

# Data Assimilation of Cloud-Affected Radiances

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# Outline

- Motivation
- Methodology
- Results
  - Observation operators
  - Assimilation and verification
- Summary and conclusions

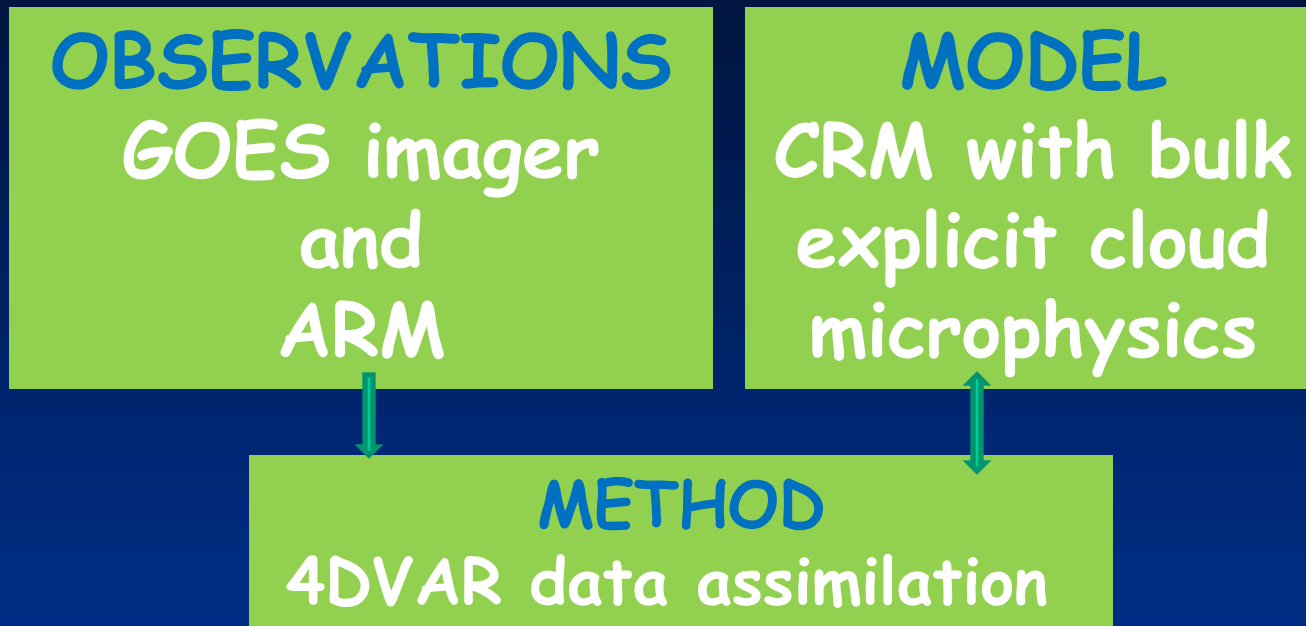
# Motivation

## 4D atmospheric data analysis with clouds

- Initialization of atmospheric state with clouds in NWP
- Validation and advanced development of cloud microphysical parameterizations
- Dynamically consistent cloud and state climatology

# Research approach: Evaluate feasibility under best scenario

- Use observations with highest expected information content with respect to clouds, including spatial and temporal variability
- Use cloud resolving model
- Use data assimilation method which allows 4D dynamically consistent analysis

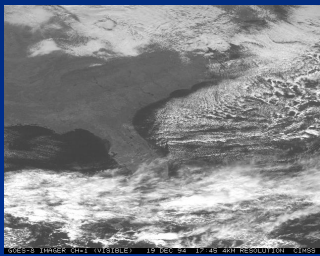


Collaborators : Tom Greenwald (CIMMS, formerly CIRA), Milija Zupanski (CIRA), Dusanka Zupanski (CIRA), Manajit Sengupta (formerly CIRA), Frank Evans (ATOC/CU) and Rosanne Polkinhorne (ATOC/CU)

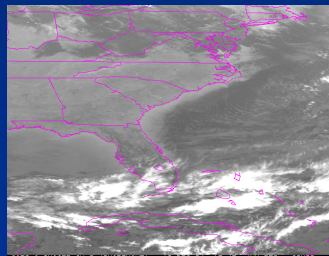
# GOES imager

GOES Channel	Wavelength (μm)	Central Wavelength (μm)	Detector Resolution (km)
1	0.52-0.72	0.7	1
2	3.78-4.03	3.9	4
3	6.47-7.02	6.7	8
3 G12	5.77-7.33	6.5	4
4	10.2-11.2	10.7	4
5	11.5-12.5	12.0	4
6 G12	12.9-13.7	13.3	8

15 minute data

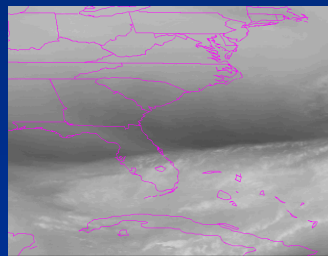


VIS



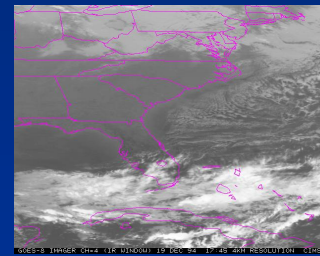
Near IR

Diff between ice and water clouds

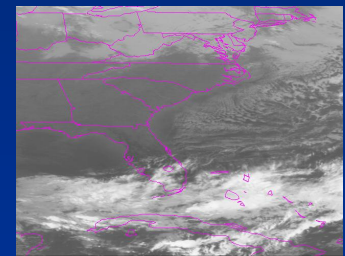


IR

upper water vapor



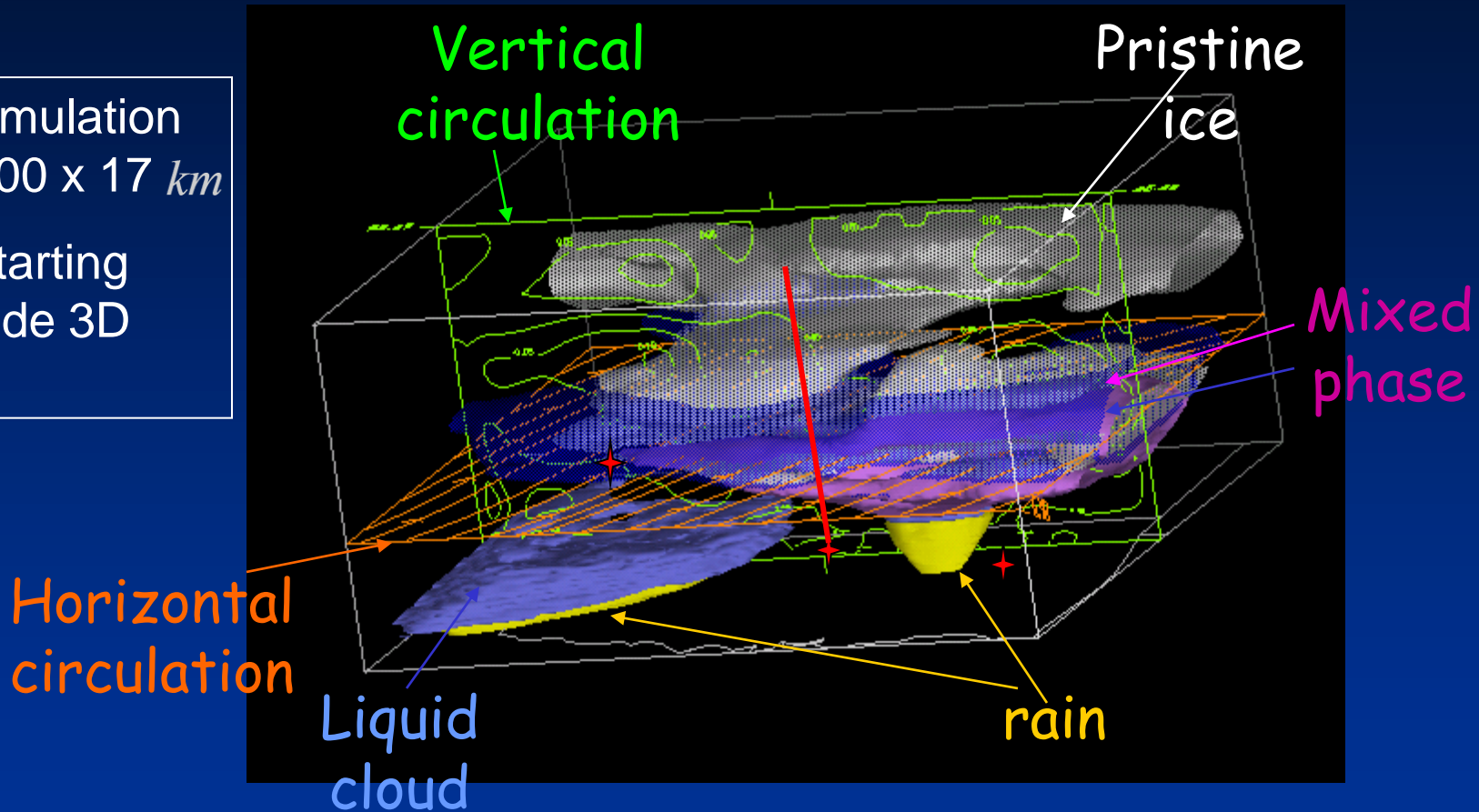
IR clouds and surface



IR clouds, surface and low level vapor

# Cloud resolving model representation of cloudy atmosphere

CRTM simulation  
in  $300 \times 300 \times 17 \text{ km}$   
domain starting  
from a crude 3D  
analysis



CRTMs have skill

# CRTM in this study

## RAMS

- Bulk, 2 moment cloud microphysics for ice: pristine ice, aggregates, snow, graupel and hail
- 1 moment for liquid: cloud droplets and rain
- Prognostic mixing ratio and number concentration for ice
- Assumed Gamma distribution with prescribed width
- Nonhydrostatic dynamics
- High resolution regional simulations



