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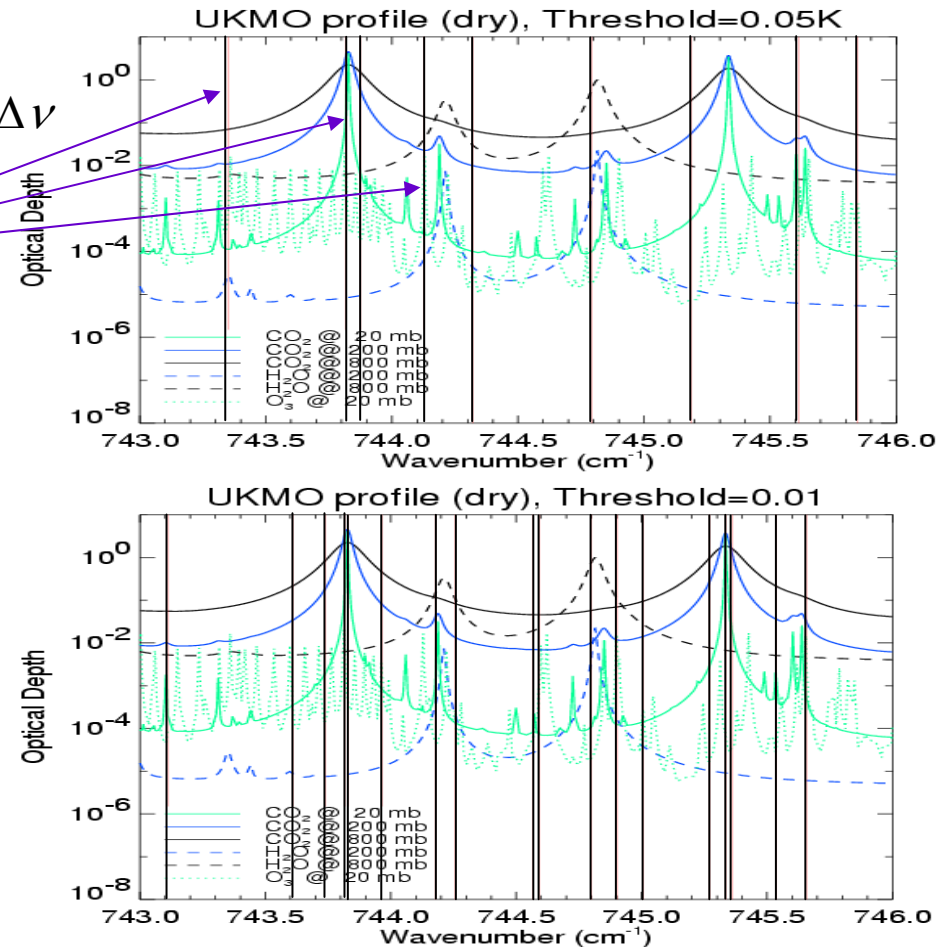
- OSS overview
 - Overview of the approach
 - Forward model
 - General attributes
- OSS/OPTRAN comparison
- Generalized training
 - Clear/cloudy training
 - Inversion issue
- Treatment of multiple scattering
 - Validation against CHARTS
 - Application to AIRS
- Summary/future work

Overview of the OSS approach

- OSS method (Moncet *et al.* 2003, 2001) models the channel radiance as

$$\bar{R} = \int_{\Delta\nu} \phi(\nu)R(\nu)d\nu \cong \sum_{i=1}^N w_i R(\nu_i); \quad \nu_i \in \Delta\nu$$

- Wavenumber ν_i (nodes) and weights w_i are determined by fitting "exact" calculations (from line-by-line model) for globally representative set of atmospheres (training set)
- Radiance training is fast and provides mechanism for directly including slowly varying functions (e.g. Planck, surface emissivity) in the selection process



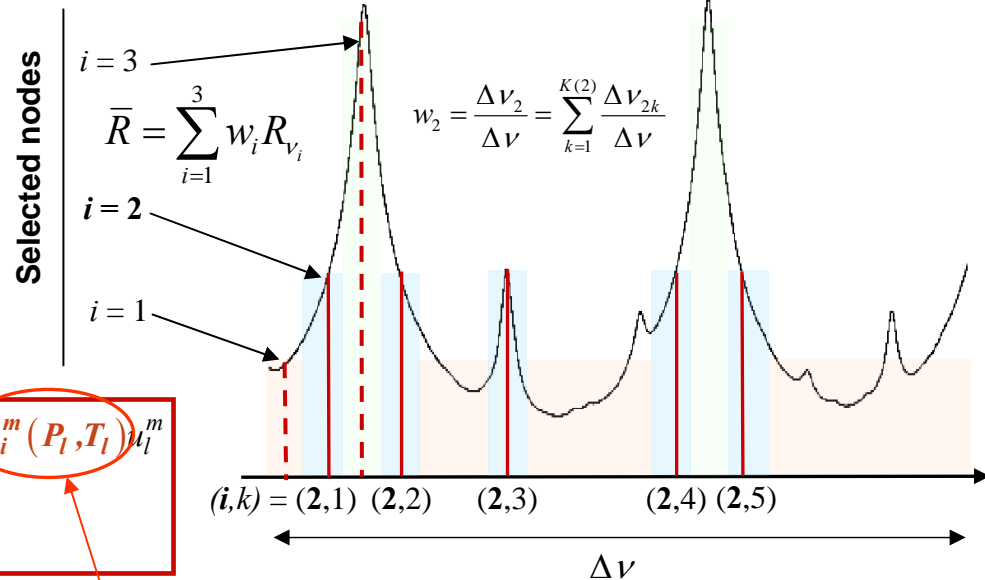
ESFT (Wiscombe and Evans, 1977) for single layer, single absorber case:

$$\bar{\tau}(u) = \int_{\Delta\nu} e^{-k_\nu u} d\nu \approx \sum w_i e^{-k_i u}$$

Extension to multiple absorbers along inhomogeneous path (e.g. Armbruster and Fisher, 1996)

$$\bar{\tau}(p) = \int_{\Delta\nu} \tau_\nu(p) d\nu \approx \sum w_i e^{-\sum_l \sum_m k_i^m(P_l, T_l) u_l^m}$$

OSS solution:
$$\bar{\tau}(p) \approx \sum w_i e^{-\sum_l \sum_m k_{\nu_i}^m(P_l, T_l) u_l}$$

