

# OAR/ESRL/GSD

## Activities

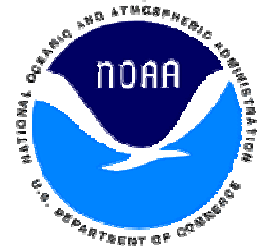
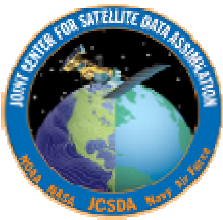
May 30-31, 2007 SSC

Dan Birkenheuer

(for Steve Koch)

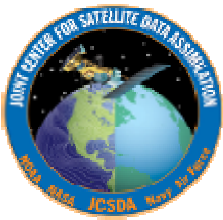
303 497 5584

[daniel.l.birkenheuer@noaa.gov](mailto:daniel.l.birkenheuer@noaa.gov)

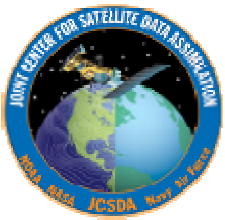


# Overview

- Background covariance work
- Bias correction for the current GOES (12)
- Model initialization, using cloud data
- Gradient assimilation
- Summarize



# Mesoscale background error covariance recovery using time-lagged ensembles



# Estimating background error covariance

$$\left\{ \begin{aligned} J(\mathbf{x}) &= \frac{1}{2}(\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) + \frac{1}{2}(\mathbf{y} - \mathbf{H}(\mathbf{x}))^T \mathbf{R}^{-1}(\mathbf{y} - \mathbf{H}(\mathbf{x})) \\ \mathbf{x}_a &= \mathbf{x}_b + \underbrace{\mathbf{B}\mathbf{H}^T(\mathbf{H}\mathbf{B}\mathbf{H}^T + \mathbf{R})^{-1}}_{\mathbf{K}} \mathbf{d} \end{aligned} \right.$$

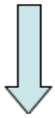
**B** is crucial for the Kalman-gain matrix in data assimilation

	<i>u</i>	<i>v</i>	<i>θ</i>	<i>p</i>	<i>q</i>
<i>u</i>	<i>B<sub>uu</sub></i>	<i>B<sub>uv</sub></i>	<i>B<sub>uθ</sub></i>	<i>B<sub>up</sub></i>	<i>B<sub>uq</sub></i>
<i>v</i>	<i>B<sub>uv</sub></i>	<i>B<sub>vv</sub></i>	<i>B<sub>vθ</sub></i>	<i>B<sub>vp</sub></i>	<i>B<sub>vq</sub></i>
<i>θ</i>	<i>B<sub>uθ</sub></i>	<i>B<sub>vθ</sub></i>	<i>B<sub>θθ</sub></i>	<i>B<sub>θp</sub></i>	<i>B<sub>θq</sub></i>
<i>p</i>	<i>B<sub>up</sub></i>	<i>B<sub>vp</sub></i>	<i>B<sub>θp</sub></i>	<i>B<sub>pp</sub></i>	<i>B<sub>pq</sub></i>
<i>q</i>	<i>B<sub>uq</sub></i>	<i>B<sub>vq</sub></i>	<i>B<sub>θq</sub></i>	<i>B<sub>pq</sub></i>	<i>B<sub>qq</sub></i>

**Fig. 1:** The background error covariance matrix (right) shown against the background state vector (left). Each square is itself a matrix of zonal and meridional wind, potential temperature, pressure, and humidity.



State vector



Background error covariance matrix



$$B = E[(X - \bar{X})(X - \bar{X})^T] = E[\delta X \delta X^T] =$$

$$\delta X = \begin{pmatrix} u_{11} \\ u_{12} \\ u_{13} \\ \vdots \\ u_{nn} \\ v_{11} \\ v_{12} \\ v_{13} \\ \vdots \\ v_{nn} \end{pmatrix}$$

$u_{11}u_{11}$	$u_{11}u_{12}$	$u_{11}u_{13}$	$\dots$	$u_{11}u_{nn}$	$u_{11}v_{11}$	$u_{11}v_{12}$	$u_{11}v_{13}$	$\dots$	$u_{11}v_{nn}$
$u_{12}u_{11}$	$u_{12}u_{12}$	$u_{12}u_{13}$	$\dots$	$u_{12}u_{nn}$	$u_{12}v_{11}$	$u_{12}v_{12}$	$u_{12}v_{13}$	$\dots$	$u_{12}v_{nn}$
$u_{13}u_{11}$	$u_{13}u_{12}$	$u_{13}u_{13}$	$\dots$	$u_{13}u_{nn}$	$u_{13}v_{11}$	$u_{13}v_{12}$	$u_{13}v_{13}$	$\dots$	$u_{13}v_{nn}$
$\vdots$	$\vdots$	$\vdots$	$\dots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\dots$	$\vdots$
$u_{nn}u_{11}$	$u_{nn}u_{12}$	$u_{nn}u_{13}$	$\dots$	$u_{nn}u_{nn}$	$u_{nn}v_{11}$	$u_{nn}v_{12}$	$u_{nn}v_{13}$	$\dots$	$u_{nn}v_{nn}$
$v_{11}u_{11}$	$v_{11}u_{12}$	$v_{11}u_{13}$	$\dots$	$v_{11}u_{nn}$	$v_{11}v_{11}$	$v_{11}v_{12}$	$v_{11}v_{13}$	$\dots$	$v_{11}v_{nn}$
$v_{12}u_{11}$	$v_{12}u_{12}$	$v_{12}u_{13}$	$\dots$	$v_{12}u_{nn}$	$v_{12}v_{11}$	$v_{12}v_{12}$	$v_{12}v_{13}$	$\dots$	$v_{12}v_{nn}$
$v_{13}u_{11}$	$v_{13}u_{12}$	$v_{13}u_{13}$	$\dots$	$v_{13}u_{nn}$	$v_{13}v_{11}$	$v_{13}v_{12}$	$v_{13}v_{13}$	$\dots$	$v_{13}v_{nn}$
$\vdots$	$\vdots$	$\vdots$	$\dots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\dots$	$\vdots$
$v_{nn}u_{11}$	$v_{nn}u_{12}$	$v_{nn}u_{13}$	$\dots$	$v_{nn}u_{nn}$	$v_{nn}v_{11}$	$v_{nn}v_{12}$	$v_{nn}v_{13}$	$\dots$	$v_{nn}v_{nn}$

Green: variance, Blue: spatial covariance, Yellow: cross covariance, Red: cross-spatial covariance

# Time-phased model ensemble system

- Hourly runs of the 40-km RUC forecast within a 12-h cycle (previous forecast up to 12h)

- Advantages of the method:

- ① mesoscale features
- ② flow-dependent features
- ③ short-range forecast
- ④ potential large samples (e.g., multi-model ensembles)
- ⑤ on-line cycling
- ⑥ economical computing

