3000A: 12 chans, 2 angles

□ 1-year data set (2014)

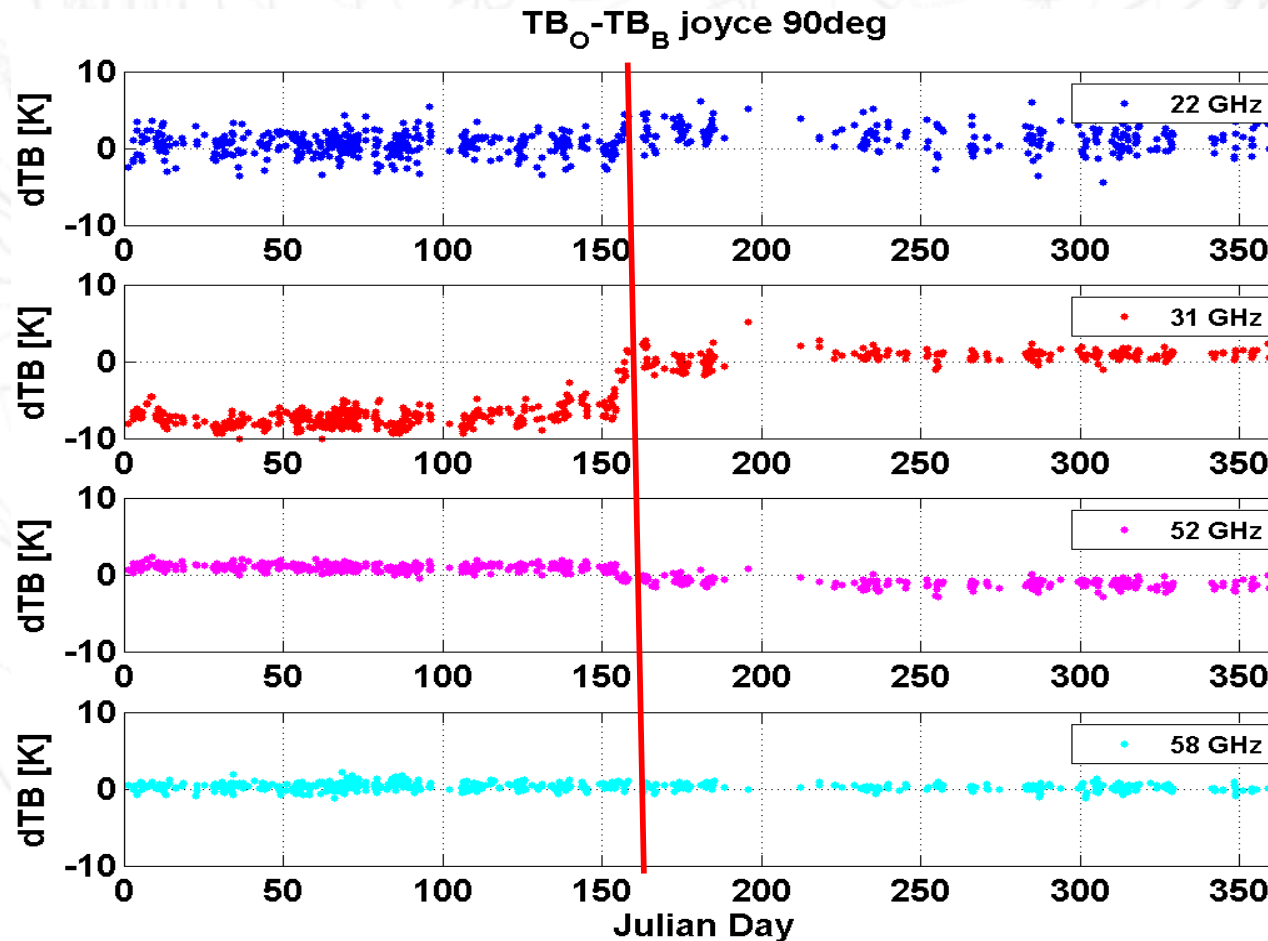


* De Angelis et al., Atm. Meas. Tech. Discuss., 2017

Software tools RTTOV-gb

MWR O-B analysis: bias monitoring

Centralized
remote
monitoring

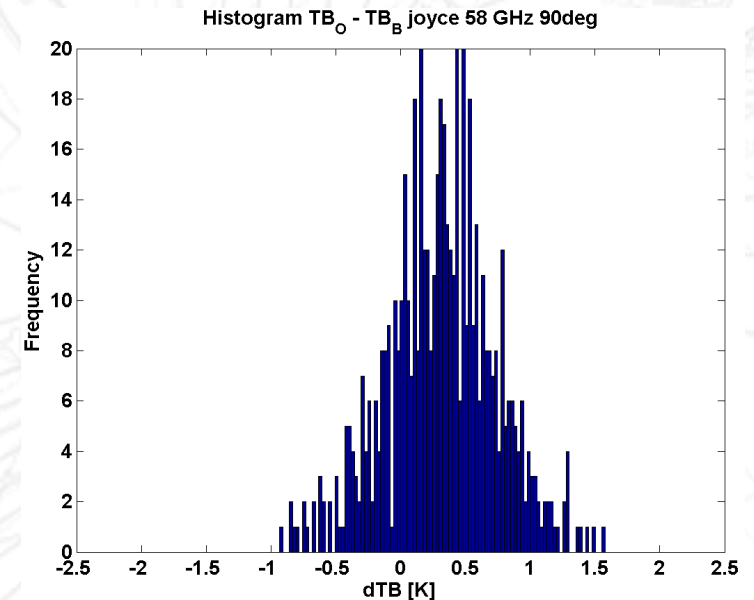
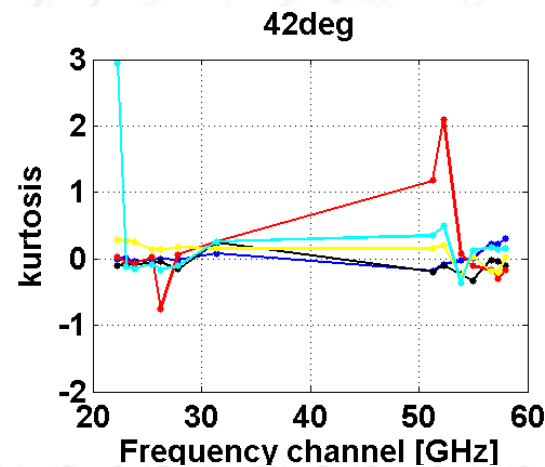
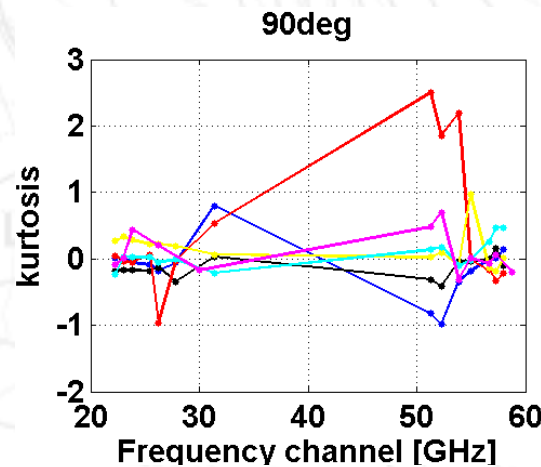