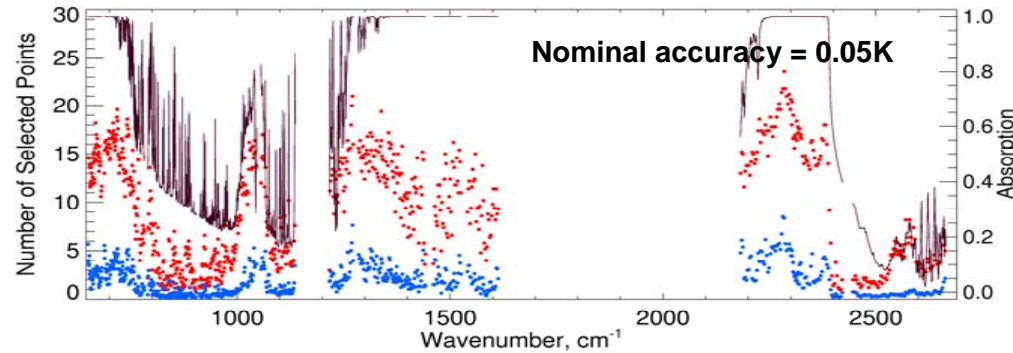


Extension of ESFT to inhomogeneous atmospheres with multiple absorbers reduces the problem to a single (wavenumber) dimension and ensures that the solution is physical

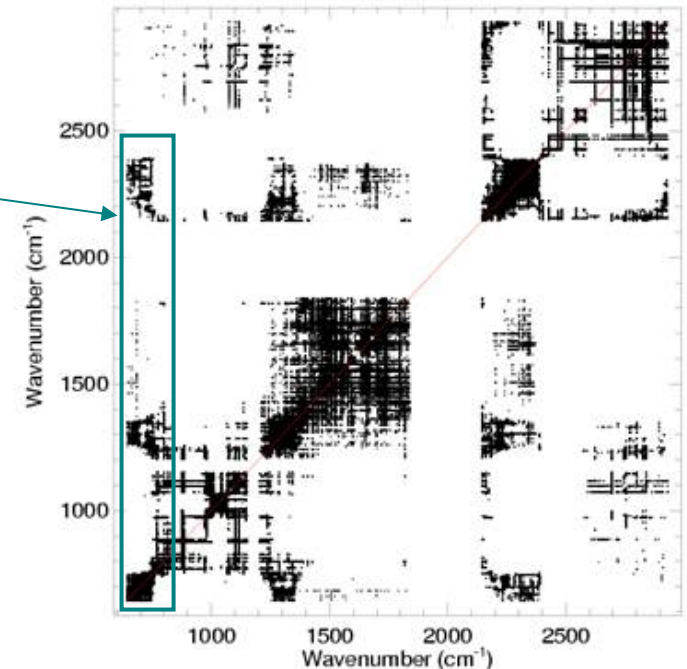
Localized versus non-localized training

- Localized training (*reference*) operates on individual channels, one at a time – node redundancy due to overlapping ILS

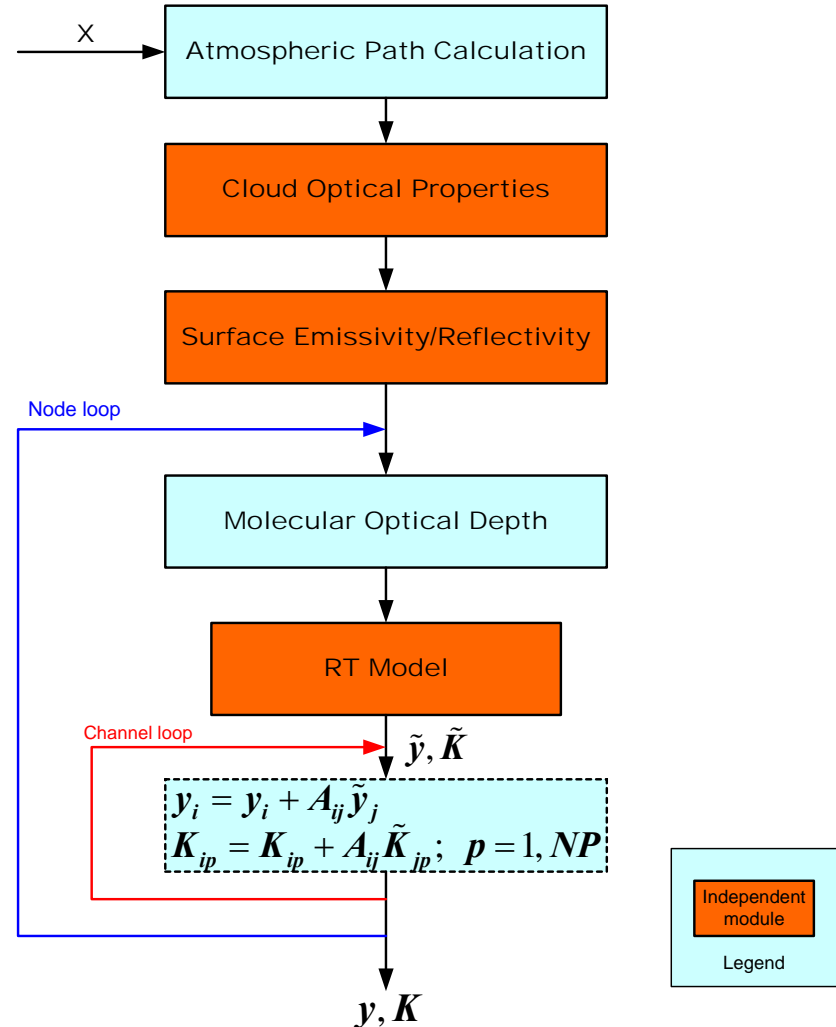
- AIRS (2378 channels):
 - Average # nodes per channel: ~9 nodes/channel
 - Total number of nodes/number of channel (i.e. no redundancy) = 1.9 nodes/channel



- Non-localized training operates on groups of N channels (up to full channel set)
 - Exploits node-to-node correlation to minimize total number of nodes across a spectral domain (*regression!!!*)
 - Results in significant increase in number of points in any given channel increases
 - Critical for MODTRAN (range 0-50,000 cm^{-1})



- **RTM structure**
 - Main loop is the node loop
 - Internal channel loop to update channel radiance and Jacobians
 - Similar structure adopted for CRTM
- **LUT of $kabs$ stored for all relevant molecules as a function of temperature**
 - Self broadening included for water vapor
 - Maximum brightness temperature error with current LUT < 0.05K in infrared and < ~0.01K in microwave
- **Use simple monochromatic RT model (clear or scattering)**
 - Jacobians (required for retrieval applications) are straightforward in the clear-sky (e.g. CrIS ATBD)