# 4DVAR algorithm

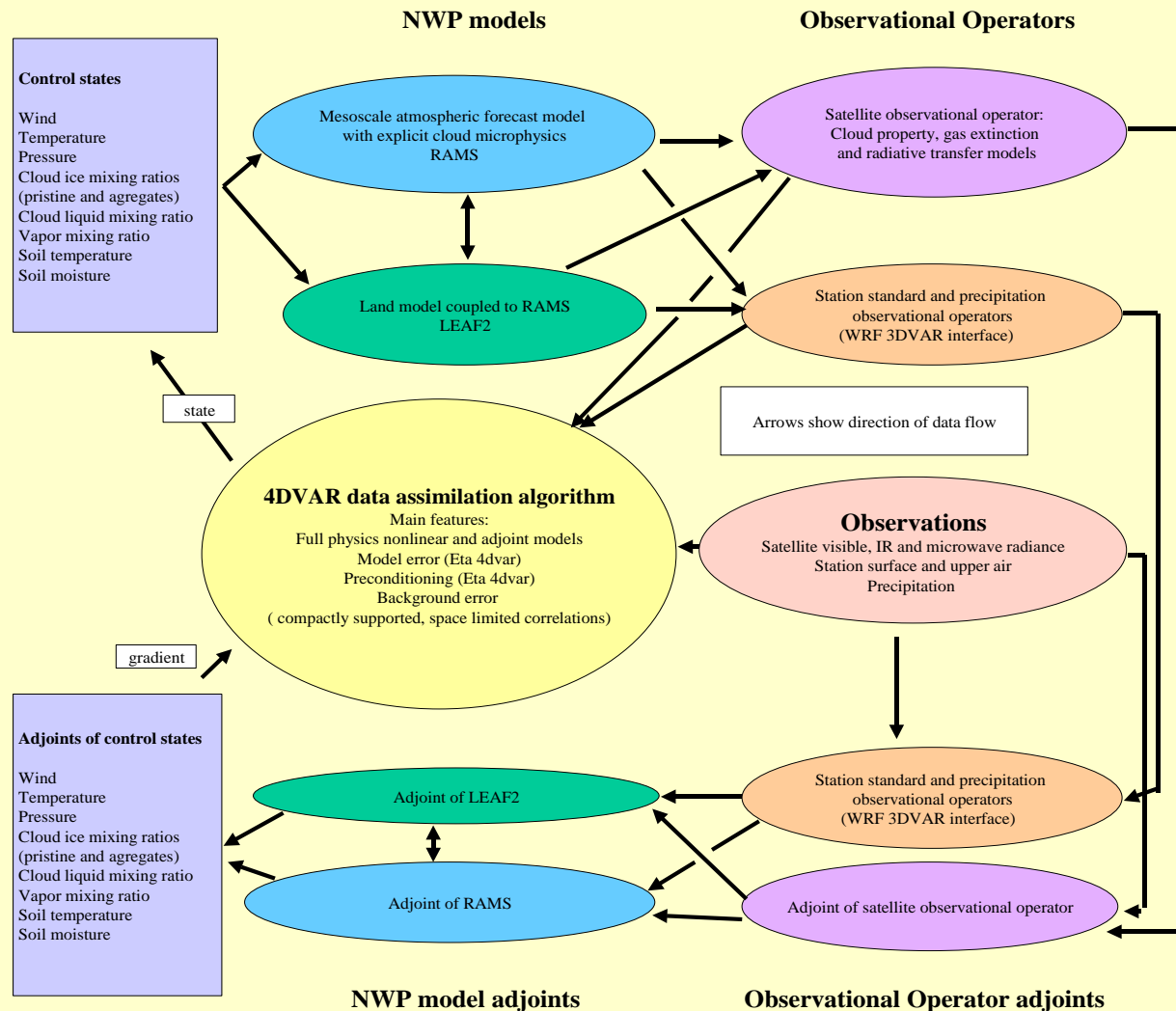
## Regional Atmospheric Modeling and Data Assimilation System

Controls include cloud  
and dynamical variables

(RAMDAS)

Vukicevic et al, 2004,2005

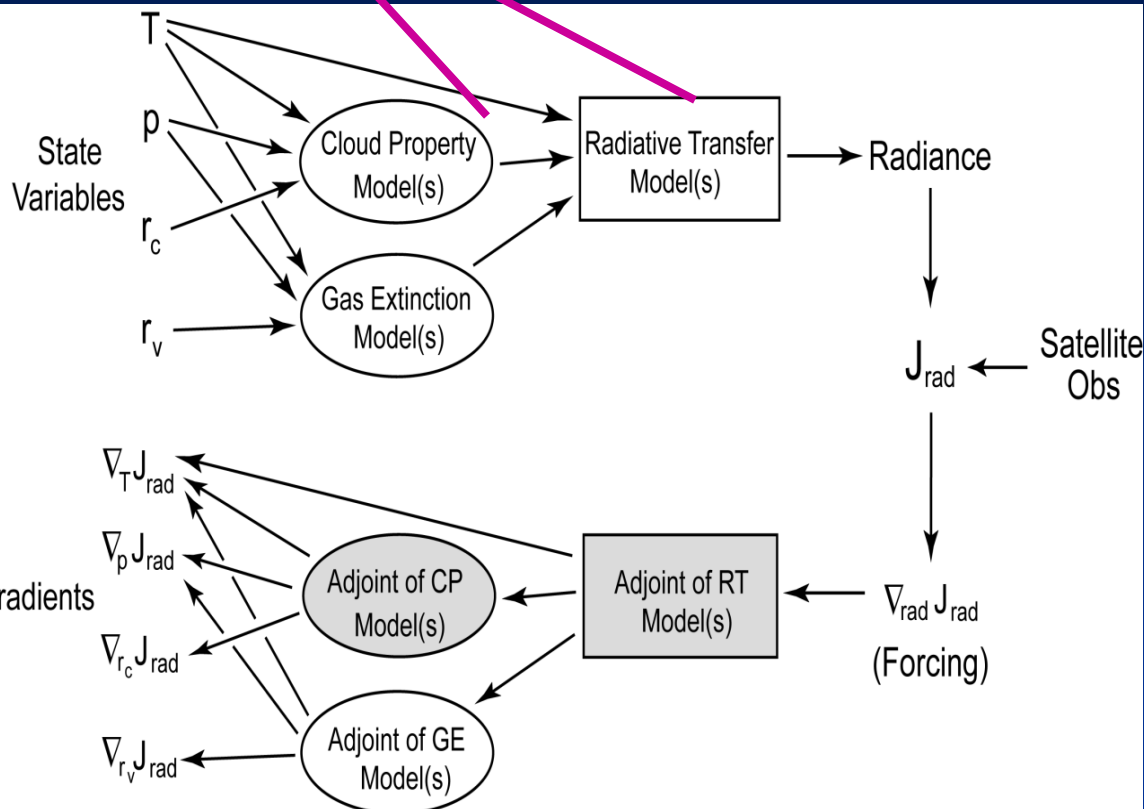
Zupanski et al 2005



Adjoint of cloud microphysics and radiative transfer

# Observation operator VIS and IR radiative transfer

$$y = H(X^t) + \varepsilon_y$$



## Version 1

Greenwald et al. 2003

Gas absorption: OPTRAN  
(McMillin et al., 1995)

Cloud properties: Anomalous  
Diffraction

Solar: SHDOM (Evans, 1998)

IR: Eddington two-stream  
(Deeter and Evans 1998)

## Version 2

SHDOMPPDA (Evans, 2007)

# SHDOMPPDA operator

## Evans (2007, JAS)

- Development was supported by JCSDA and NSF-ATM
- Unpolarized, plane parallel RT model with adjoint and tangent linear models
- Hydrometeor optical properties are determined from lookup tables as function of mass mean radius
- Scattering by look-up tables
  - Mie theory for spherical particles with Gamma or Lognormal size distributions
  - Gamma size distribution of mixture of 6 ice crystal shapes (Yang et al., 2005)

# Information content of GOES imager observations

# VIS and IR information content analysis

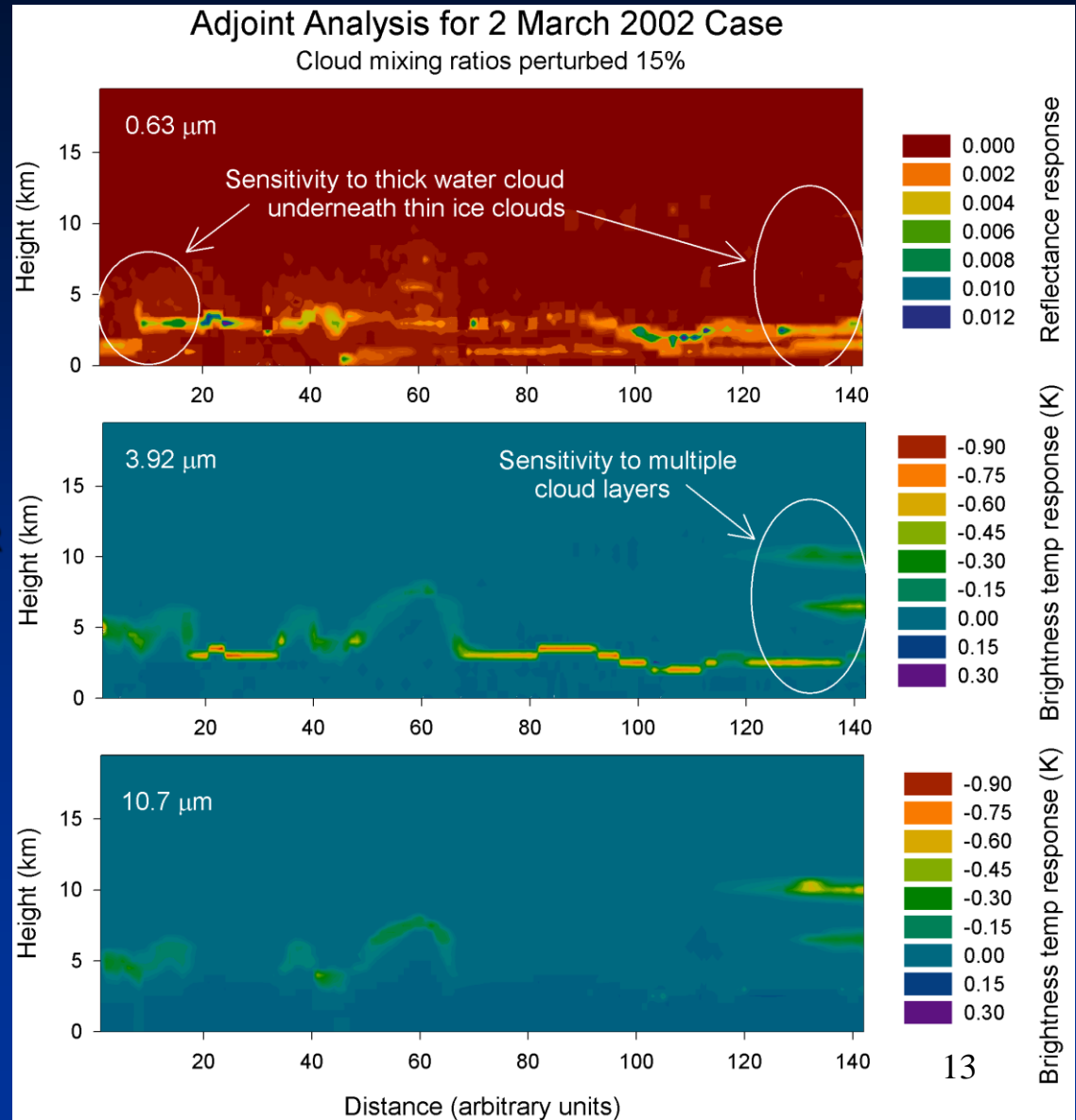
## Example of a case with mixed phase clouds