* De Angelis et al., Atm. Meas. Tech. Discuss., 2017

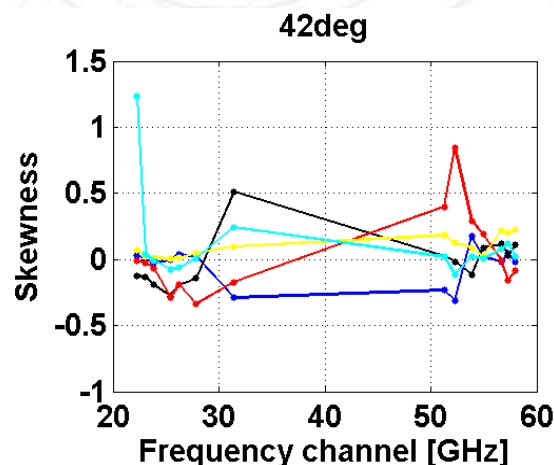
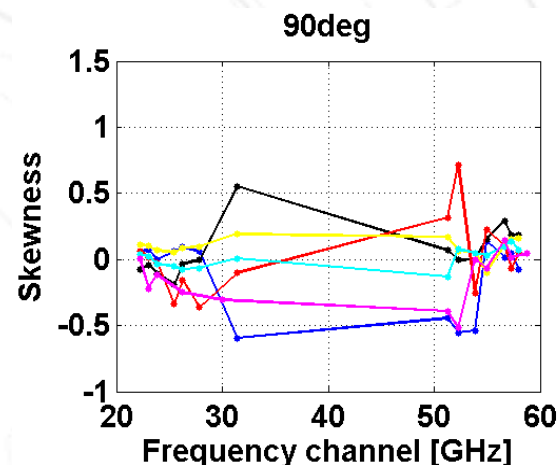
Software tools RTTOV-gb

MWR O-B analysis: distribution monitoring

CESAR
 JOYCE
 LACROS
 Payerne
 RAO
 SIRTa



95.8% excess kurtosis < 1
 99.4% skewness < 0.5



Approximately Gaussian
throughout the network

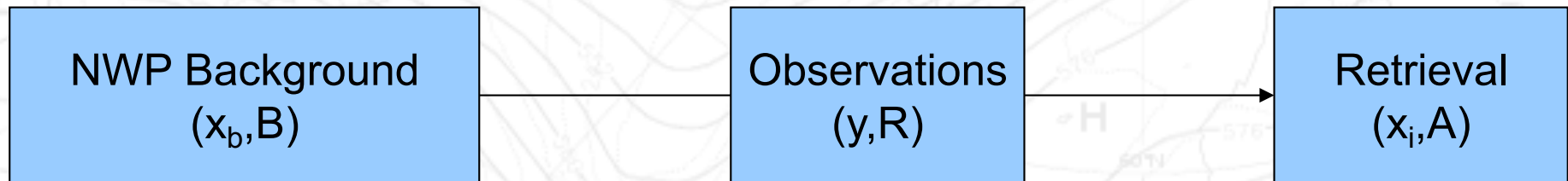
Obs and FM are consistent
with NWP and can then be
assimilated

* De Angelis et al., Atm. Meas. Tech. Discuss., 2017

Software tools

1DVAR

- ❑ Another step towards MWR data assimilation is the development of a one-dimensional variational retrieval (1DVAR)
- ❑ 1DVAR is a nonlinear retrieval technique based on Optimal Estimation Method with the a priori information taken from a NWP model output



Software tools

1DVAR

1DVAR development





- ❑ Start from the EUMETSAT NWP SAF 1DVAR package
 - Stand-alone system calling RTTOV (satellite applications)
- ❑ Modify to work with RTTOV-gb
 - NWPSAF 1DVAR v1.1 interfaced with RTTOV-gb v11.2
 - Modified the “composite instrument” option from “multiple instrument” to “same instrument with multiple elevation angles”
 - Added option to minimize humidity either in logarithm or in specific humidity
 - Extended quality check during the minimization (to avoid negative humidity and liquid water content values)
 - More flexibility in the LWP retrieval (background error defined by the user, minimization starting from the background value or a default value)