- RT model designed to handle any number of variable trace species
- Adding a new variable species requires no change in OSS parameterization
 - No change in RTM required
 - Only need to include variability in training (number of nodes may increase as a result)
- # of variable trace gases and molecule type specified on node-by-node basis (*set by the user at run time*)
 - Average number of trace gases absorbing at any given frequency \ll total number of absorbing species in the atmosphere
 - Computationally efficient and minimizes memory requirements
 - Inexpensive Jacobian computation:
$$\frac{\partial y}{\partial \mathbf{u}_l^m} = \frac{\partial y}{\partial \tau_{abs,l}^0} k_l^m$$

Optimal Spectral Sampling (OSS) method

- OSS absorption parameterization leads to *fast* and *numerically accurate* RT modeling:
 - OSS-based RT model can approach line-by-line calculations arbitrarily closely
 - Adjustable numerical accuracy:
 - Possibility of trade off between accuracy and speed
 - Unsupervised training
 - No empirical adjustment: cuts significantly on cost of testing approximations and validating model
 - Provides flexible handling of (variable) trace molecular species
 - Designed to handle large number of variable trace species w/o any change to model – low impact on computational cost
 - Selection of variable trace gases at run time
 - Memory requirements do not change whether we are dealing with one or more instruments
 - Execution speed primarily driven by total spectral coverage and maximum spectral resolution (not by number of instruments)
 - Leads to accurate handling of multiple scattering (cloudy radiance assimilation)

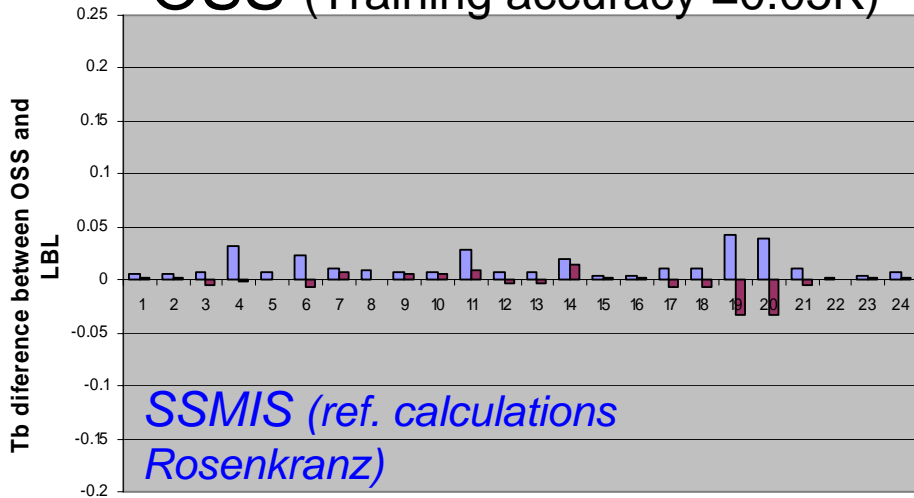
- Used in NPOESS/ CrIS, CMIS and OMPS (IR) retrieval algorithms
- JCSDA CRTM
 - Compared with OPTRAN at NOAA for AMSU, SSMIS, HIRS-3, GOES imager/sounder, AIRS
 - Accuracy and timing
 - Beta version of OSS-based CRTM about to be tested at NCEP to evaluate impact on forecast
 - Other comparison results from ITSC comparison (Garand et al. 2001), and recent ITSC AIRS comparison (Saunders et al., 2005)
- Currently working on integrating into MODTRAN (AFRL- sponsored effort)
 - Wide array of users and applications
 - Same method should cover it all
- NASA's Mars Fundamental Research Program: OSS forward model has been developed for the Thermal Emission Spectrometer (TES) onboard the Mars Global Surveyor spacecraft (Christensen et al. 2001).

JCSDA OPTRAN/OSS (localized training) comparison

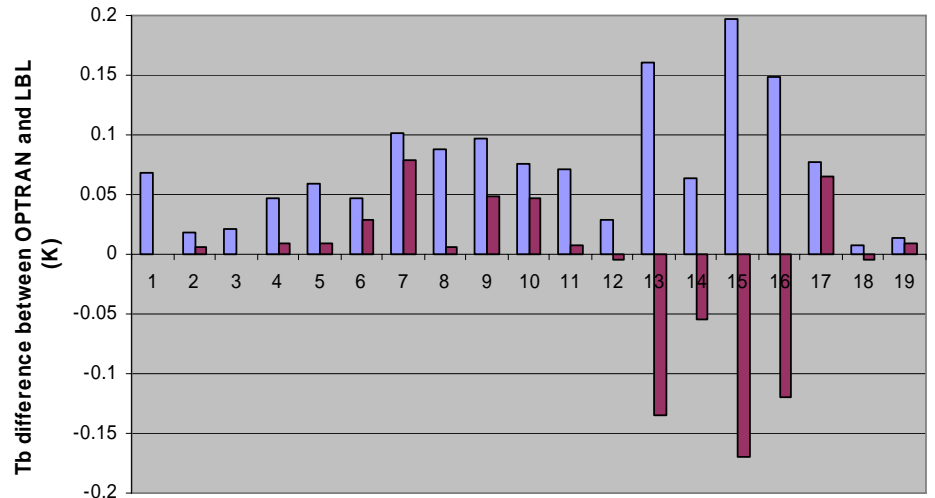
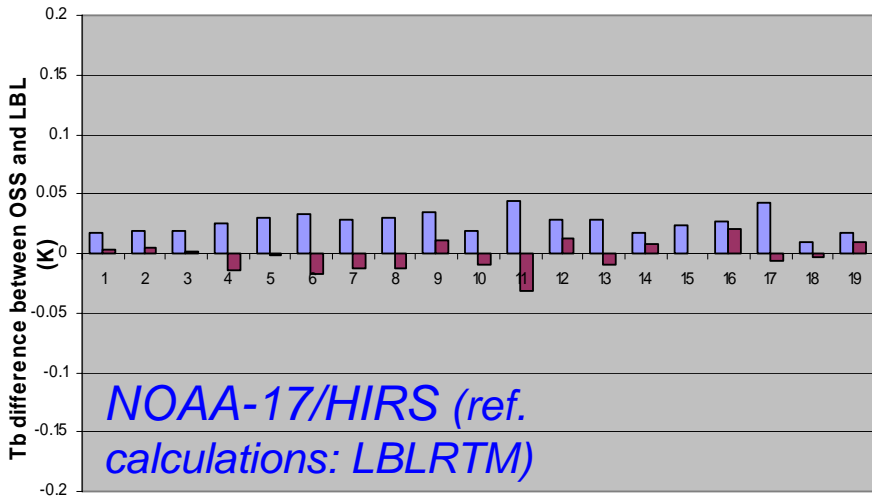
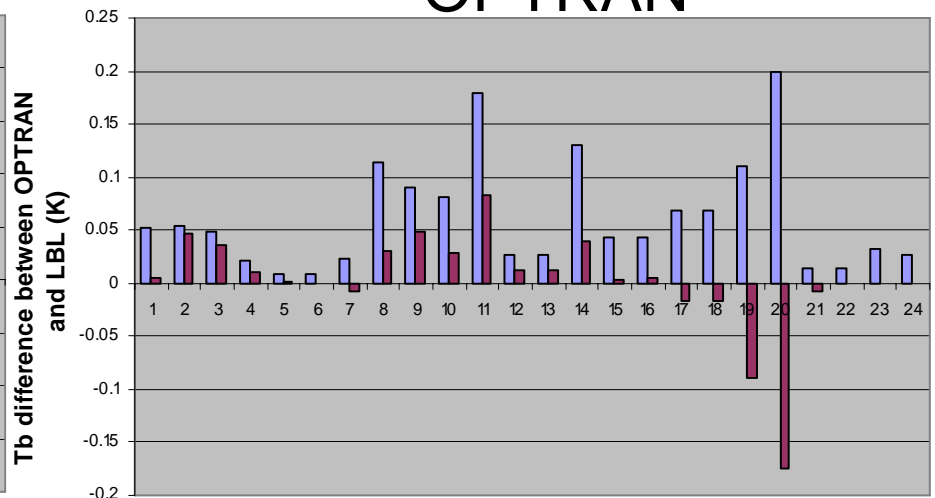
OPTRAN/OSS comparison: SSMIS & HIRS