Visible

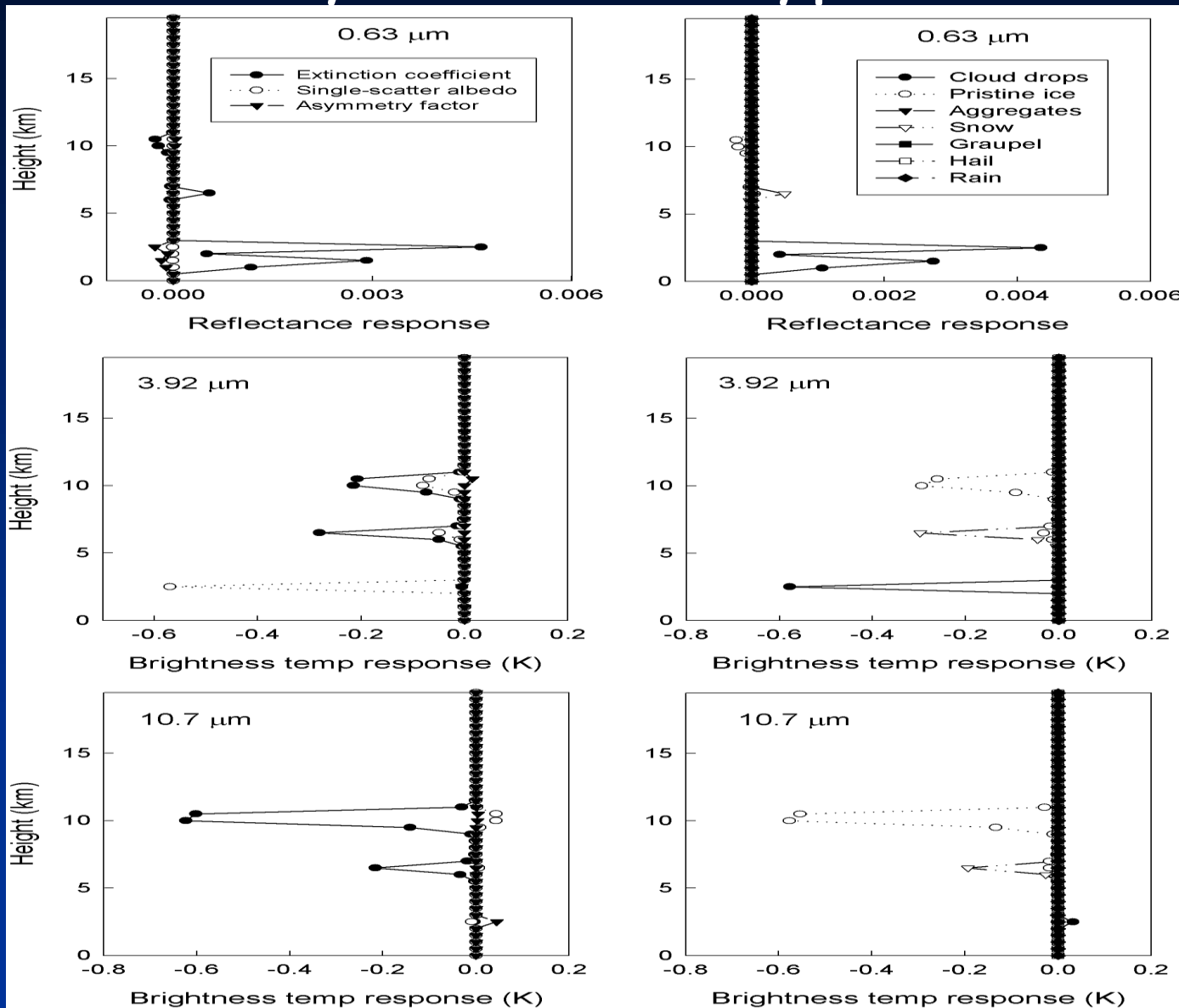
Near IR

IR

- Vertical and horizontal variability
- Sensitivity to multiple cloud layers



# Sensitivity by optical properties and hydrometeor type



# Assimilation experiments

## • Set 1 (Version 1 observation operator)

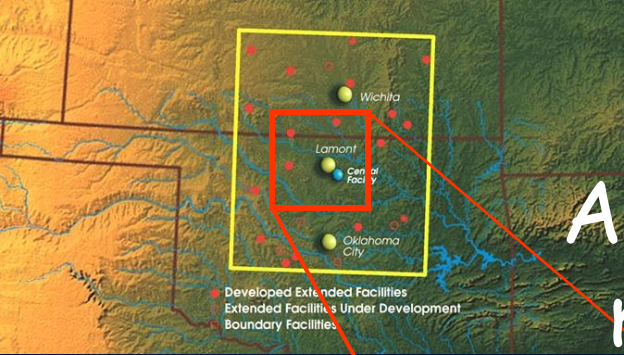
- Case with 100% cloud cover in the model domain
- Crude estimates of data assimilation parameters
- GOES IR only
- Sensitivity to observations

## • Set 2 (SHDOMPPDA operator)

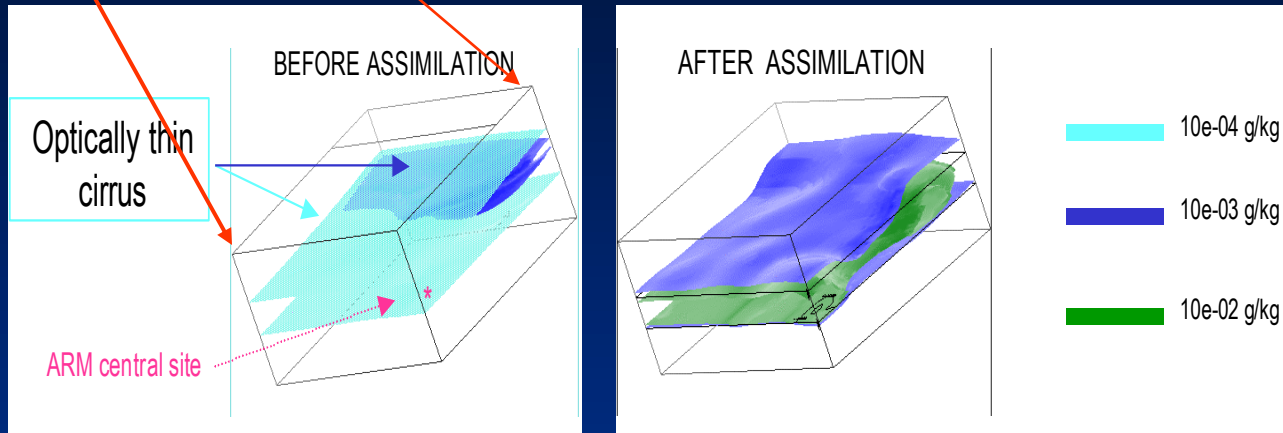
- Estimates of background biases and error correlation length from model validation with observations
- Use of cloud mask in quality control
- Cases with mixed clear and cloudy scenes
- Sensitivity to data assimilation parameters

Set 1 (Vukicevic et al, 2004, 2006)

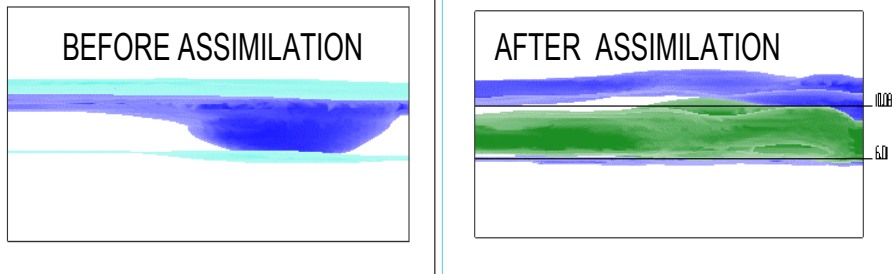
# Assimilation of GOES imager IR multi-layered stratiform case