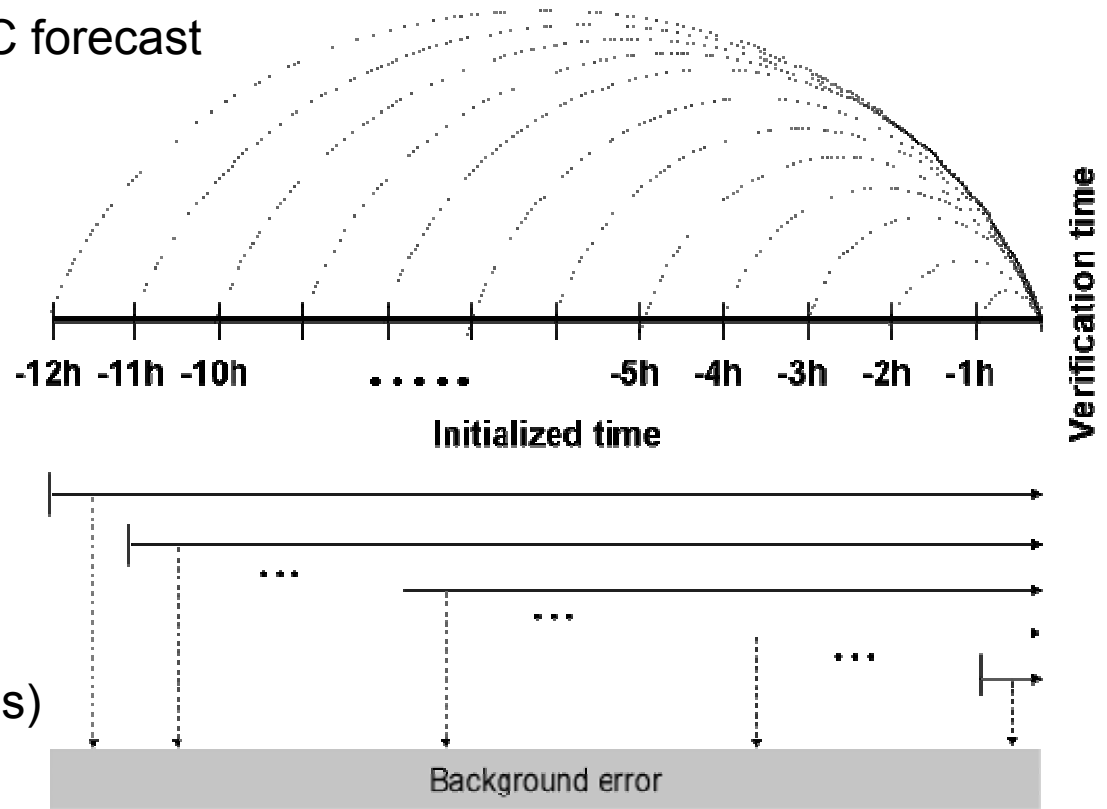
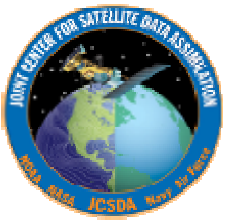
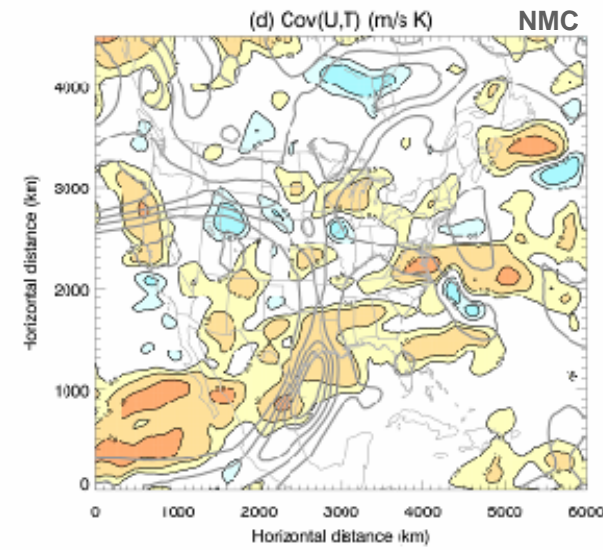
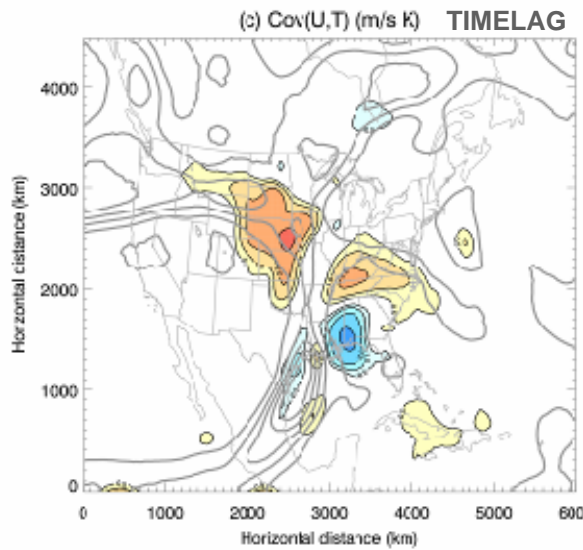
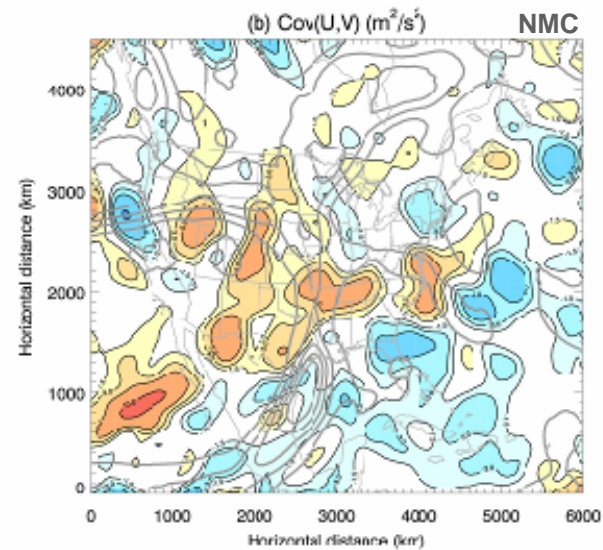
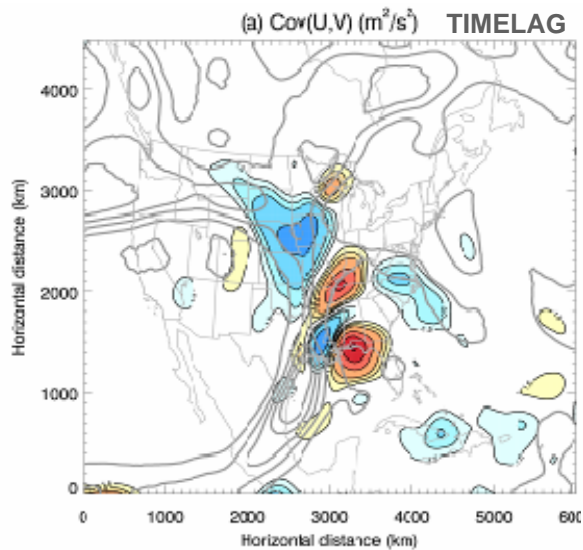


Fig. 4: Schematic illustration of time-phased ensemble method for estimating error covariance.