(from NOAA, 2005)

OSS (Training accuracy = 0.05K)



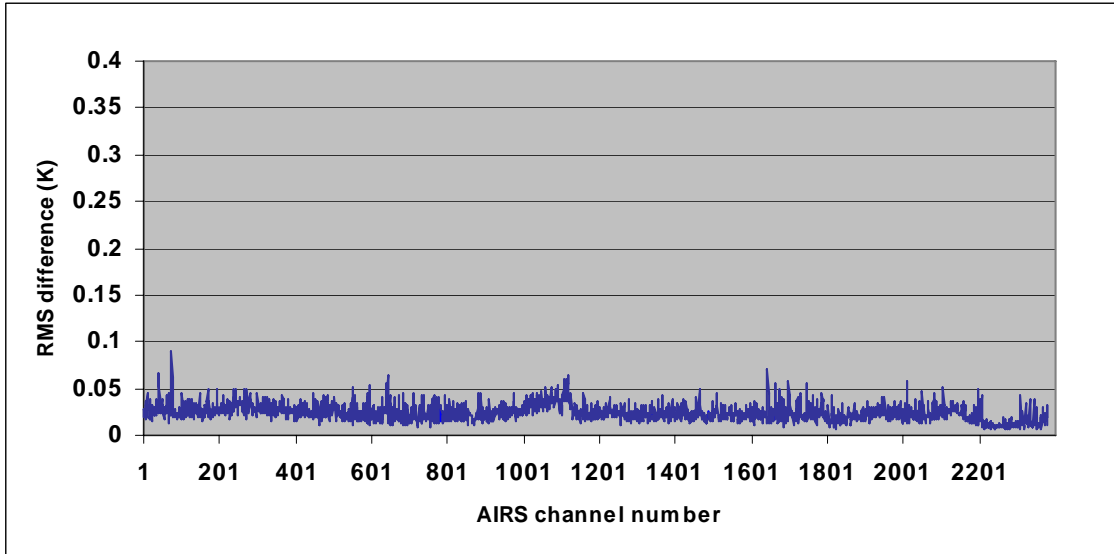
OPTRAN



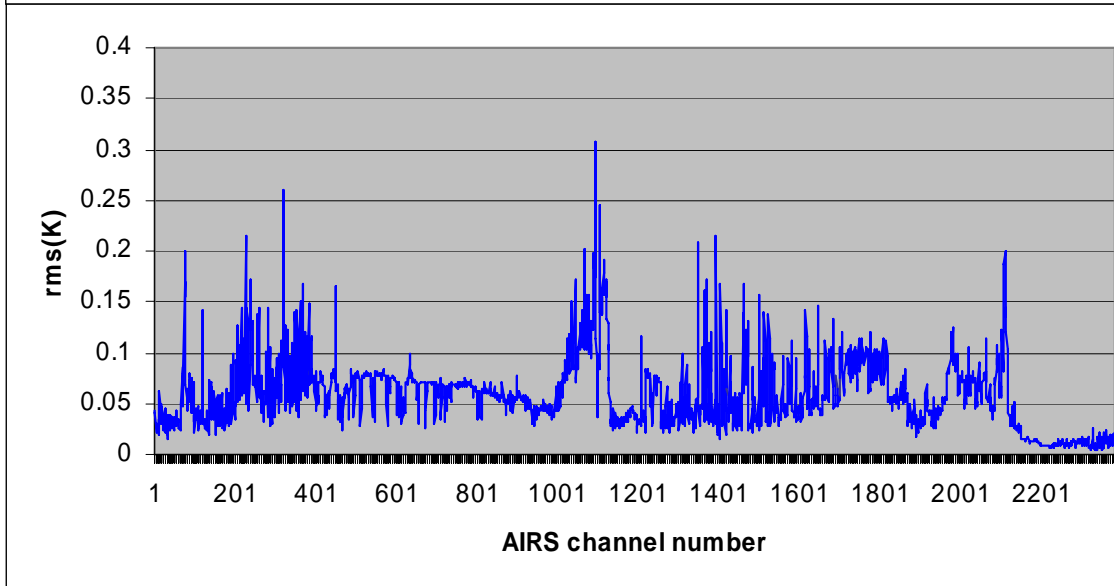
█ RMS difference
█ Mean difference

OPTRAN/OSS comparison: AIRS

(from NOAA, 2005)



OSS
Trained with ECMWF set
Tested with UMBC set
(Training accuracy = 0.05K)



OPTRAN
Trained with UMBC set
Tested with ECMWF set

OPTRAN/OSS Comparison: Computation & Memory Efficiency (from NOAA, 2005)