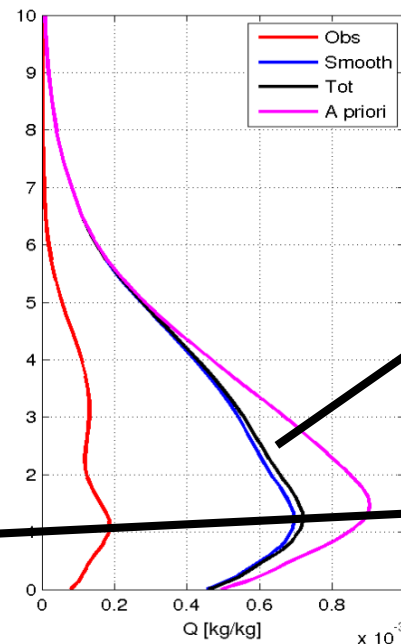
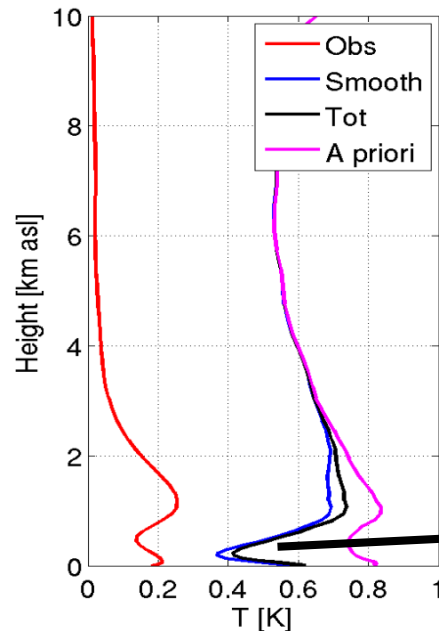
Software tools

1DVAR

1DVAR results

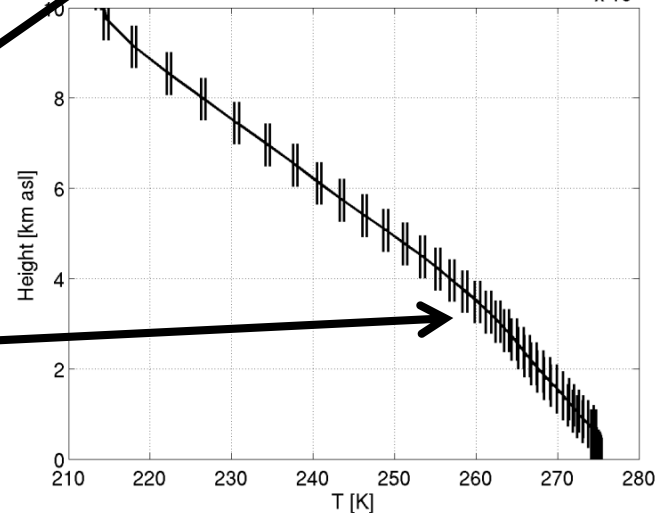
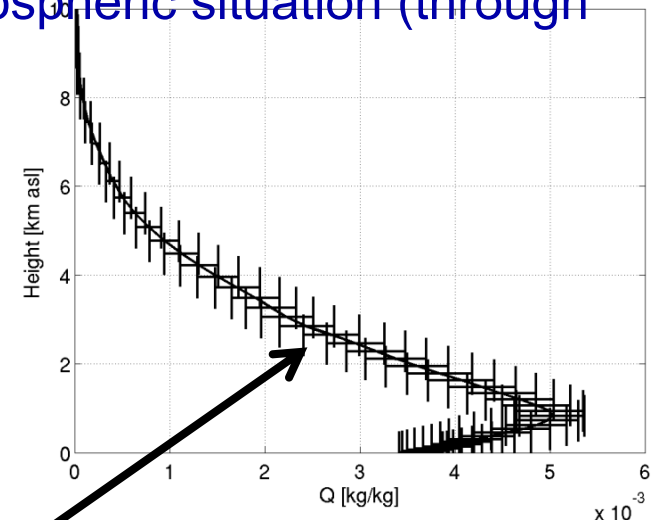
- T and Q profiles with estimated retrieval uncertainty
- Estimated uncertainties depends on the atmospheric situation (through Jacobians)