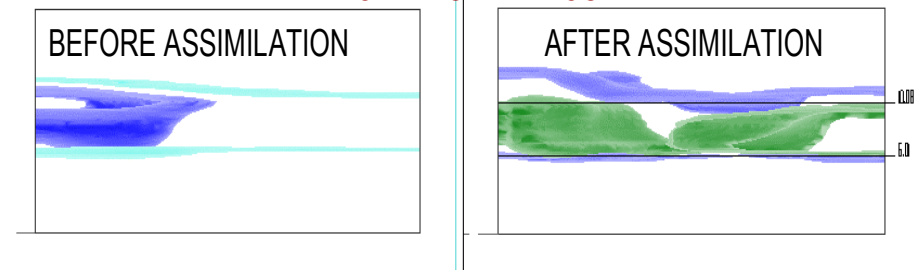
Model  
3D  
cloud



VIEW FROM SOUTHERN BOUNDARY



VIEW FROM WESTERN BOUNDARY

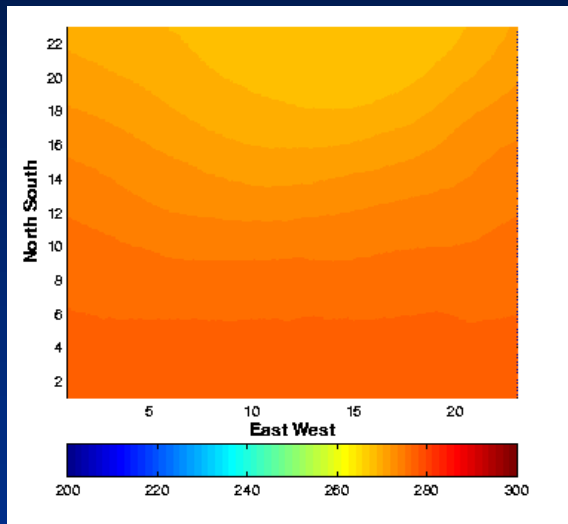


Observations every 15 min

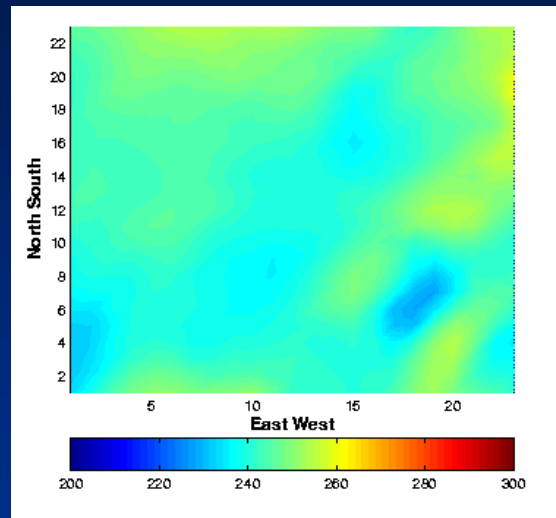
End time shown



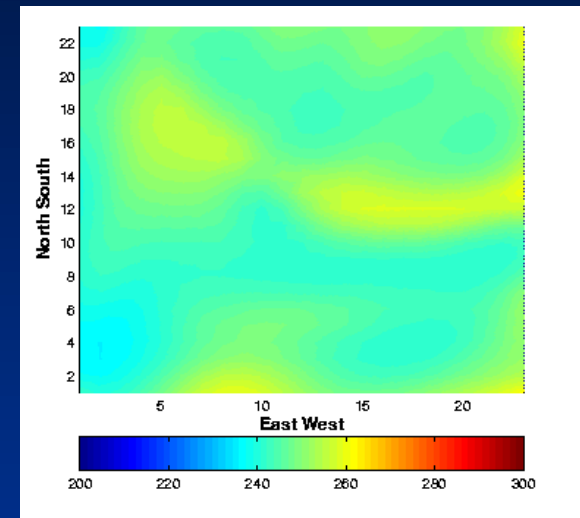
# Large amplitude bias and poor spatial variability are corrected simultaneously



prior



Observations



posterior

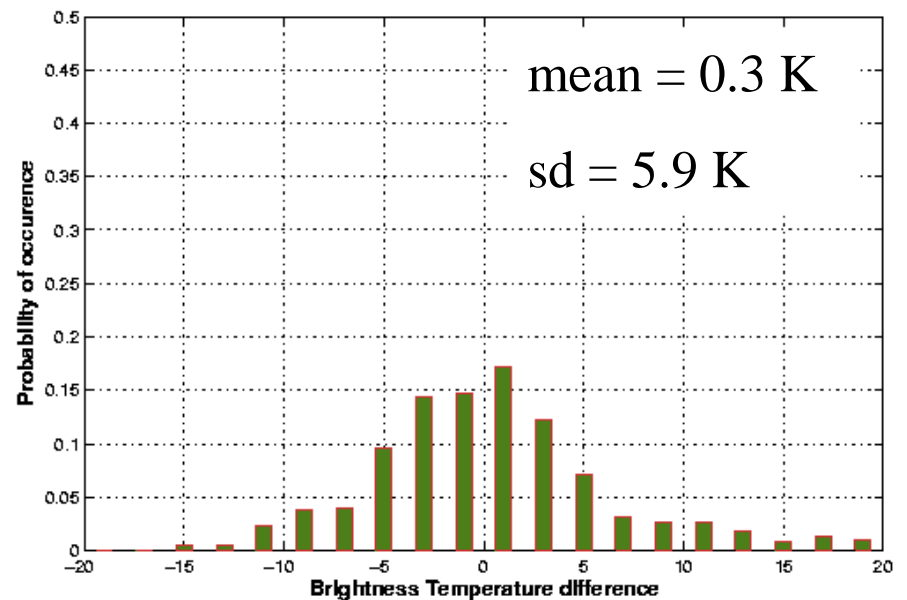
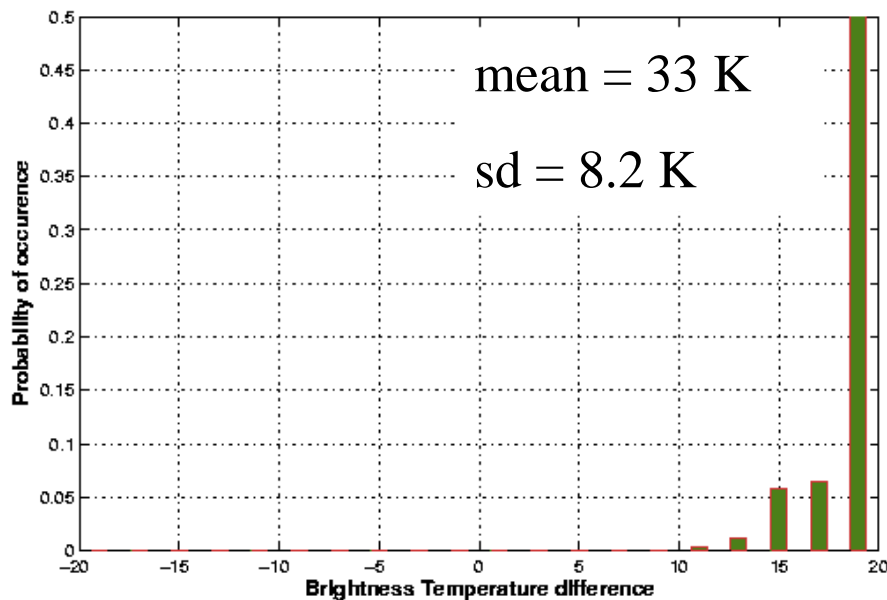
1 h window, every 15 min  
end time shown

# GOES imager IR error statistics (model - observation)

prior

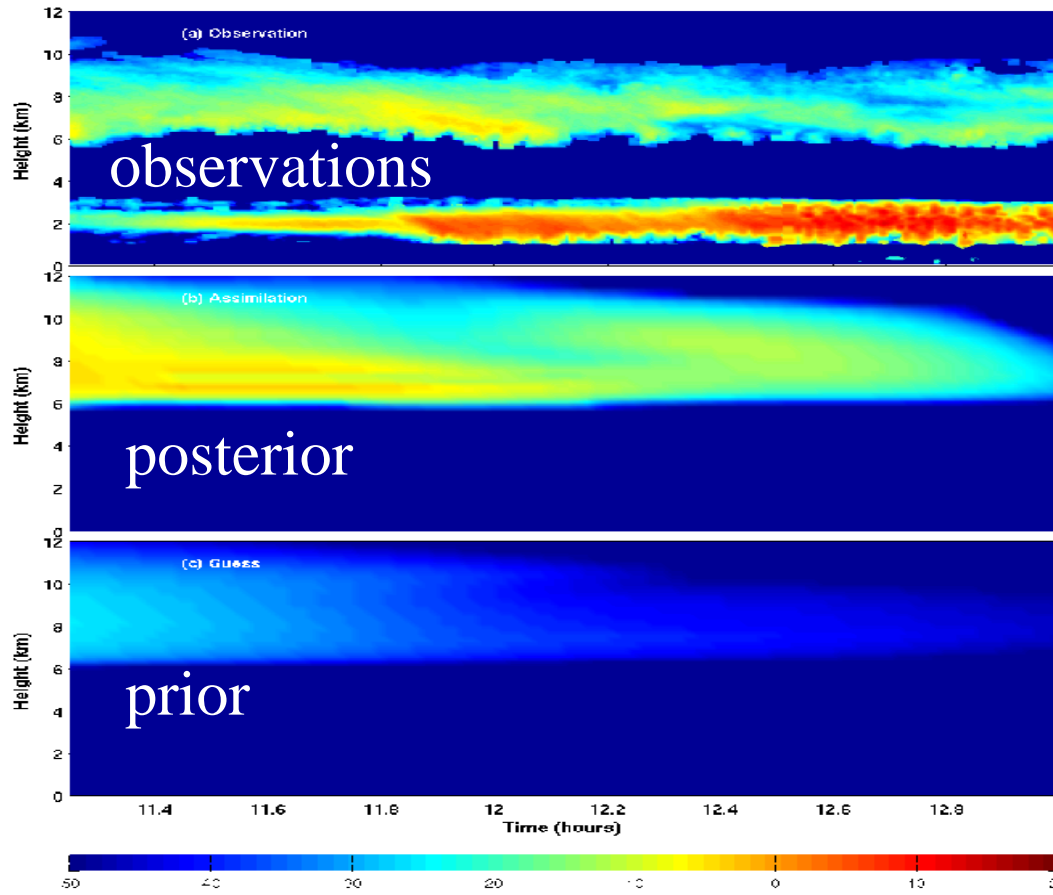
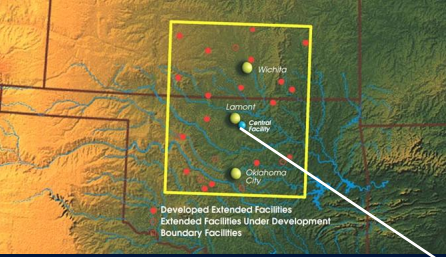
Brightness Temperature