Time needed to process 48 profiles with 7 observation angles (336 profiles)

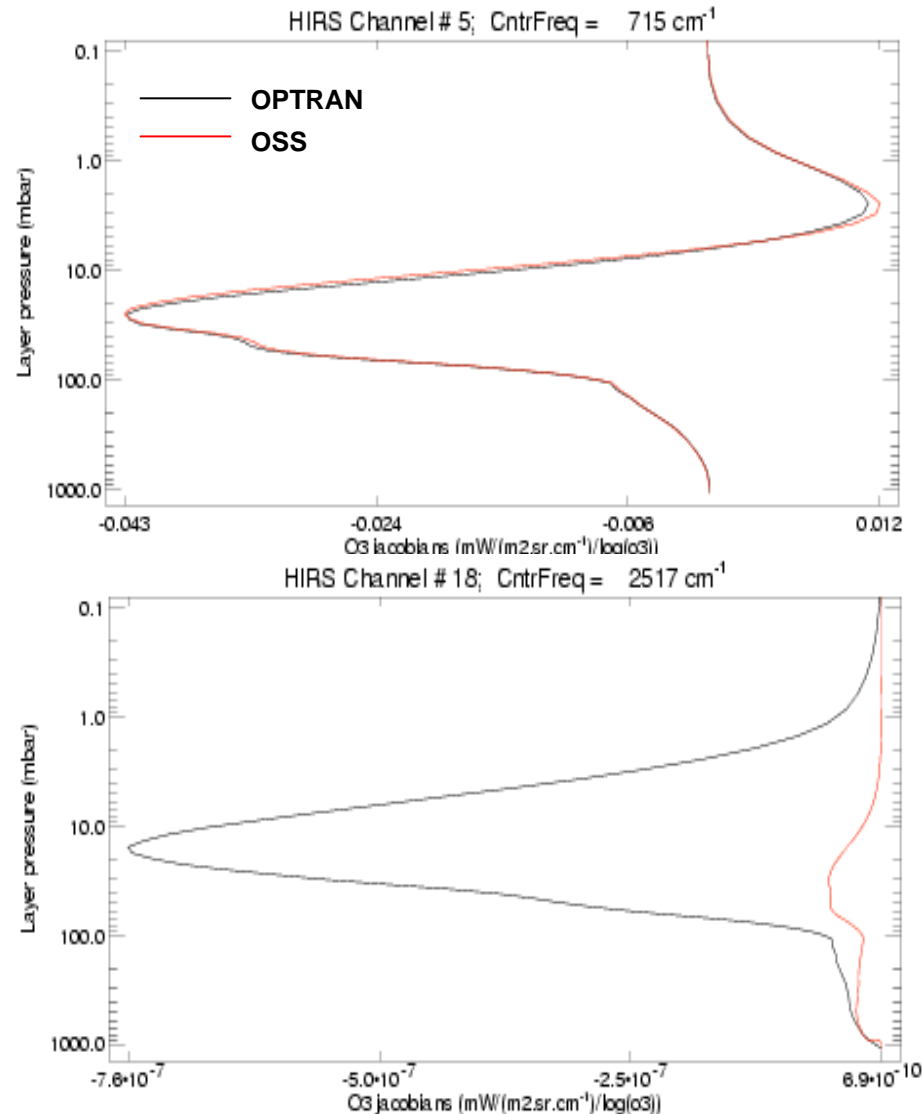
	OPTRAN-V7 Forward, Jacobian + Forward	OPTRAN-comp Forward, Jacobian + Forward	OSS Jacobian + Forward
AIRS	7m20s, 22m36s	10m33s, 35m12s	3m10s
HIRS	4s, 13s	5s, 17s	9s

Memory resource required (Megabytes)

	OPTRAN-V7 single, double precision	OPTRAN-comp double precision	OSS
AIRS	33, 66	5	97**
HIRS	0.26, 0.5	0.04	4

****With OSS: Based on 0.05K accuracy -
No increase in size when adding other IR instruments**

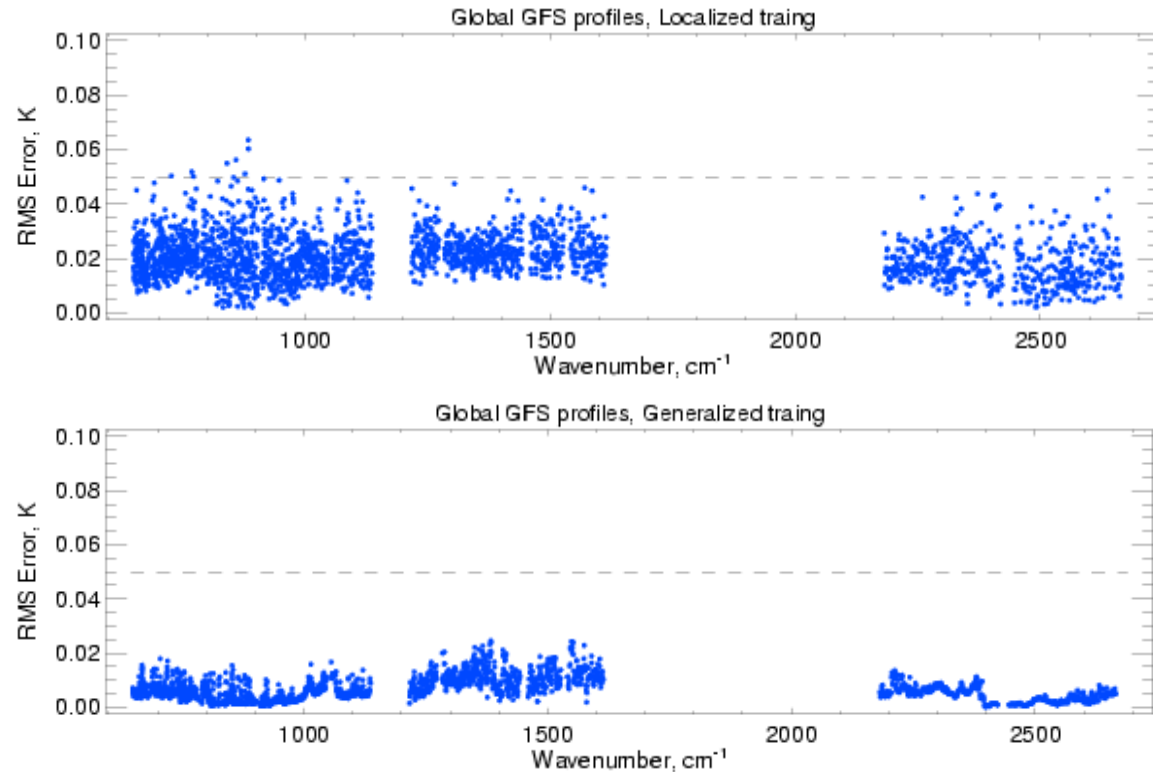
- OSS Jacobian accuracy commensurate with model accuracy
 - Unlike OPTRAN (trained to fit transmittances for individual absorbers), OSS fits total transmittance/radiance (OPTRAN equivalent training obtained by zeroing out major absorber concentration)
 - Jacobians for weakly absorbing constituents not as accurate when impact on radiances of (global) variability in concentration is less than model accuracy