 Background error
 Total uncertainty
 Smoothing error
 Instrumental error



Error covariance

$$A = \left(B^{-1} + K_i^T R^{-1} K_i \right)^{-1}$$



Software tools

Net1D

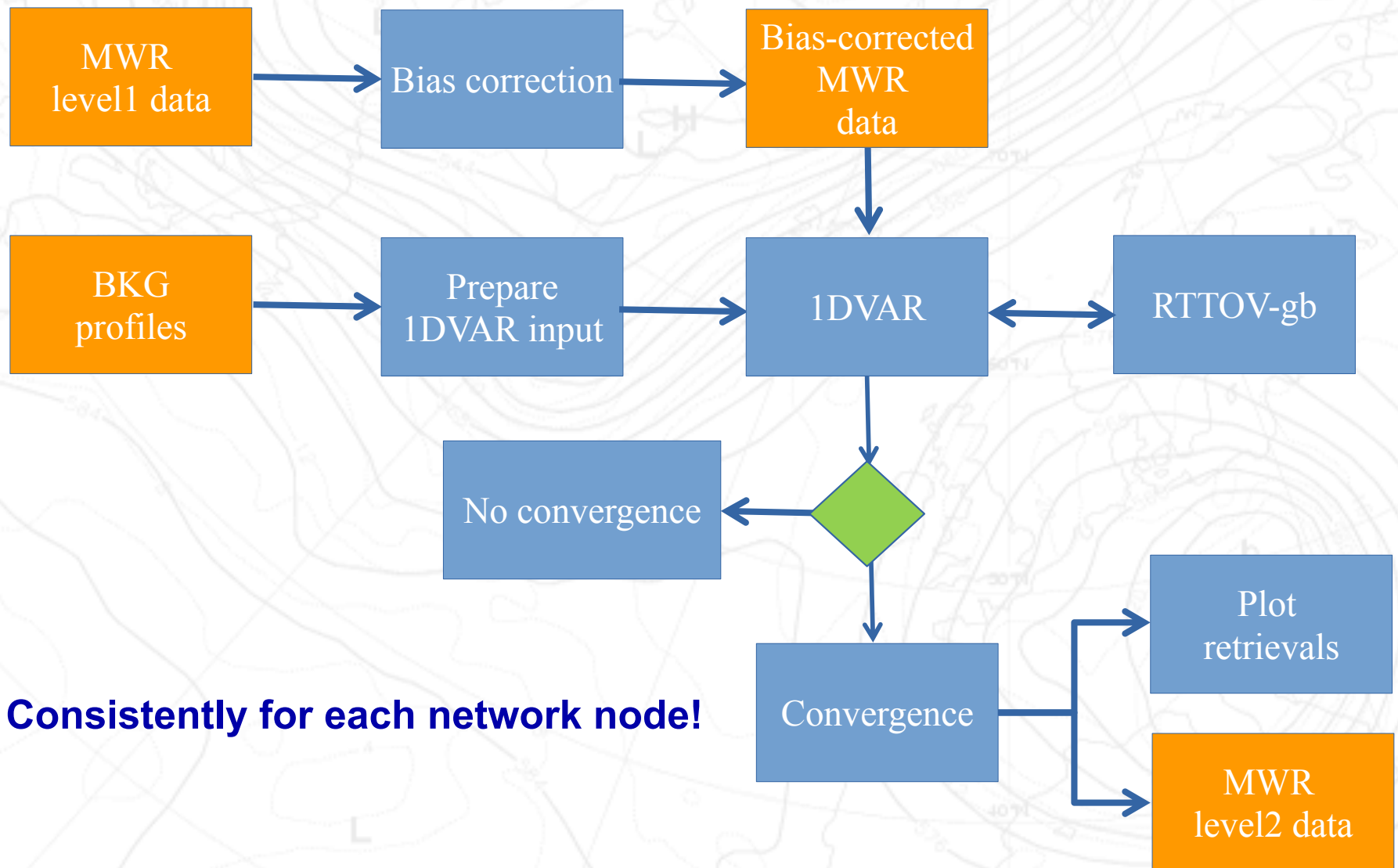
How can we test this for a MWR network?

- ❑ Consider a prototype-network
- ❑ Develop a flexible processing chain to perform 1DVAR retrievals on different sites with different instrument and configurations
- ❑ Evaluate retrieval accuracy and dominant sources of errors (uncertainty / systematic / drifts)

Software tools

Net1D

Net1D processing chain



Consistently for each network node!