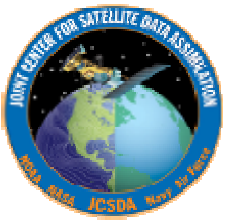


TIMELAG



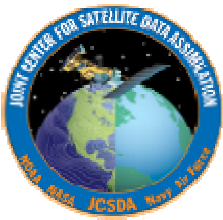
NMC

Horizontal plot of Cov: TIMELAG vs. NMC

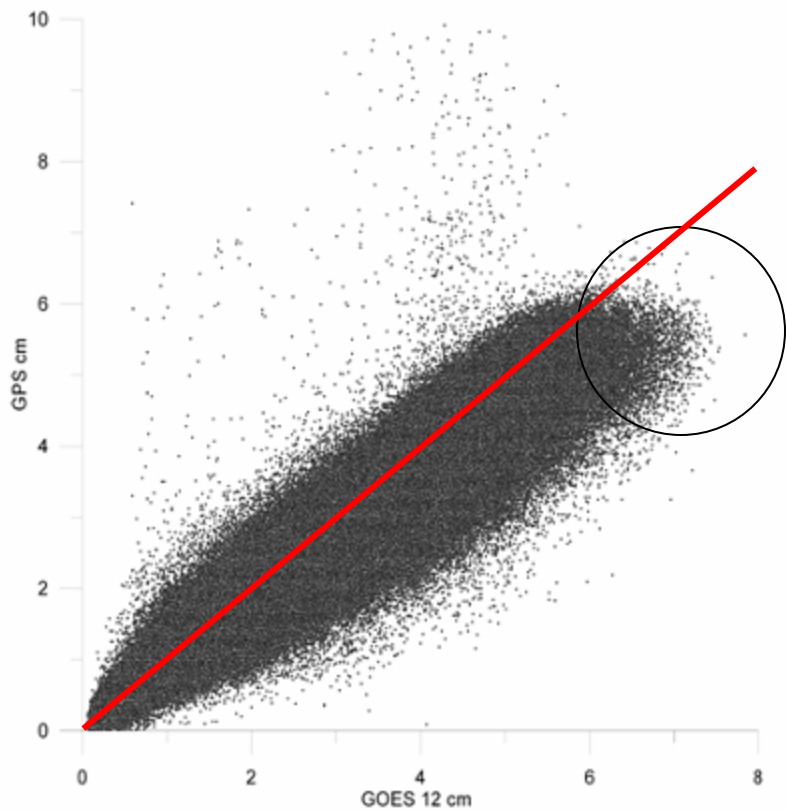


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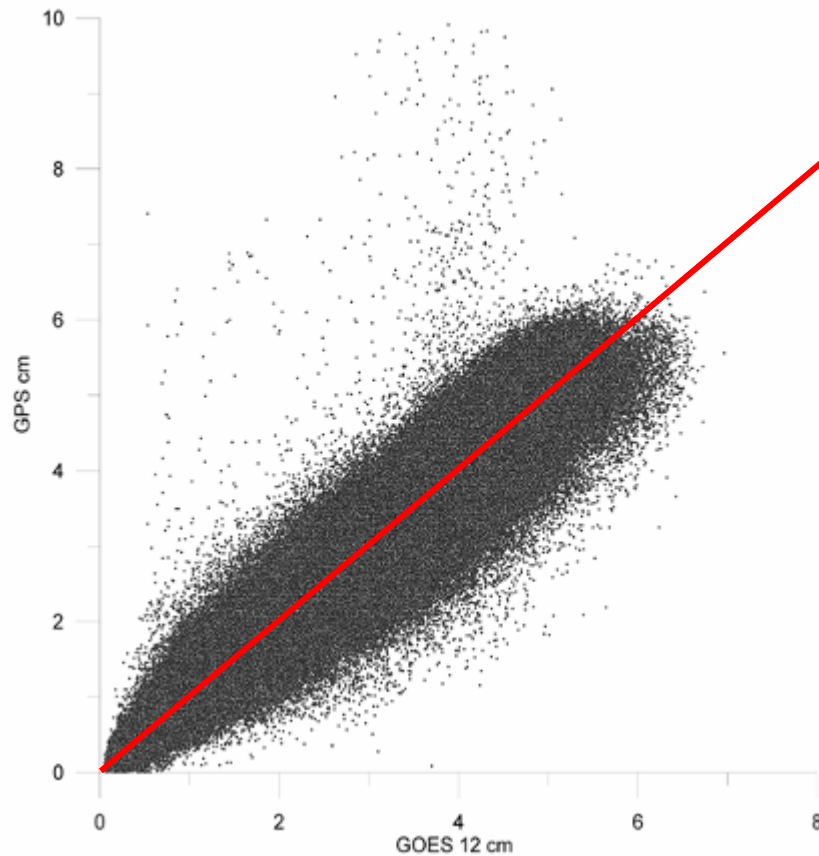




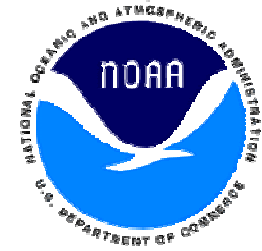
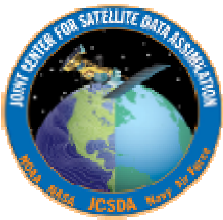


Scatter plot of raw data

Scatter plot of bias corrected data



1.8 Million Data Pairs (20min, 10km)



# Algorithm

$$g_c = ag^b$$

a

b

h

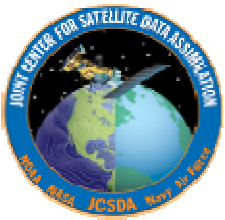
0.979470611	0.952045858	0
0.96386236	0.958807886	1
0.951016307	0.962379932	2
0.932851493	0.974993765	3
0.938412488	0.973992229	4
0.928518832	0.971161544	5
0.932472348	0.975237787	6
0.936737478	0.97503674	7
0.943030536	0.971995413	8
0.945574582	0.972088754	9
0.953864217	0.967487574	10
0.952823639	0.967738211	11
0.944226384	0.970142543	12