Generalized training

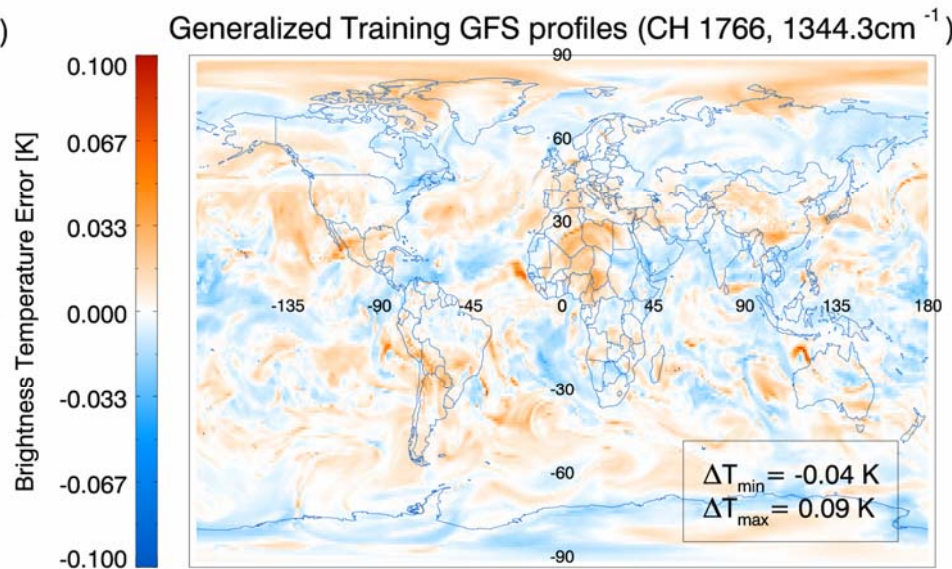
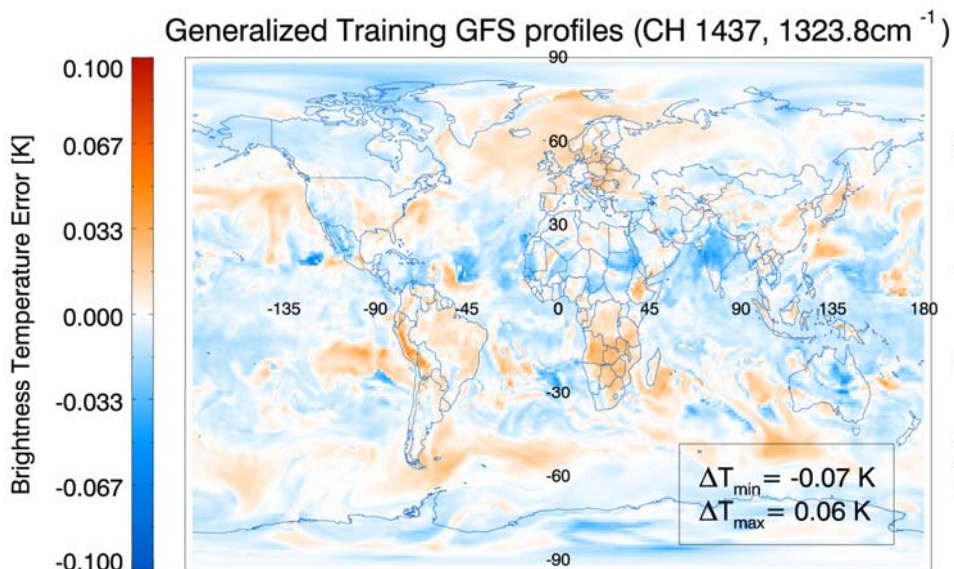
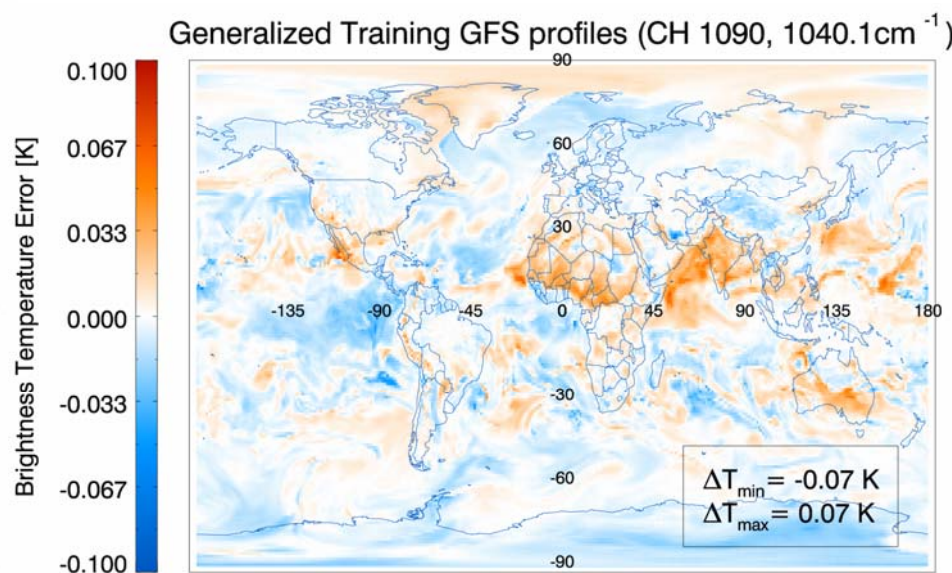
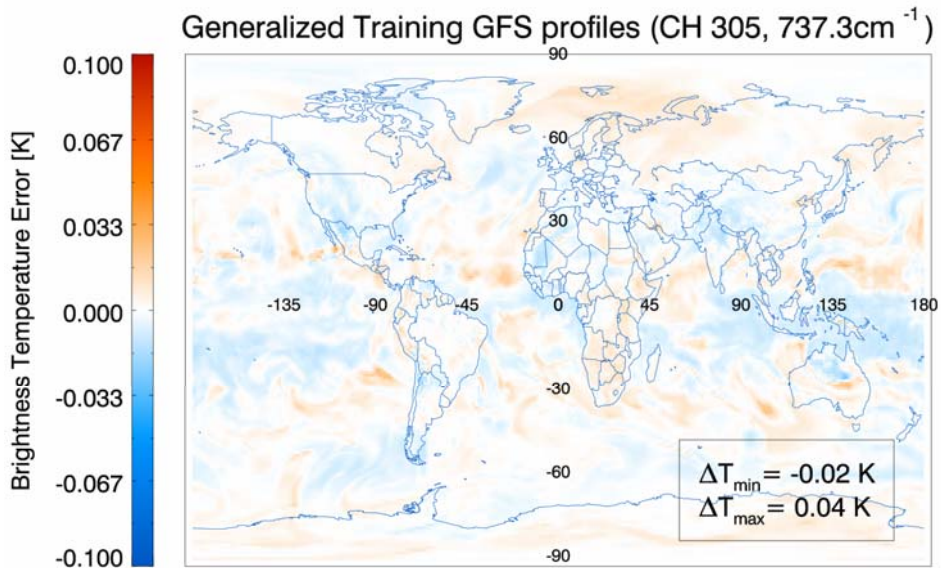
Performance example (AIRS)