posterior



# Verification with independent cloud observations

## ARM Cloud Radar reflectivity



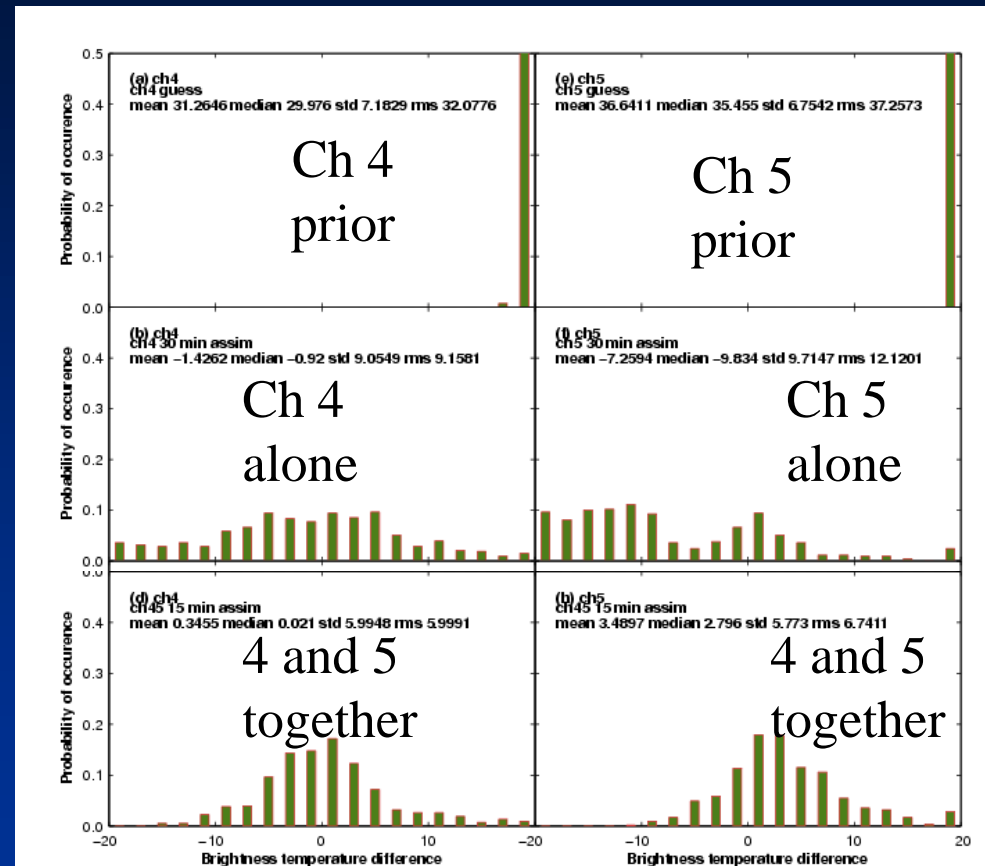
Ice cloud  
Liquid cloud

Time



# Complementary information from IR channels

Sensitivity of Tb in 10.7nm and 12.0 nm to clouds is very similar.



Model - Observations  
brightness temperature

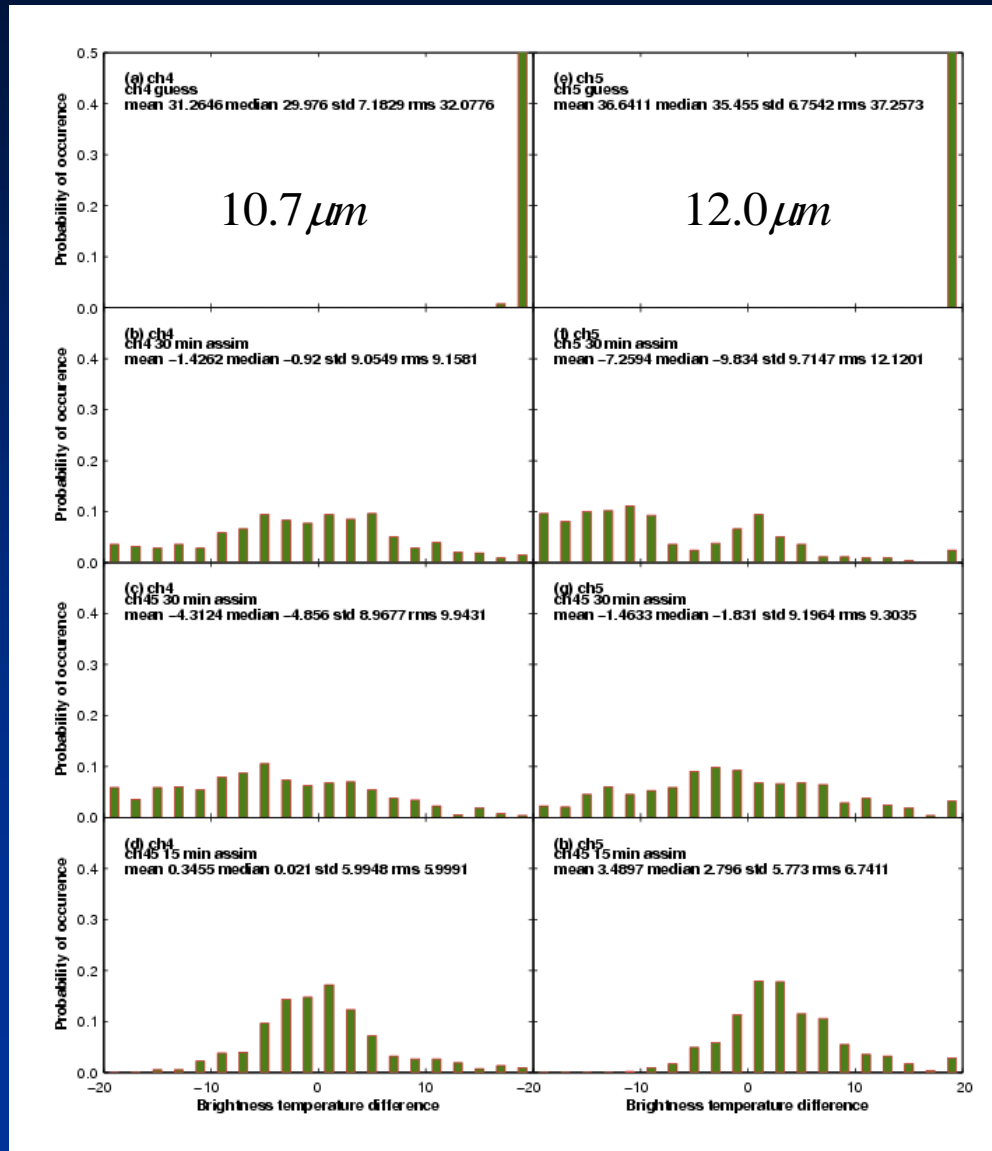
# Sensitivity to observation frequency

Tb errors

Worst



Best



Guess

Single channel assimilation, 30 min frequency

2-channel assimilation, 30 min frequency

2-channel assimilation, 15 min frequency

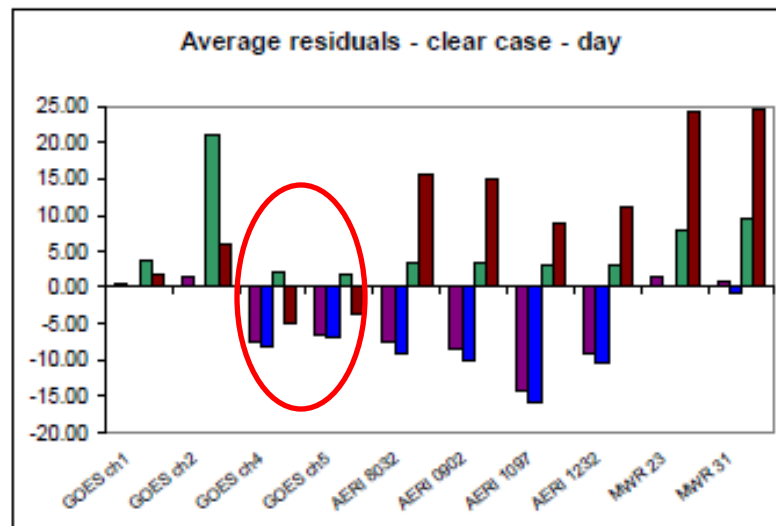
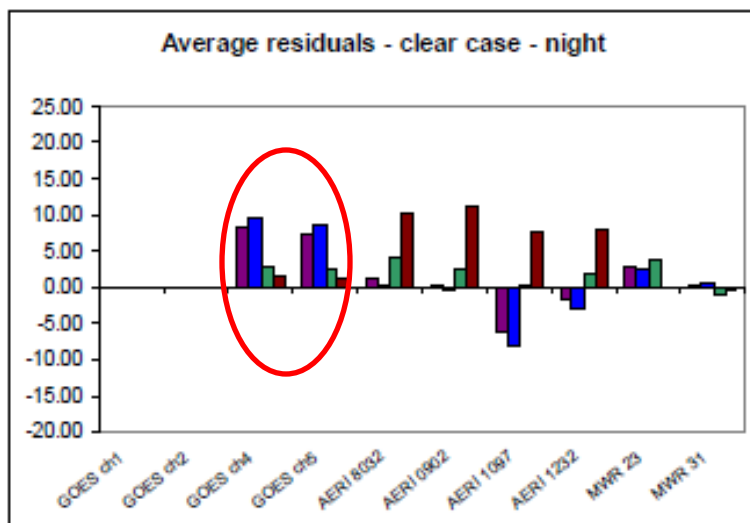
## Set 2

- Estimate of background biases and error correlation length from model validation with observations
- Use of cloud mask in quality control
- Cases with mixed clear and cloudy scenes
- Assimilation of visible and ground-based ARM observations
- Sensitivity to data assimilation parameters
  - QC
  - Background error decorrelation length
  - Spatial smoothing in RAMS adjoint
  - Length of assimilation window

# Estimates of background statistics using GOES imager and ARM data (Polkinghorne et al., 2008)

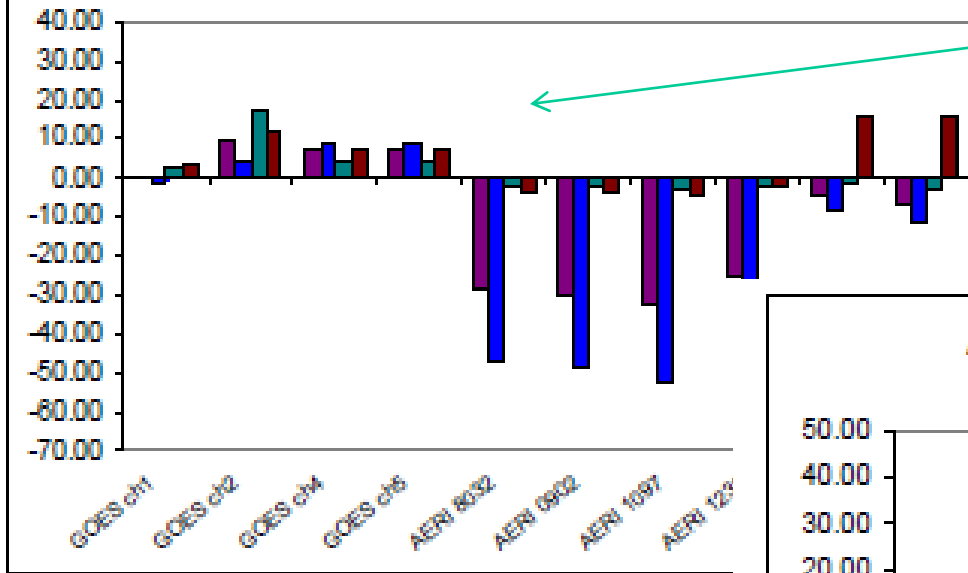
- 280 verification times
- Domain centered on ARM central facility
- grid dx=4km