a

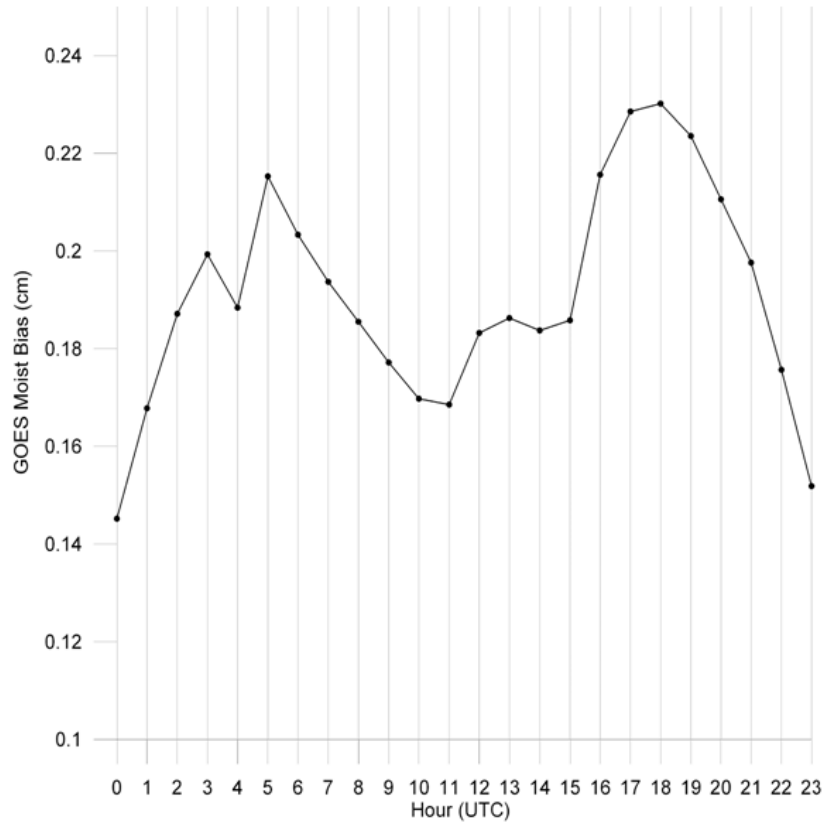
b

h

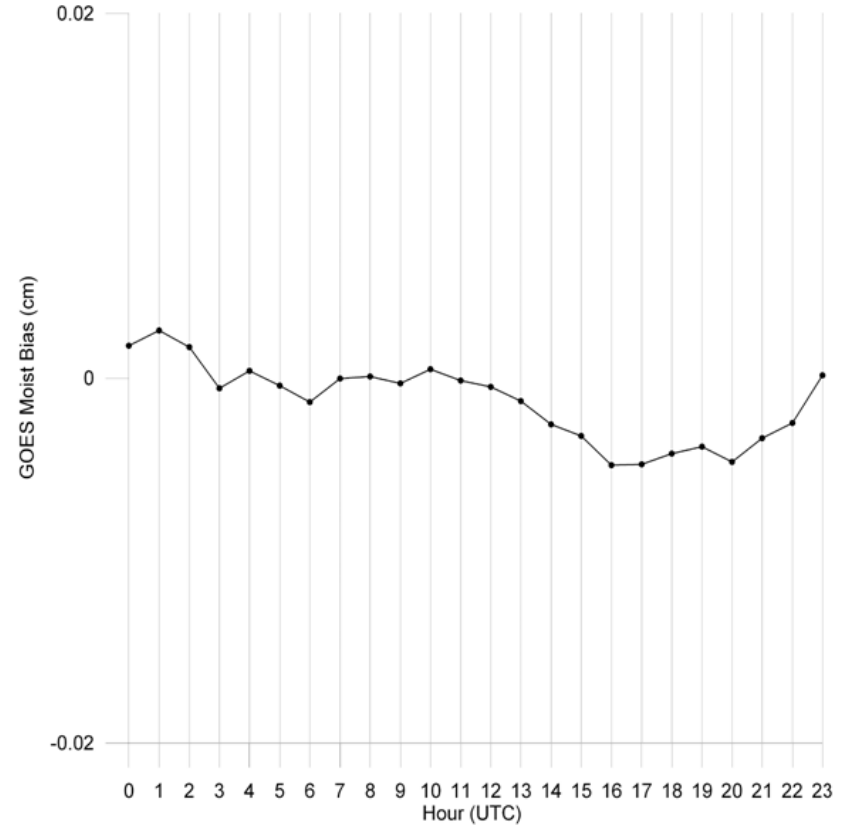
0.934683204	0.977410853	13
0.928368866	0.98369354	14
0.923411667	0.988313854	15
0.90421778	0.997356713	16
0.896550059	1.00138319	17
0.896099865	1.00216639	18
0.900296807	1.00008261	19
0.905209124	1.00010216	20
0.923843801	0.986412048	21
0.942986071	0.975428104	22
0.970267594	0.958948851	23



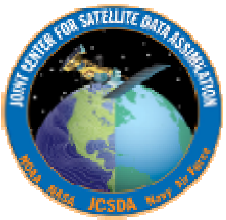
# Bias Correction



RAW data

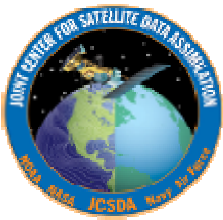


Bias Corrected



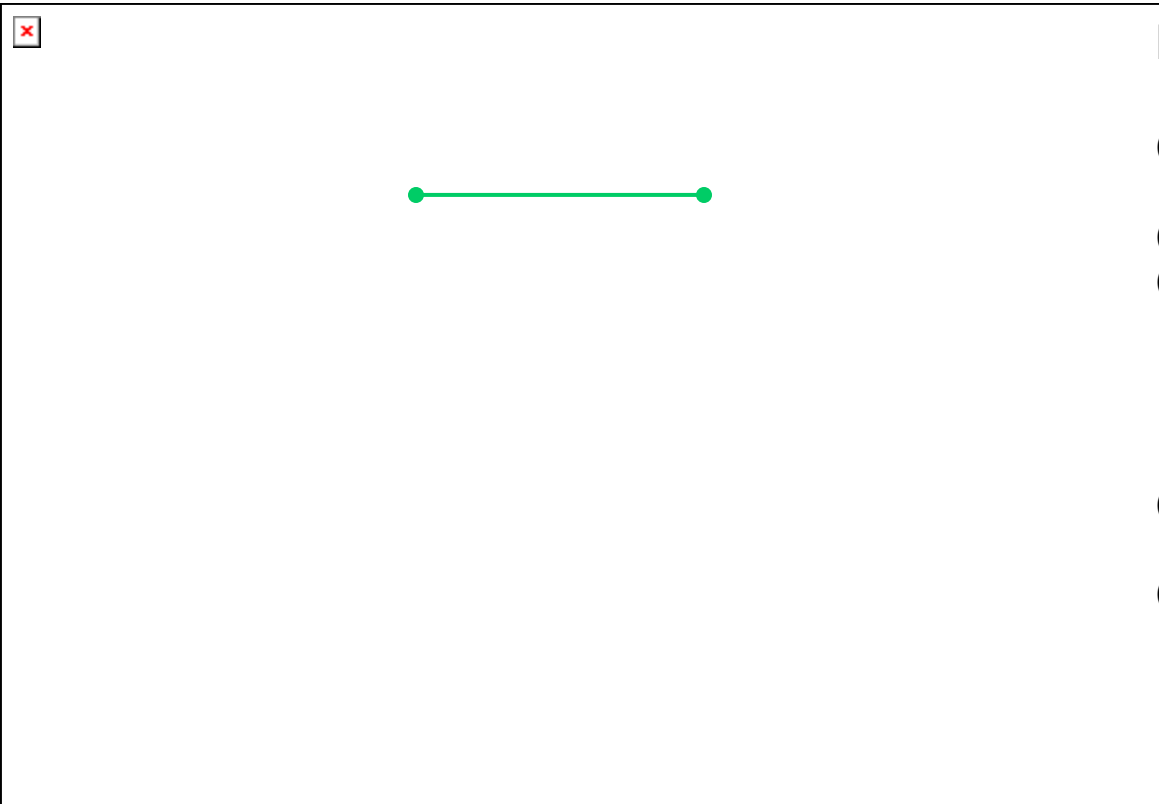
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# Surface Temperature and Pressure

## 21 UTC 2002 June 12

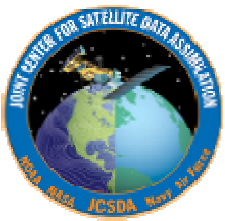


### Diabatic Initialization Procedure

- (1) Identify clouds in radar, **satellite**, and surface observations
- (2) Determine cloud depth
- (3) Assign a parabolic vertical velocity profile such that maximum vertical velocity is  $10 \text{ m s}^{-1}$  for a cloud depth of 10 km and model grid point spacing of 1 km.
- (4) Set cloud relative humidity with respect to water to 110%.
- (5) Compute balanced wind and pressure fields with 3D variational analysis using the continuity equation and velocity tendencies as constraints.

WRF-ARW simulation with 1-km grid increment

Light blue line shows the position of cross sections in subsequent slides. It intersects a developing storm within the >60 dBZ area.



# Diabatic Initialization

## 21 UTC 2002 June 12

WRF initialization fields (timestep=0)

Upper Right:

Green shading is relative humidity

White contours for  $RH > 100\%$  and  $110\%$

Black contours cloud+ice  $m_r > 0.1, 0.5$  g/kg

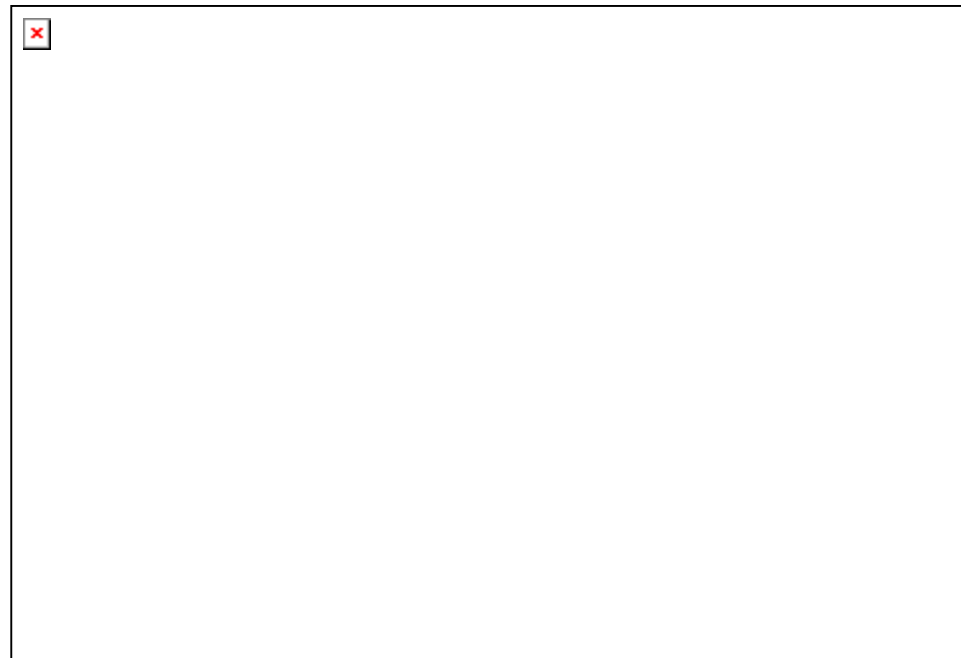
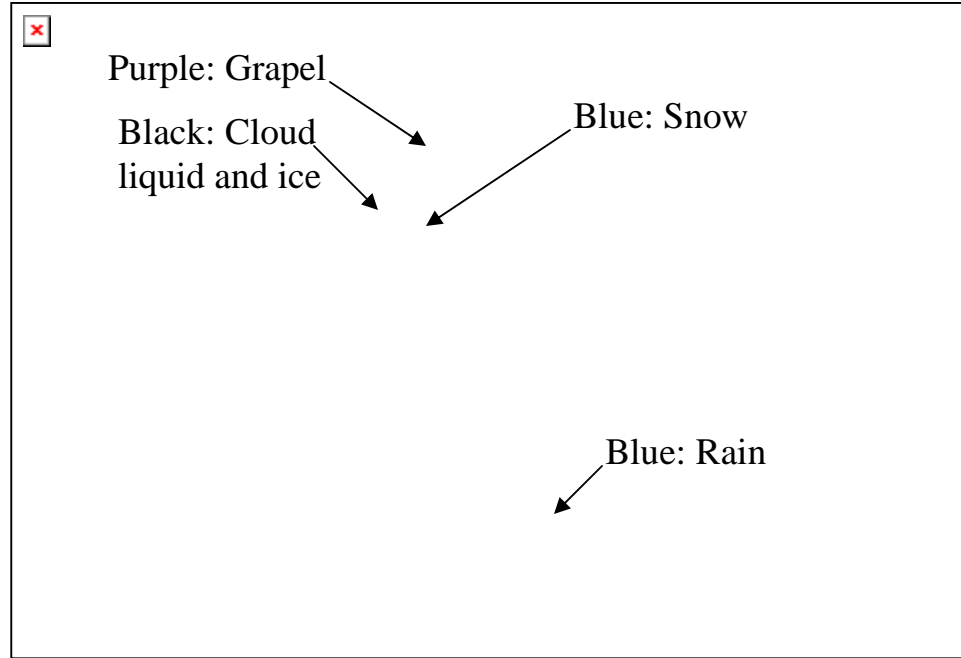
Blue contours snow  $m_r > 4.0, 8.0, 12.0$  g/kg

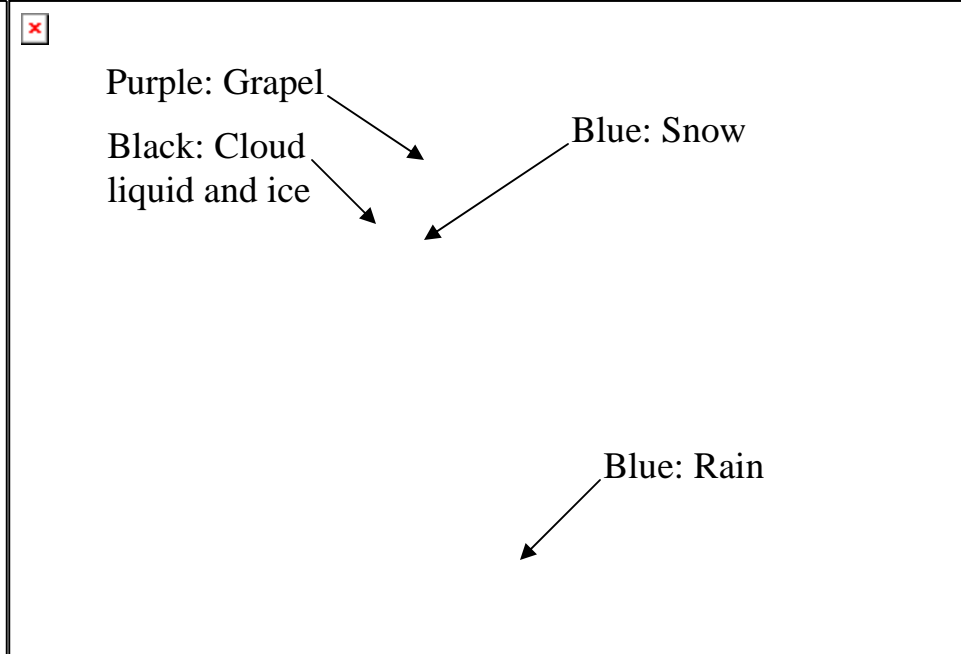
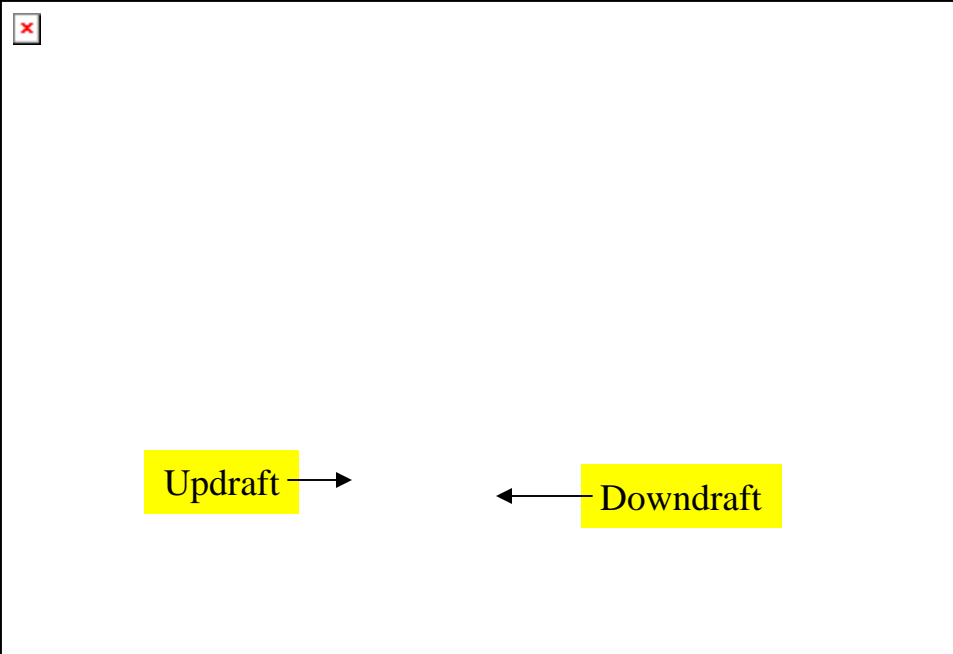
Blue contours rain  $m_r > 0.01, 0.1$  g/kg

Purple contours graupel  $m_r > 1.0, 2.0$  g/kg

Lower Right:

Reflectivity from Brian Colle's equation





WRF 5 second fields (timestep=1)

Upper Left:

Red shading diabatic heating

Blue shading is negative diabatic heating

Black contours are  $w > 0.2, 2.0$  by  $0.2$  m/s

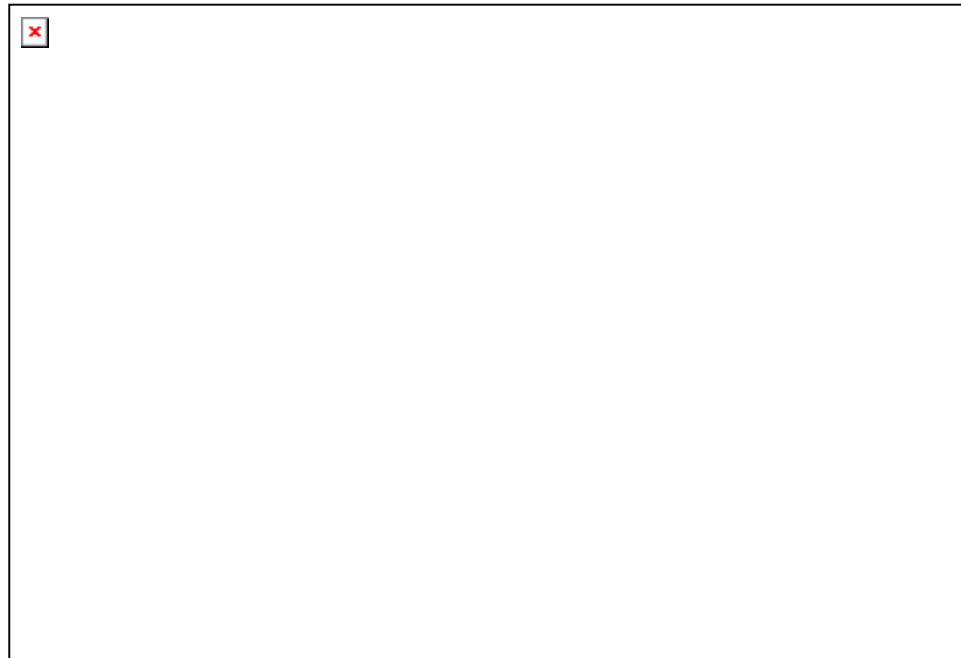
Black dash contours are  $w < -1.0, -10.0$  by  $1.0$  m/s

Upper Right:

Same as green scale in wrf initialization plot

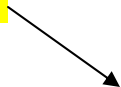
Lower Right:

Reflectivity from Brian Colle's equation





Updraft



Purple: Grapel

Black: Cloud  
liquid and ice

Blue: Snow



Blue: Rain



WRF 30 second fields (timestep=6)

Upper Left:

Red shading diabatic heating

Blue shading is negative diabatic heating

Black contours are  $w > 0.2, 2.0$  by  $0.2$  m/s

Black dash contours are  $w < -1.0, -10.0$  by  $1.0$  m/s

Upper Right:

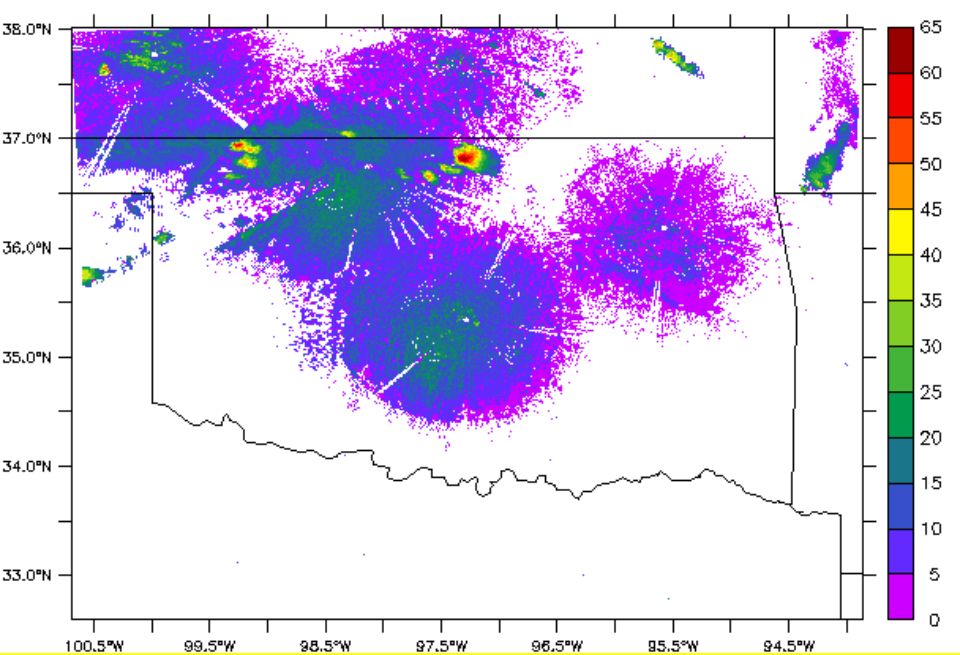
Same as green scale in wrf initialization plot

Lower Right:

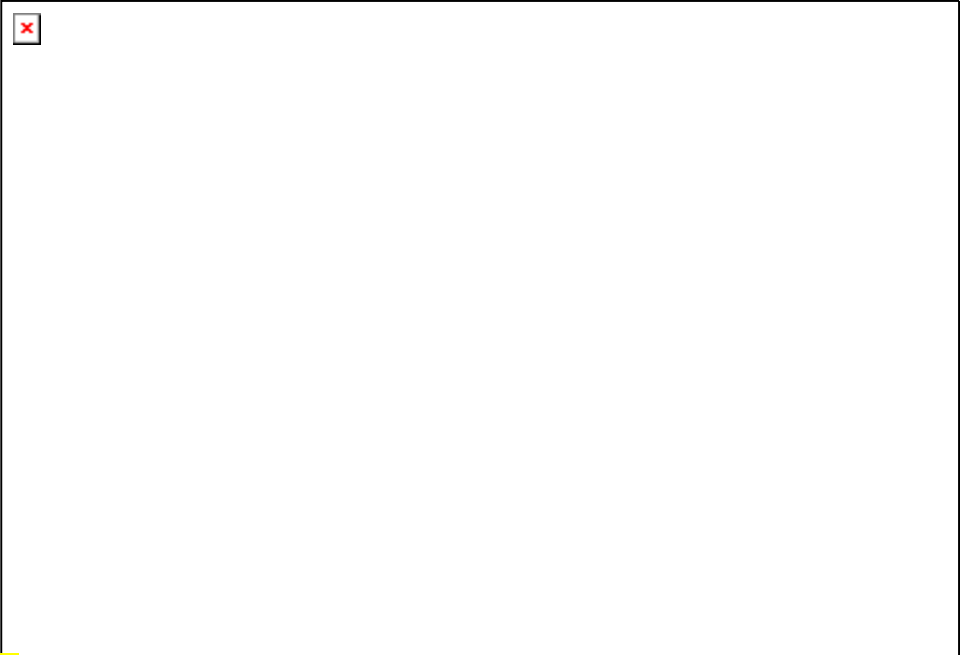
Reflectivity from Brian Colle's equation