Software tools

Net1D

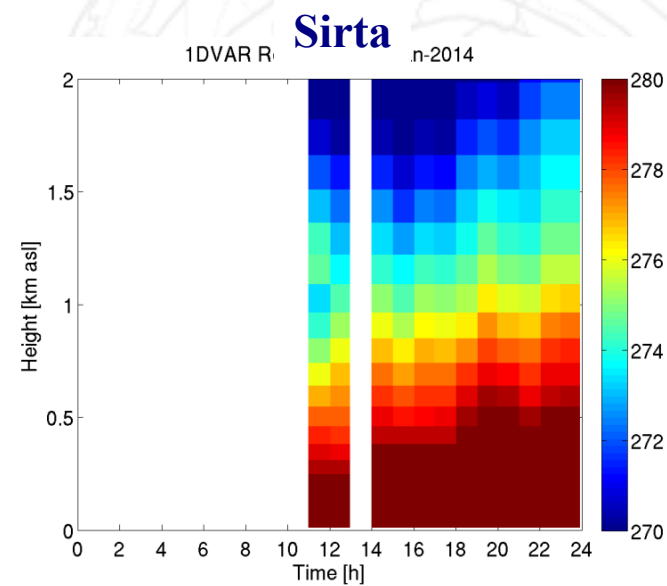
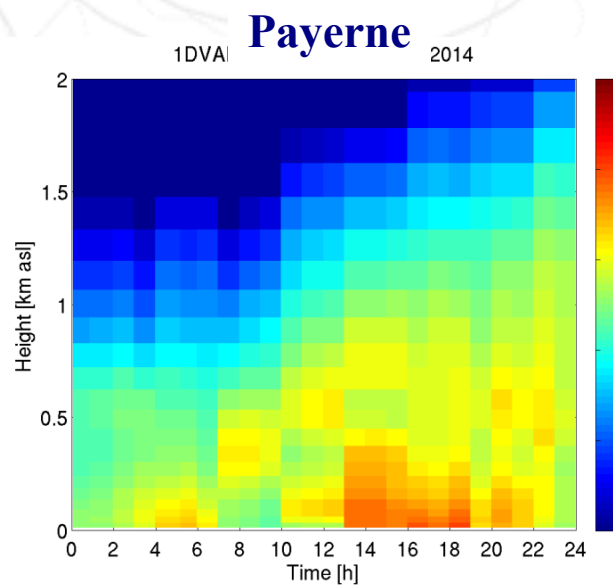
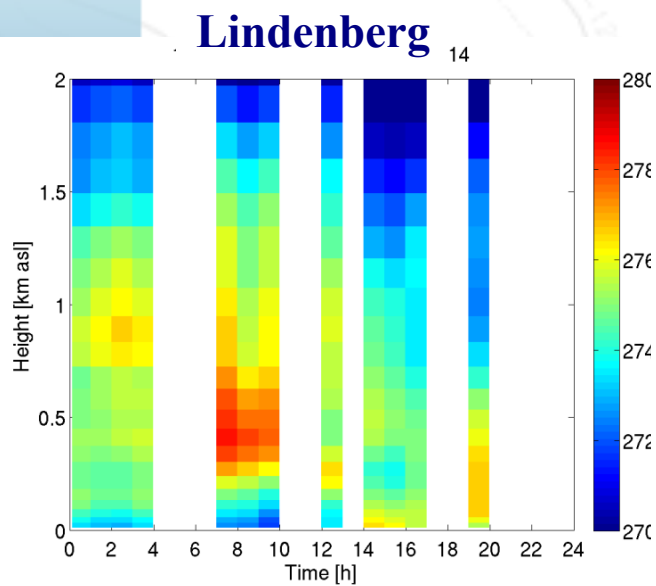
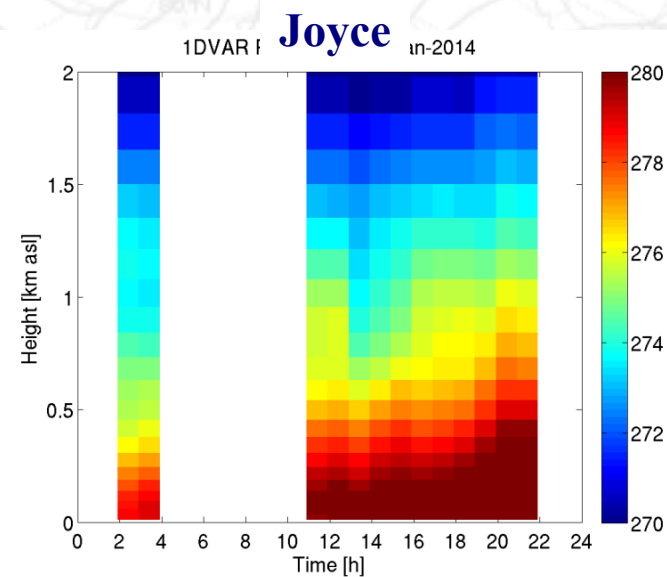
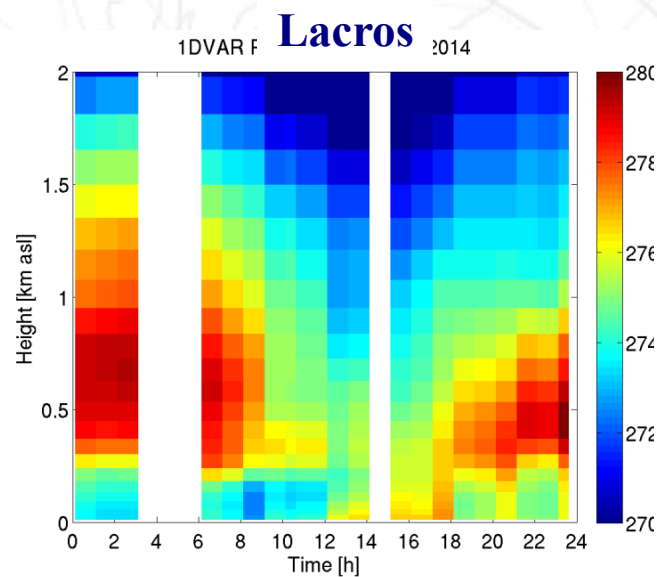
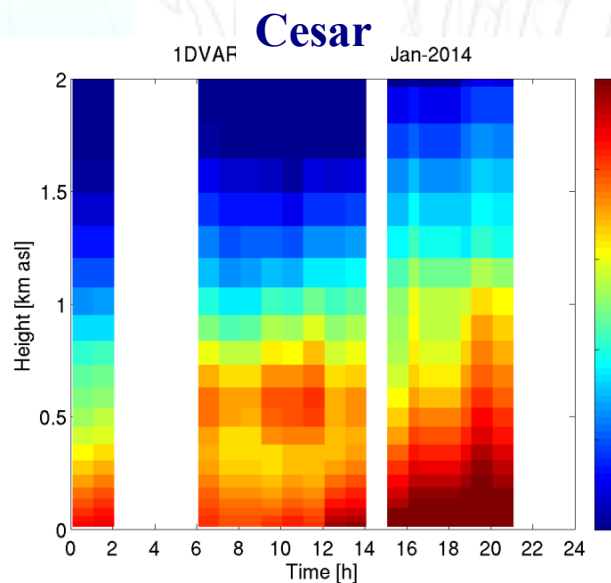
- ❑ The whole configuration is defined in one config file:
 - input data and paths
 - station (location, instrument)
 - period to process
 - sampling strategy
 - channels and elevation angles to be used
 - control variables
 - R and B matrix
 - cloud configuration
 - etc..