Clear sky : emphasis on surface temperature bias



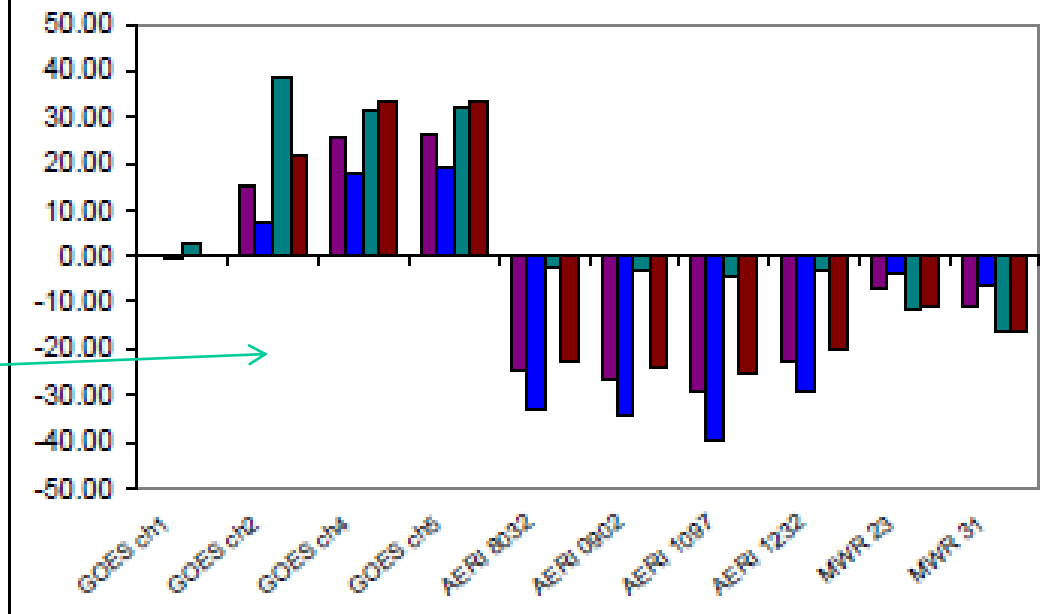
# Biases in cloudy conditions

Average residuals - low cloud - overall



Low clouds

Average residuals - high cloud - overall



High clouds



# Large biases in cloudy condition motivate design of cloud mask for QC

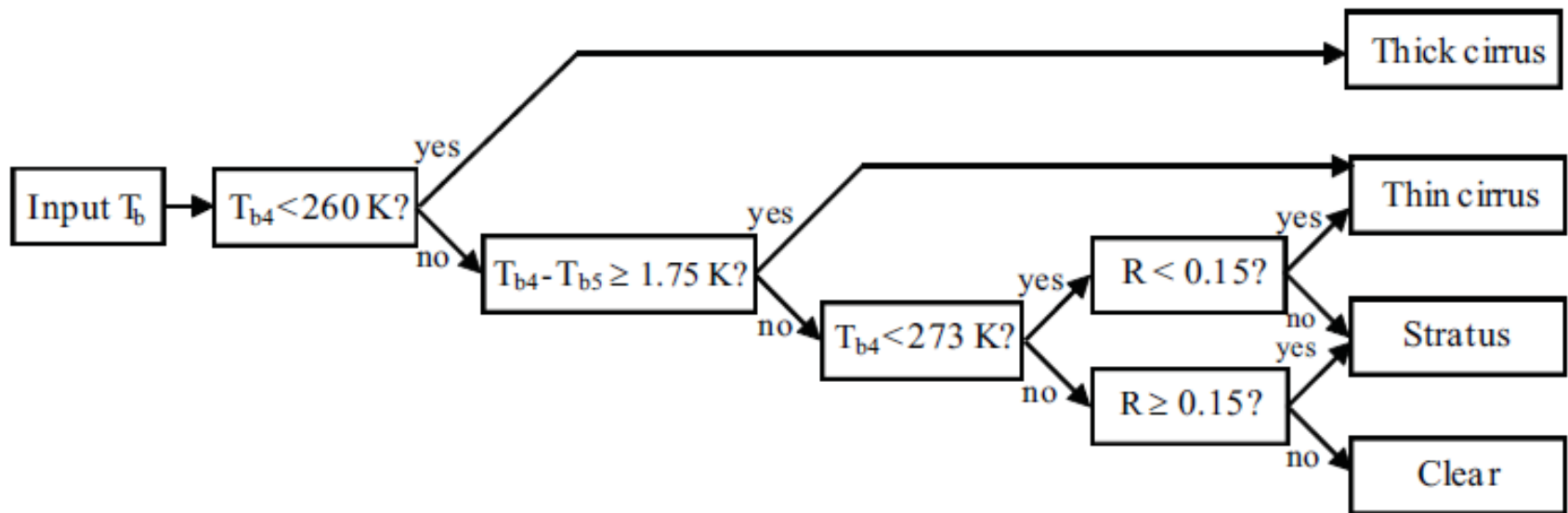
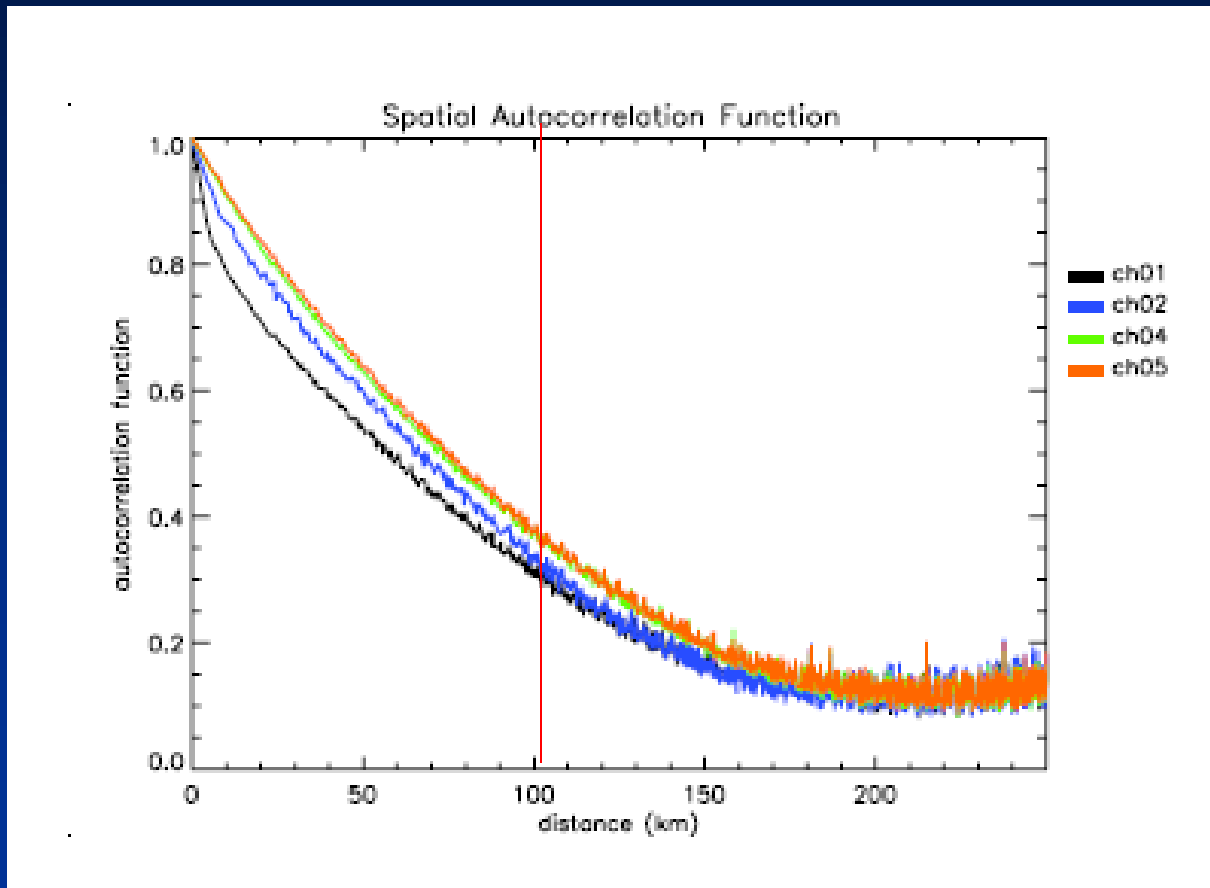


Figure 4: A flowchart representing the cloud mask algorithm.  $T_b$  is brightness temperature,  $T_{b4}$  is GOES channel 4 brightness temperature,  $T_{b5}$  is GOES channel 5 brightness temperature,  $R$  is GOES channel 1 reflectivity.

# Background error correlation for cloud variables



# Experiments with mixed clear and cloudy scenes (Polkinghorne and Vukicevic, 2010)

## Observed and background IR

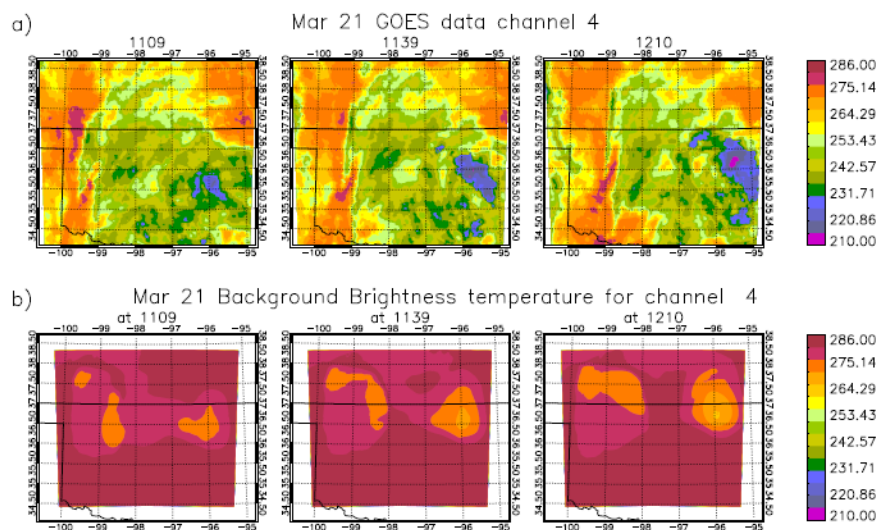


Figure 2: a) observed and b) simulated GOES channel 4 on Mar 21, 2000 at 1109, 1139, and 1210 UTC.