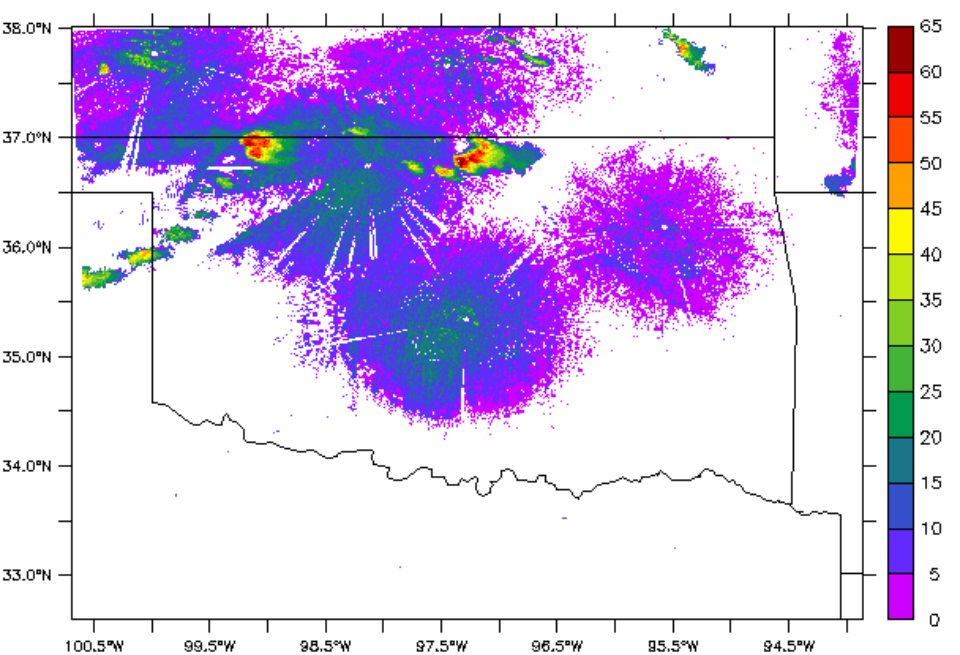




LAPS Analysis Column Maximum Reflectivity 10 minutes



WRF Column Maximum Reflectivity 10 minutes

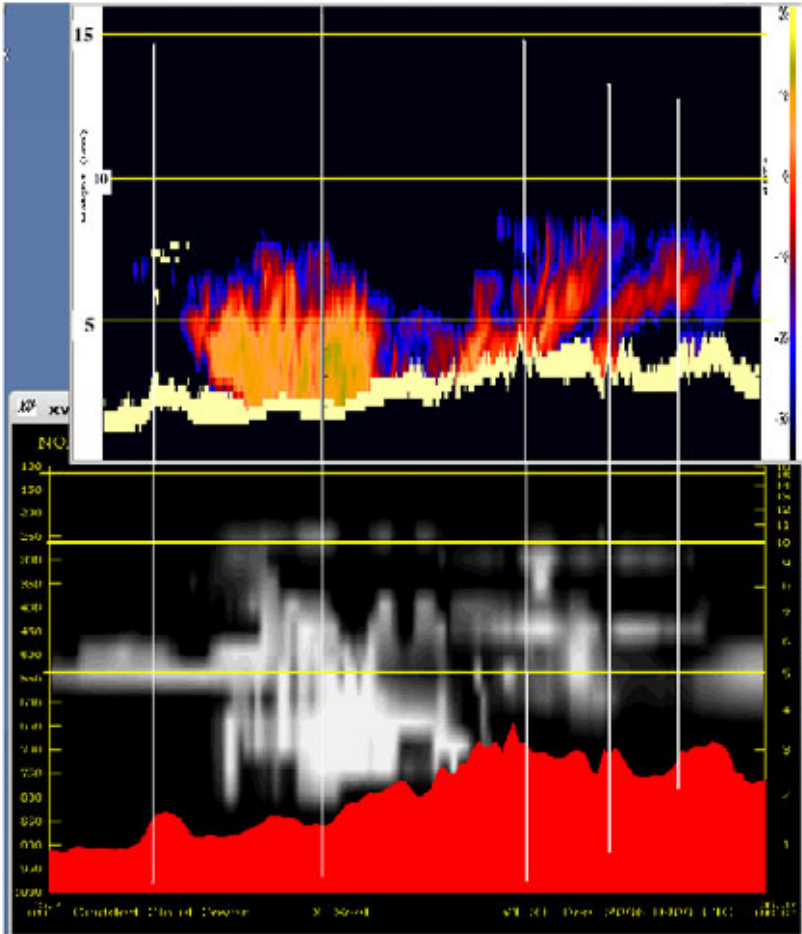


LAPS Analysis Column Maximum Reflectivity 30 minutes



WRF Column Maximum Reflectivity 30 minutes

# CloudSat and LAPS Comparisons



CloudSat

LAPS Cloud Analysis

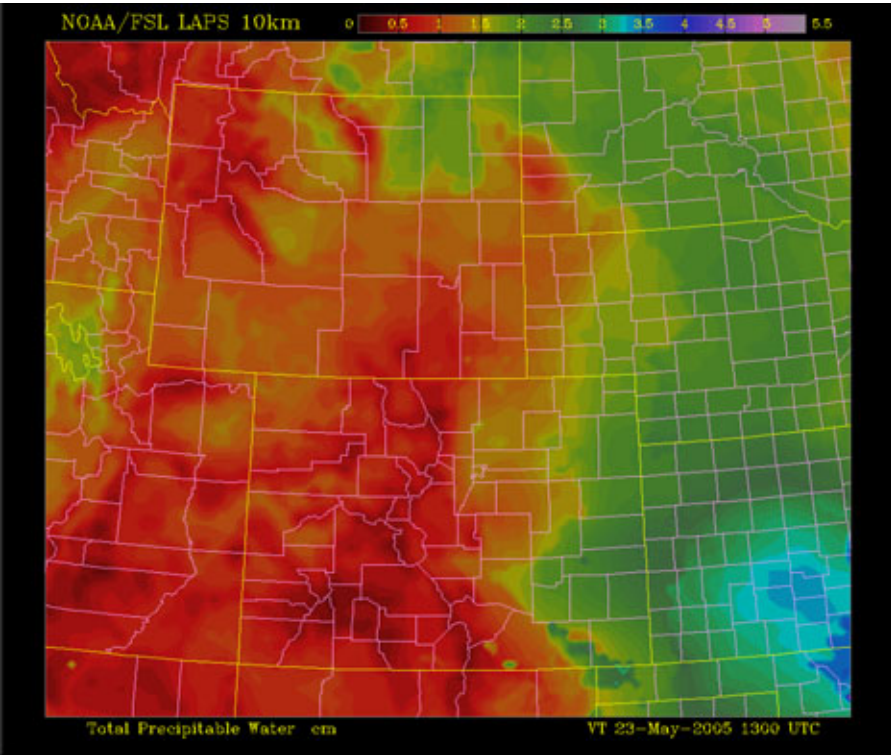
December 2006 Snowstorm case, Cross section running NE to SW through ROC

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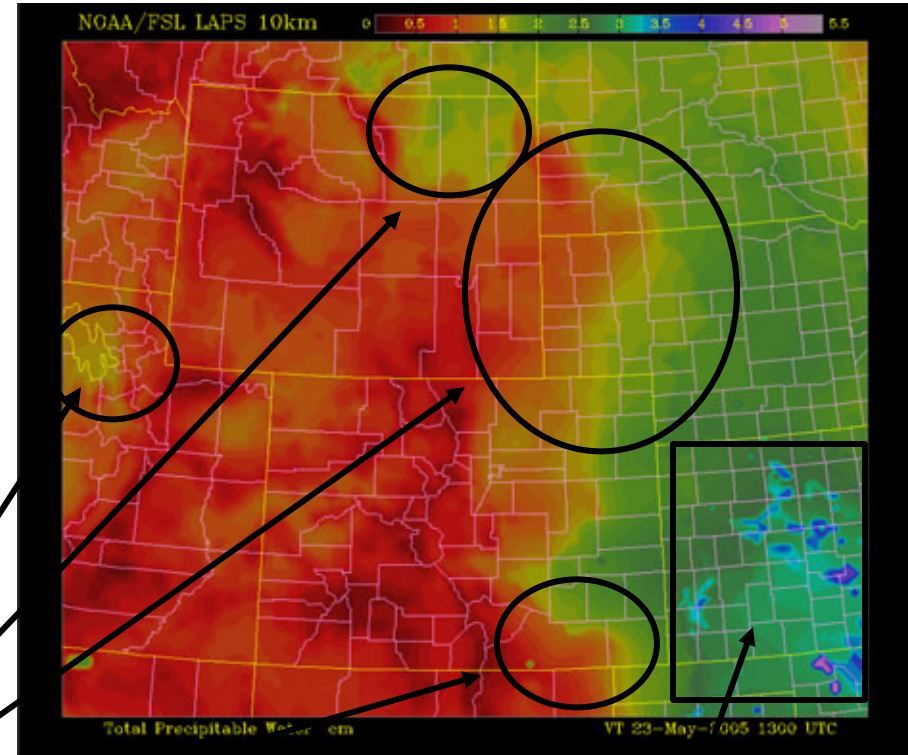