- ❑ First evaluation of Net1D chain on one day and six stations

Software tools

Net1D

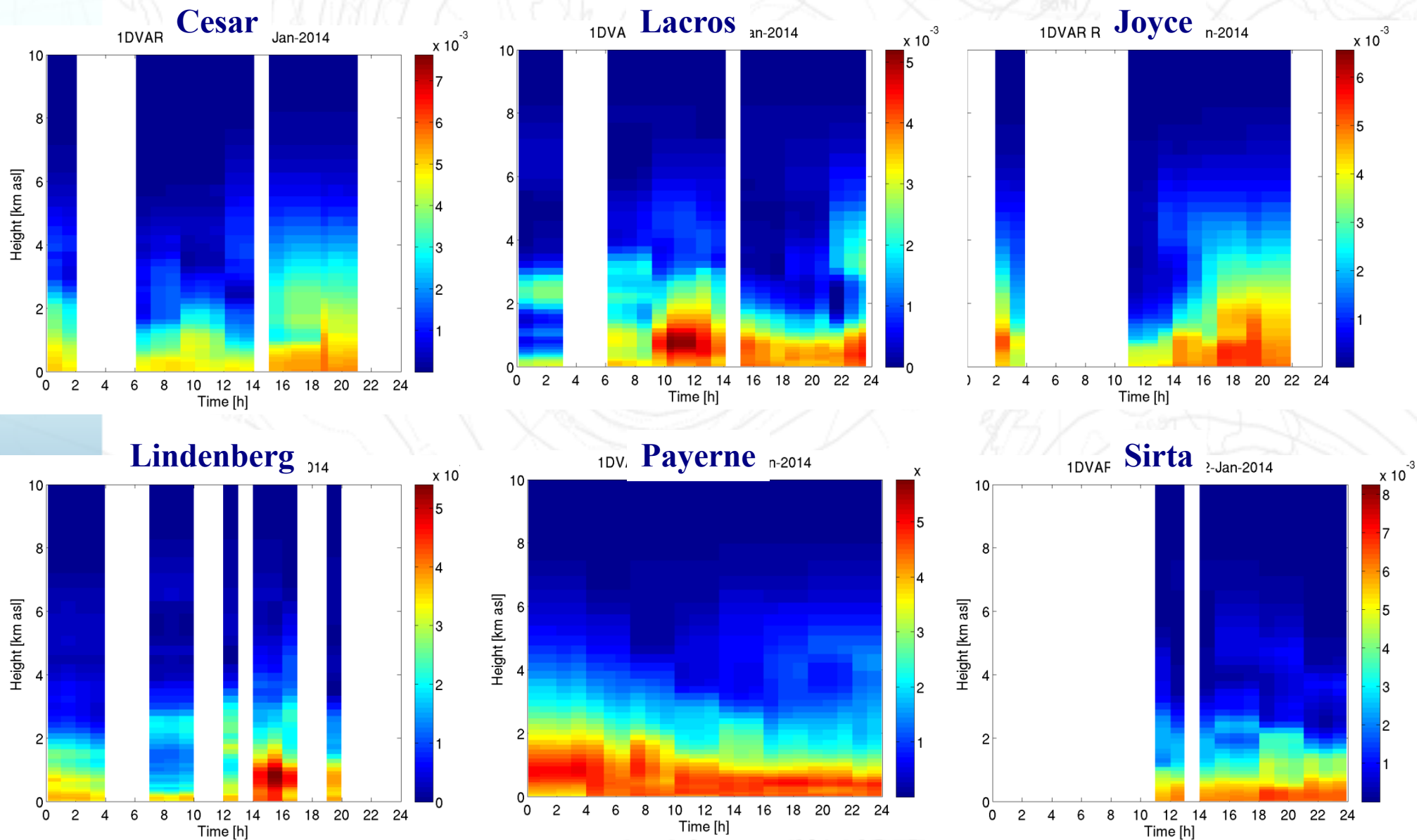
Temperature profile retrievals



Software tools

Net1D

Humidity profile retrievals

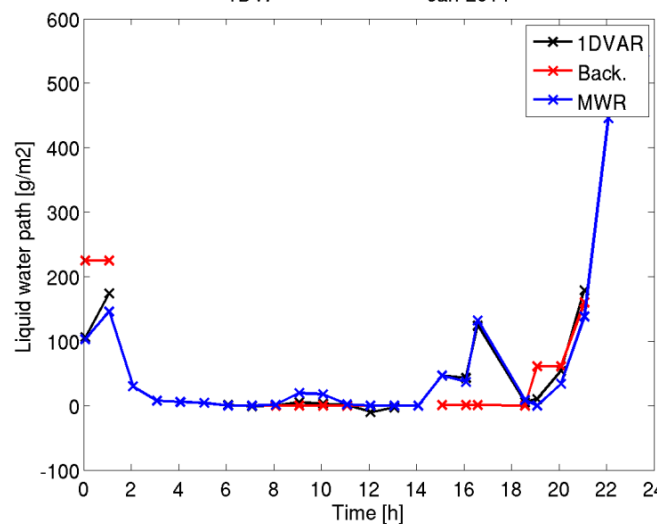


Software tools

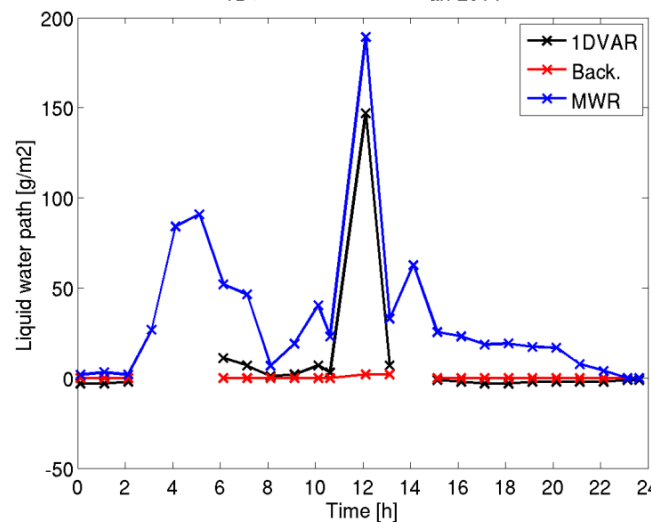
Net1D

Liquid water path retrievals

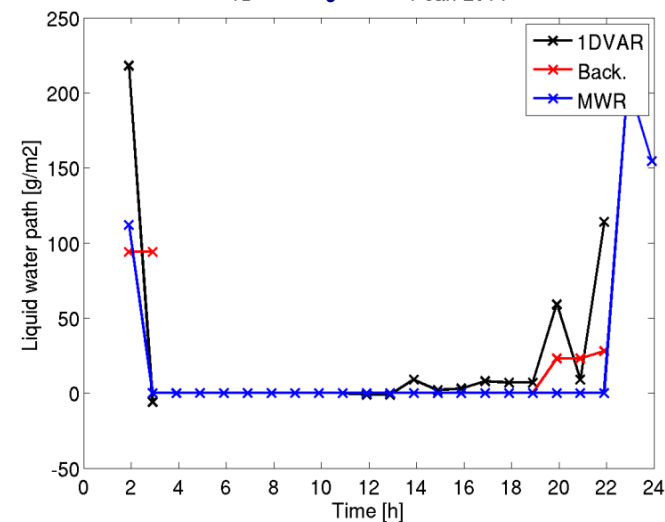