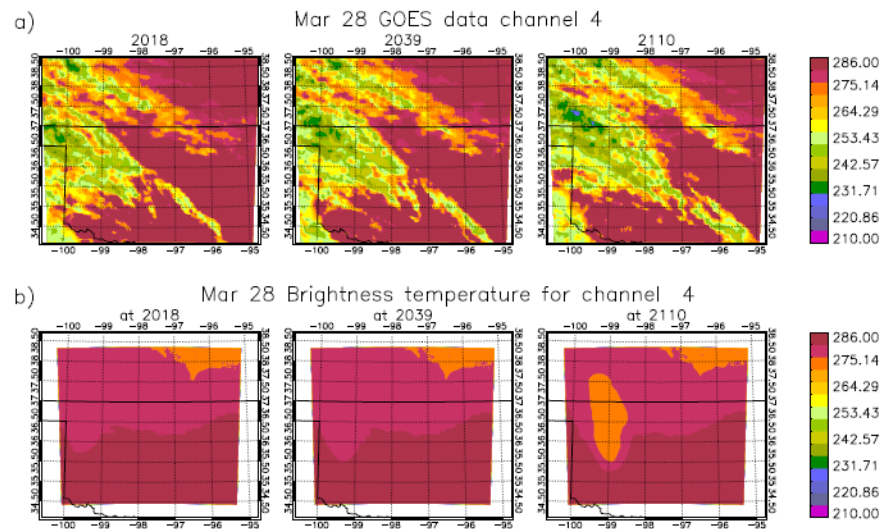


Figure 3: a) observed and b) simulated GOES channel 4 on Mar 28, 2000 at 2018, 2039, and 2110 UTC.

QC based only on maximum residual ; simple estimate of decorrelation length; IR channels ; 1 h window

1 -QC based on cloud mask ; simple estimate of decorrelation length; IR channels ; 1 h window ; variable max residual

2 - As 1d with spatial smoothing of RAMS adjoint solution

3 – As 1d with observation-based decorrelation length

4 – As 1d with 2 h window

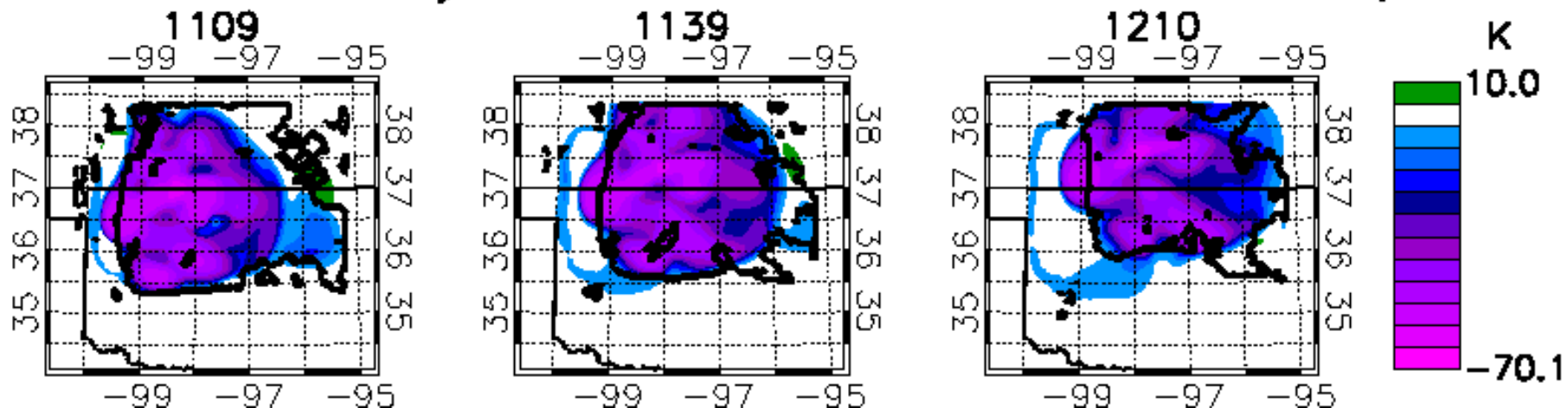
5 – As 1d with ARM observations

6 – As 1d with ARM observations

Exp #	Sensitivity experiments to data assimilation parameters
Control	Simple QC mask, residuals greater than 50 K excluded
1a	Cloud mask applied, hh residuals greater than 20 K excluded
1b	Cloud mask applied, hh residuals greater than 30 K excluded
1c	Cloud mask applied, hh residuals greater than 40 K excluded
1d	Cloud mask applied, hh residuals greater than 50 K excluded
2a	Experiment 1d with smoother applied to the adjoint solution, span=1
2b	Experiment 1d with smoother applied to the adjoint solution, span=2
3	Experiment 1d with background decorrelation length=100 km
4	Experiment 1d with longer assimilation window
5a	Experiment 1d with ground-based data assimilated at satellite assimilation times
5b	Experiment 1d with ground-based data assimilated every 5 minutes
5c	Experiment 5b with the background decorrelation length=100 km
6	Experiment 3 with assimilation of GOES channels 1 and 2

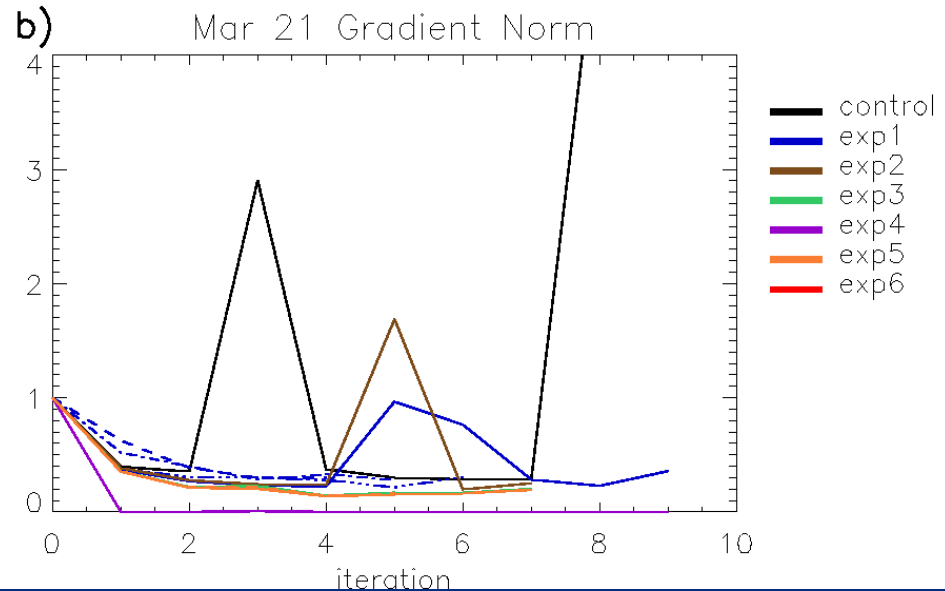
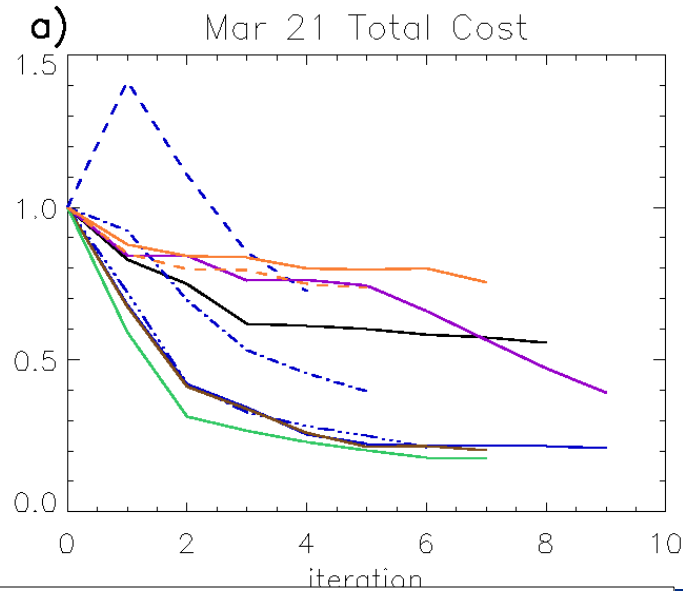
# Application of cloud mask in QC

d) Mar 21 Analysis—Back for channel 4 – exp 4

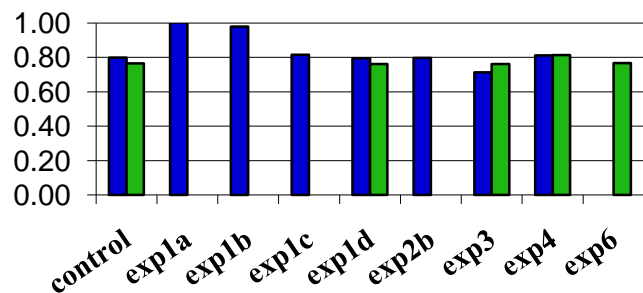


- Black contours mark boundary of regions within which the observations are used in assimilation
- Color shows impact of observations in the experiment with 2 h assimilation window

# Bulk results : convergence and global fit to observations



Change in fit to independent obs



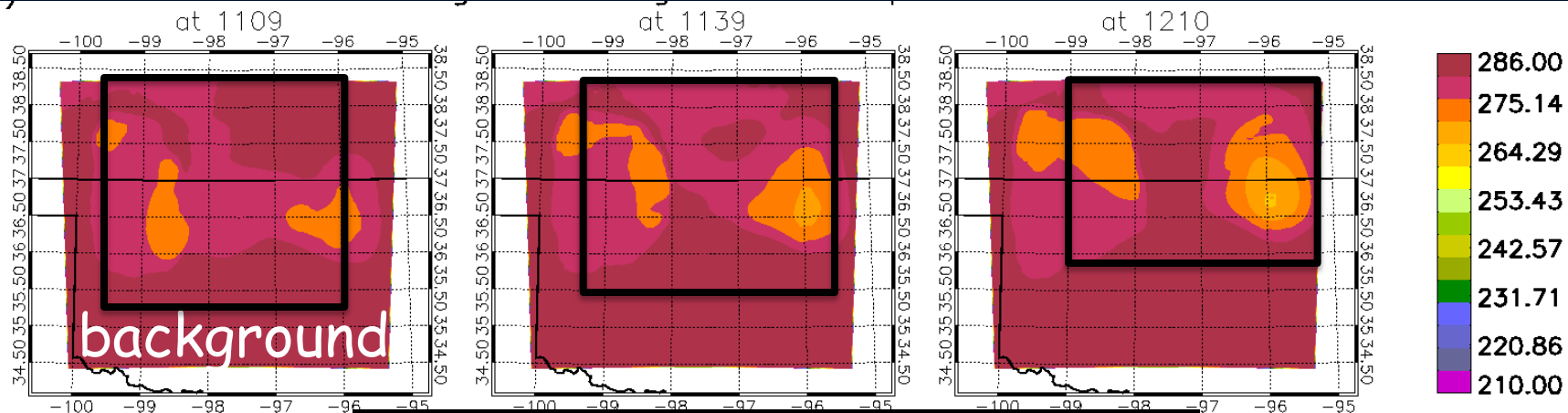
See additional slides for description of experiments