# Gradient approach is to utilize only satellite gradients in the analysis



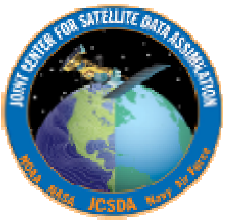
Old method

Less moist bias

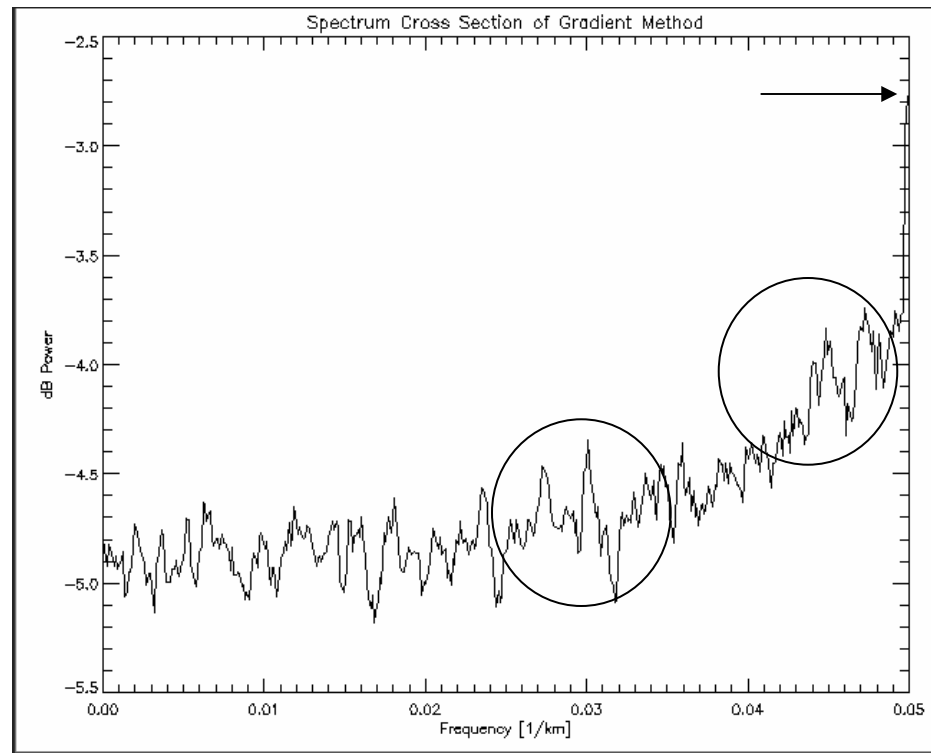
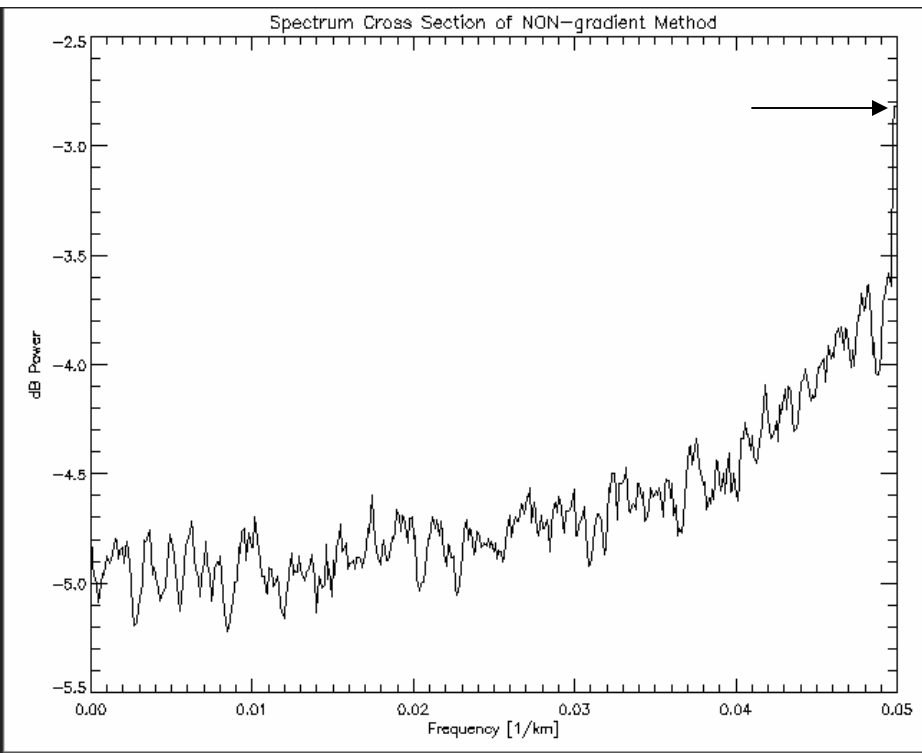


New method

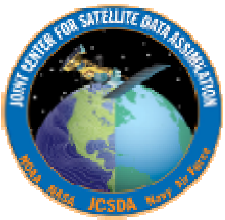
Better structure



# Non-Gradient Method Spectra



# Gradient Method Spectra



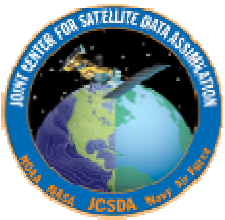
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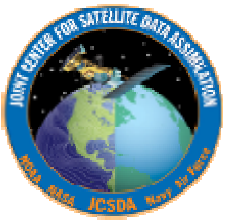
# Summary

- NOAA/OAR is well positioned to work in areas of JCSDA interest.
- In order for GSD as part of OAR to continue our relationship with JCSDA, we must get more support.
- No funding was transferred to OAR from the Joint Center in 06.
- Work shown here will die on the vine unless there is some change.



# It Goes Deeper than Funding Support

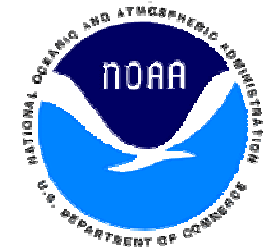
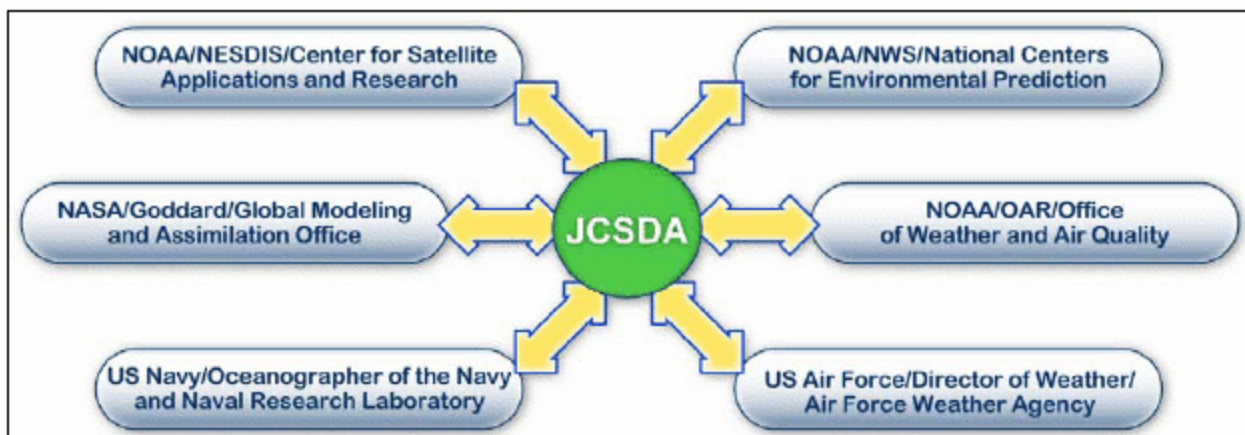
- Last year we prepared a comprehensive summary of cases to ascertain GOES moisture problems. This fostered little interaction.
- This year we have provided bias corrections for GOES 12 moisture to NESDIS and no action has resulted.
- We must cooperate to be successful.





# From the JCSDA Strategic Plan

## Inter-Agency Relationships



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