1DV Cesar Jan-2014



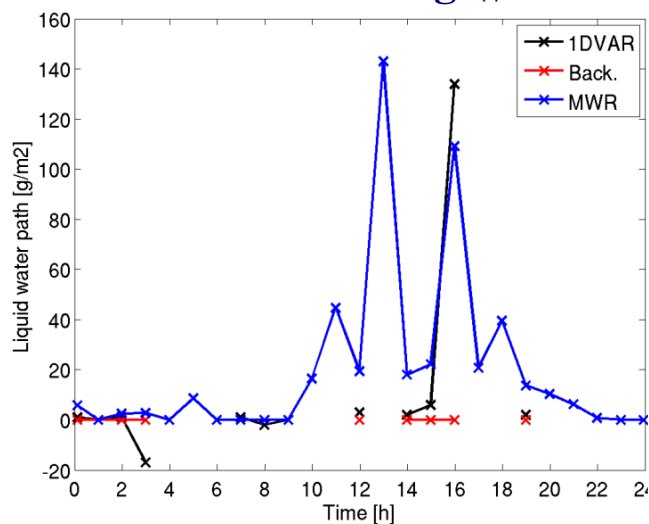
1DV Lacros an-2014



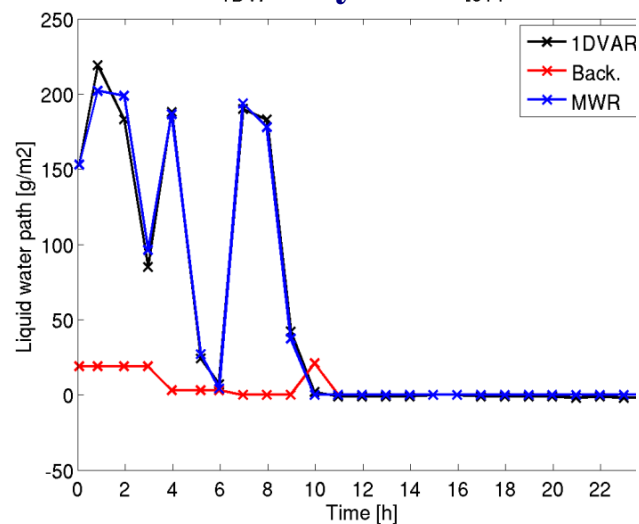
1DV Joyce 1-Jan-2014



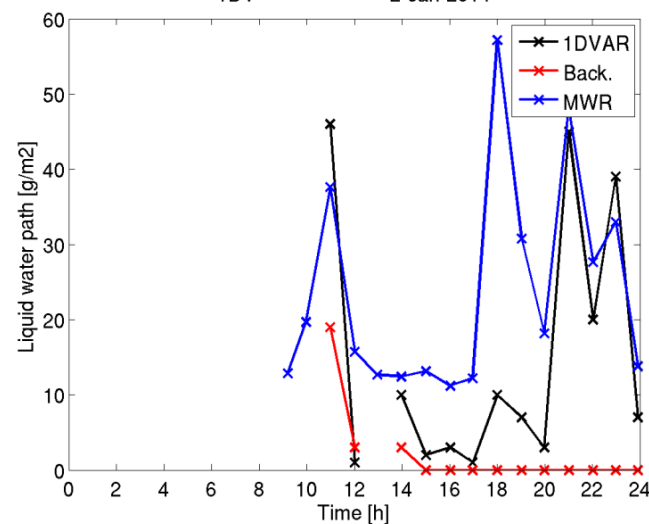
1DV Lindenberg 14



1DV Payerne 2014



1DV Sirta 2-Jan-2014



Software tools

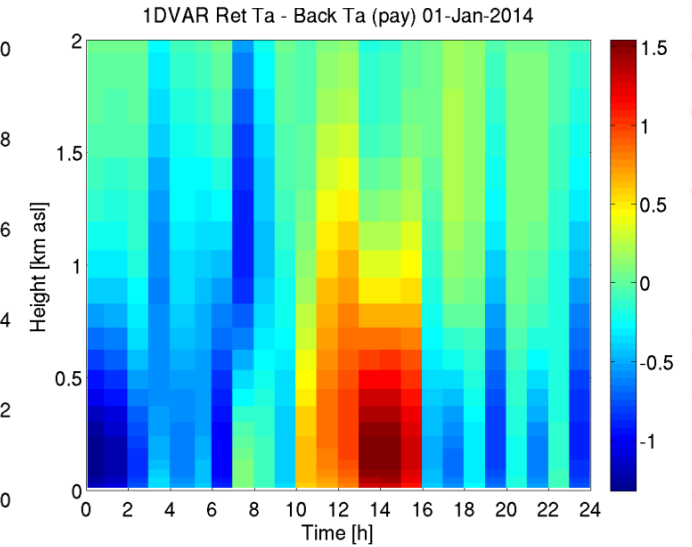
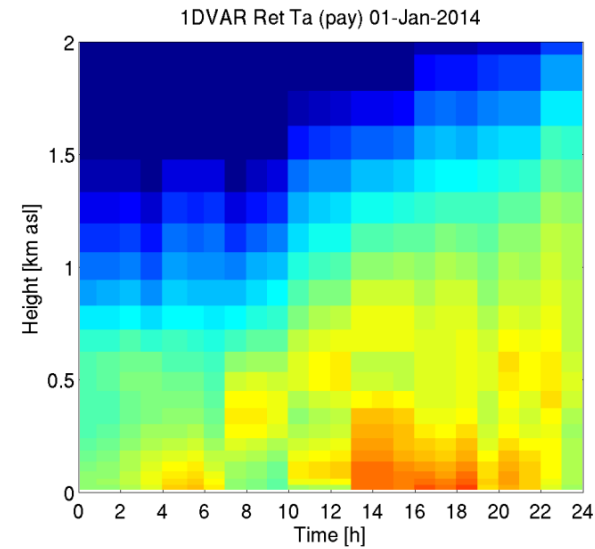
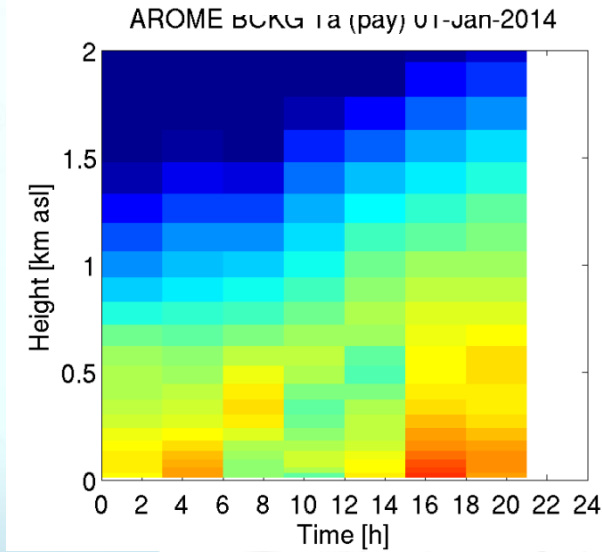
Net1D