- Different convergence rate
- Similar final global fit to observations

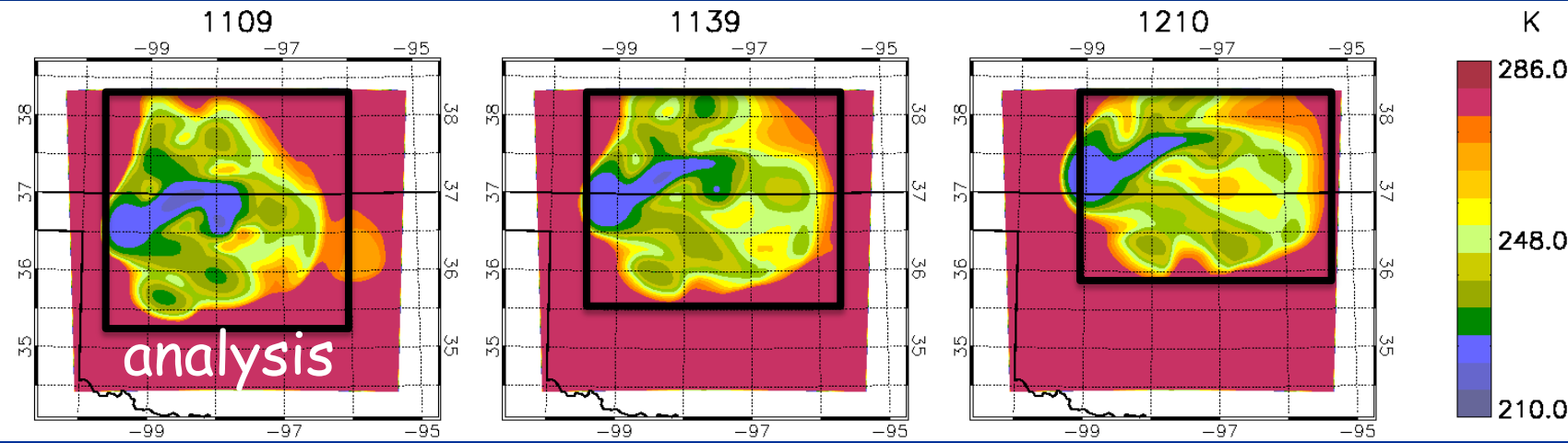
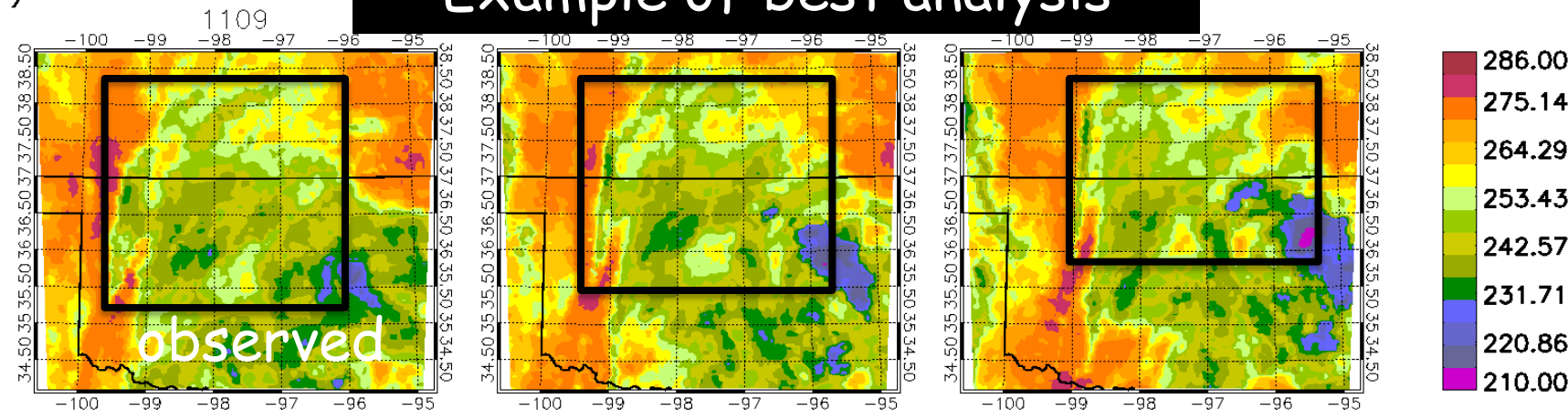
# Quality of analysis

- Despite small differences in the global fit to observations there are significant differences in quality of analysis between different experiments
- Best analysis is produced in the experiments that include the cloud mask in QC together with large allowed maximum residual, observation based decorrelation length and longer assimilation window
- Small but positive impact of VIS and ground-based remote sensing observations





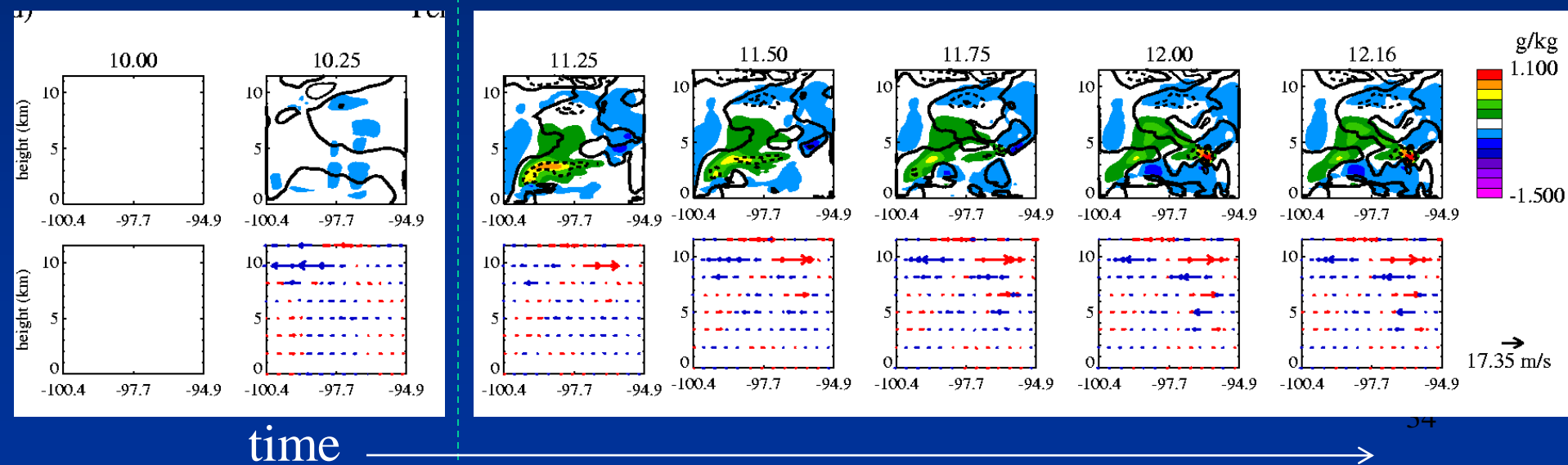
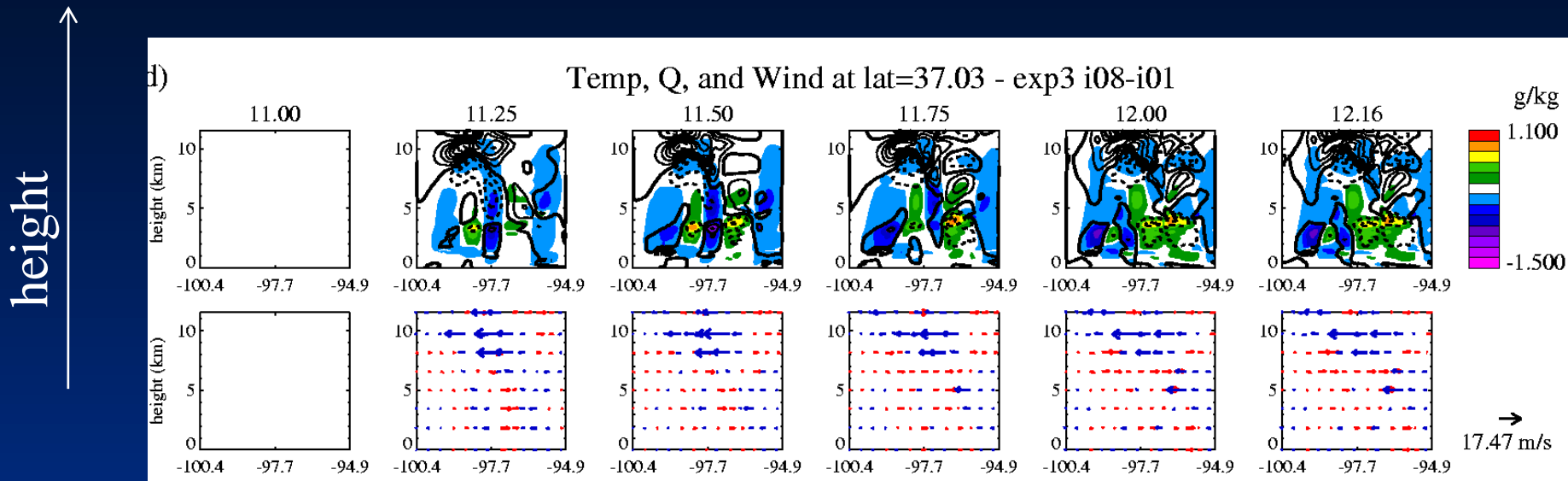
a) **Example of best analysis**







# Impact of longer assimilation window less noise/more balance



# Summary

- 4D, dynamically consistent analysis of cloudy atmosphere by assimilation of GOES imager observations is feasible
- The assimilation benefits from the use of cloud-mask based QC with large maximum residuals
- Balanced analysis requires sufficiently long assimilation window
- More frequent observations improve the analysis
- Window IR channels have complementary information
- Assimilation of visible observations has small impact in the studied cases that are dominated by ice clouds
- Assimilation of ground based remote sensing has small but positive local impact