Impact on the background

T AROME

T 1DVAR

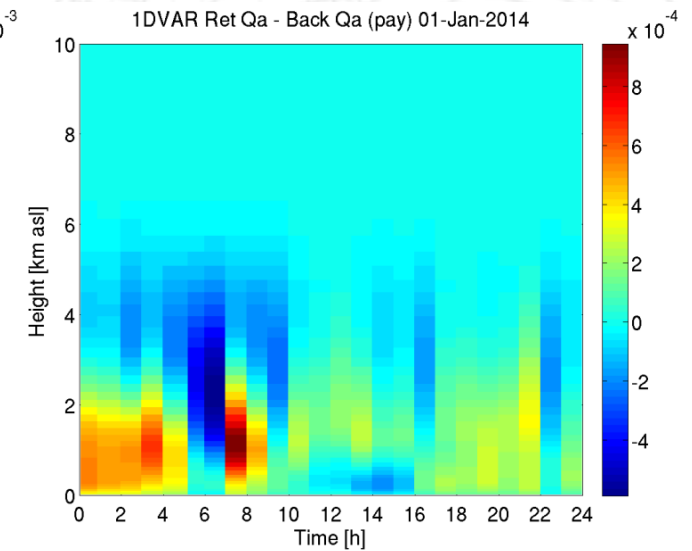
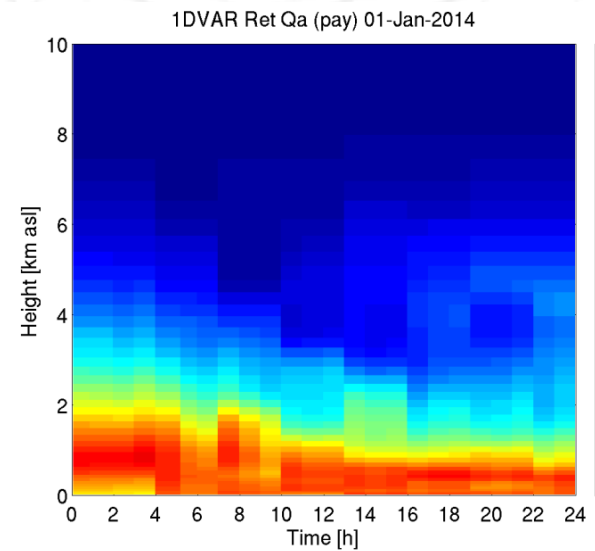
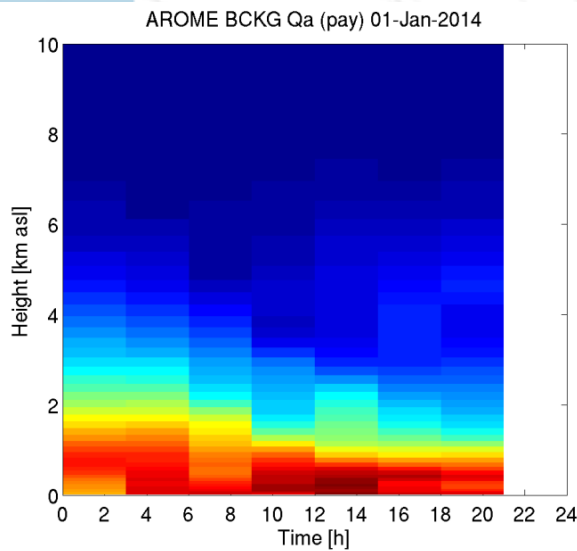
Increment



Q AROME

Q 1DVAR

Increment



Software tools

Net1D

- ❑ Next steps:
 - perform one year of retrievals
 - validation against radiosondes where available

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Summary and practical information

Software tools:

- **RTTOV-gb**
 - fast FM for ground-based MWR
 - Ingests atmospheric profiles
 - Computes Tb and Jacobians
- **1DVAR**
 - fast inversion scheme
 - Ingests ground-based MWR obs
 - Computes retrievals of T and H profiles and LWP
- **Net1D**
 - network 1DVAR retrievals
 - Ingests ground-based MWR obs from a network (curr. 6 nodes)
 - Computes 1DVAR retrievals consistent throughout the network

Summary and practical information

Availability:

- **RTTOV-gb**
 - will be available through NWP SAF
 - needs a free RTTOV licence
- **1DVAR**
 - available through P. Martinet (Météo France)
 - needs RTTOV-gb
- **Net1D**
 - currently available on GitHub (Matlab code)
 - plans for translating to python
 - needs RTTOV-gb and 1DVAR

**Thanks much for
your attention!**