- Localized training (0.05K accuracy):
 - ~2nodes /channel
 - ~5000 monochromatic calculations for full AIRS channel set
- Generalized training:
 - ~0.1 node/channel
 - Reduces number of monochromatic calculations to ~250



Speed gain ~ 20 compared to localized training for AIRS

Examples of error spatial distribution



Non-localized cloudy training

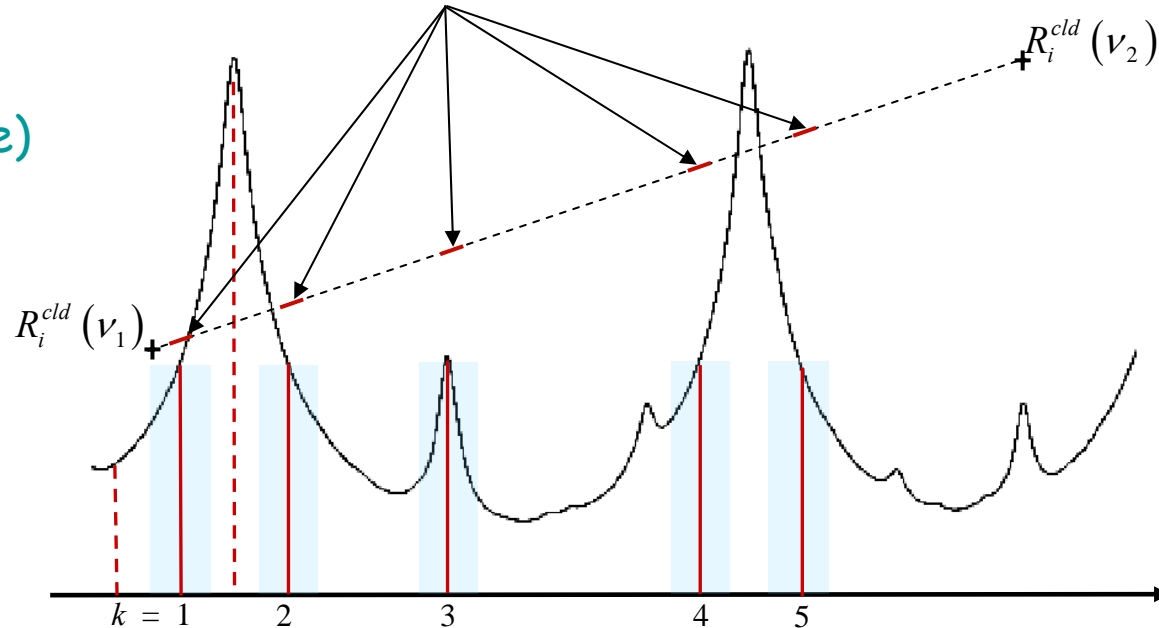
- Must include slowly varying cloud/aerosol optical properties in training
 - Over wide bands: training can be done by using a database of cloud/aerosol optical properties
 - More general training obtained by breaking spectrum in intervals of the order of 10 cm^{-1} in width (impact of variations in cloud/aerosol properties on radiances is quasi-linear) and by performing independent training for each interval
 - lower computational gain but increased robustness
- Direct cloudy radiance training not recommended !
 - Clouds tend to mask molecular structure which makes training easier
 - If "recipe" for mixture of clear and cloudy atmospheres in direct training not right: clear-sky performance degrades

Single/multi-channel cloudy training over wide spectral domains

Alternate two-step training preserves clear-sky solution

- **First step:**
normal clear-sky
(transmittance/radiance)
training to model
molecular absorption
- **Second step:**
duplicate + redistribute
nodes across spectral
domain and recompute
weights to incorporate
slowly varying functions
into the model

$$R_i^{cld}(v_k) = a_{ik} R_i^{cld}(v_1) + (1 - a_{ik}) R_i^{cld}(v_2)$$



$$\bar{R} = \sum_i w_i \sum_k (a_{ik} R_i(v_1) + (1 - a_{ik}) R_i(v_2)) \frac{\Delta v_{ik}}{\Delta v_i} = \sum_i w'_i R_i(v_1) + (w_i - w'_i) R_i(v_2)$$

Robust, physical approach for including slowly varying functions (e.g. cloud optical properties, surface emissivity) into OSS formalism

Cloudy training preserves clear-sky solution

Variational retrieval methods:

- Average channel uses ~150 nodes
- Mapping Jacobians from node to channel space partially offsets speed gain

$$\delta x_{n+1} = (\mathbf{K}_n^T \mathbf{S}_\varepsilon^{-1} \mathbf{K}_n + \mathbf{S}_x^{-1}) \mathbf{K}_n^T \mathbf{S}_\varepsilon^{-1} \left[(y_n - y^m) + \mathbf{K}_n \delta x_n \right],$$

where,

$$\mathbf{y} = \mathbf{A} \tilde{\mathbf{y}} \text{ and}$$

$$\mathbf{K} = \mathbf{A} \tilde{\mathbf{K}}$$

Alternatives:

- PC (reduces first dimension of matrix A)
- Operate directly in node space

$$\mathbf{y}^m = \mathbf{A} \tilde{\mathbf{y}}^m \rightarrow \hat{\mathbf{y}}^m = \mathbf{H} \mathbf{y}^m$$

- Avoids Jacobians transformation all together and reduce K-matrix size (inversion speed up)

- for AIRS: 2378 channels -> 250 nodes

$$\tilde{\mathbf{y}}^m = (\mathbf{A}^T \mathbf{S}_\varepsilon^{-1} \mathbf{A})^{-1} \mathbf{A} \mathbf{S}_\varepsilon^{-1} \mathbf{y}^m$$

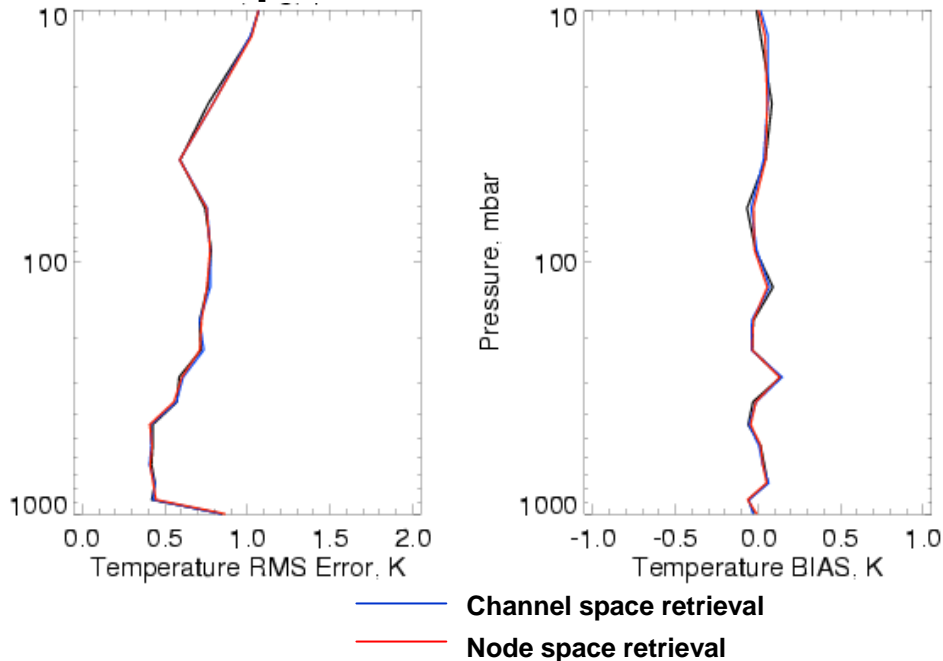
$$\tilde{\mathbf{S}}_\varepsilon^{-1} = \mathbf{A}^T \mathbf{S}_\varepsilon^{-1} \mathbf{A}$$

$$\delta \mathbf{x}_{n+1} = (\tilde{\mathbf{K}}_n^T \tilde{\mathbf{S}}_\varepsilon^{-1} \tilde{\mathbf{K}}_n + \mathbf{S}_x^{-1}) \tilde{\mathbf{K}}_n^T \tilde{\mathbf{S}}_\varepsilon^{-1} \left[(\tilde{\mathbf{y}}_n - \tilde{\mathbf{y}}^m) + \tilde{\mathbf{K}}_n \delta \mathbf{x}_n \right]**$$

**Equivalent to

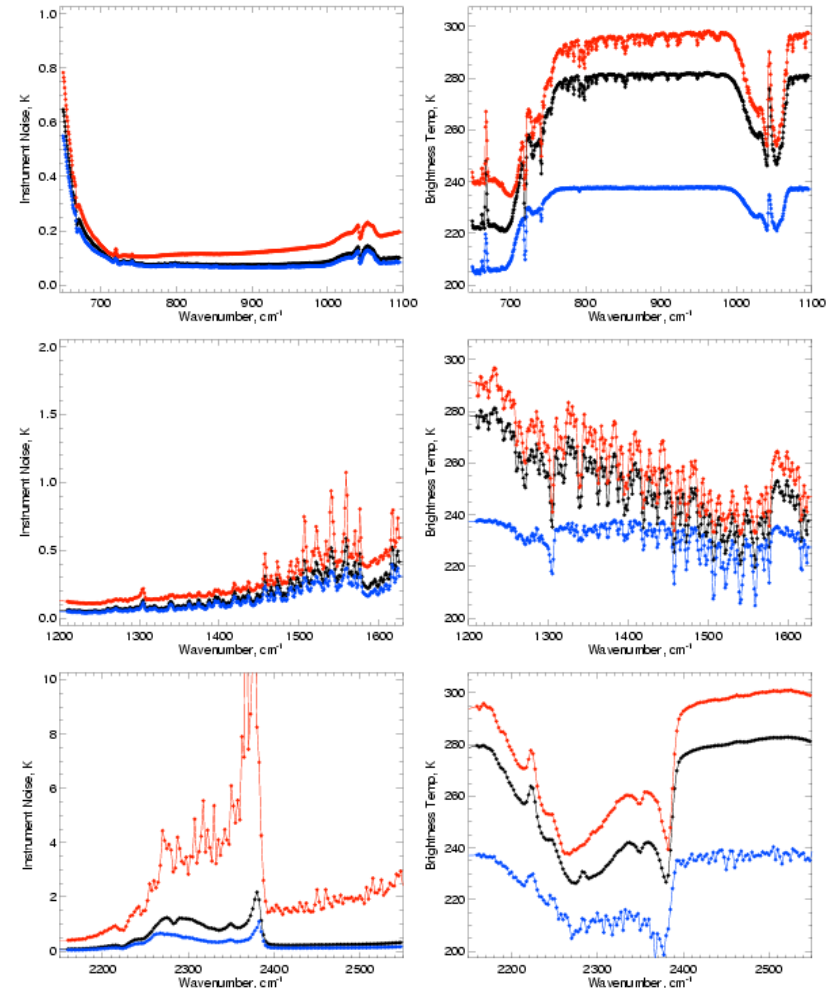
$$\delta \mathbf{x}_{n+1} = (\mathbf{K}_n^T \mathbf{S}_\varepsilon^{-1} \mathbf{K}_n + \mathbf{S}_x^{-1}) \mathbf{K}_n^T \mathbf{S}_\varepsilon^{-1} \left[(\mathbf{A} \mathbf{H} \mathbf{y}_n - \tilde{\mathbf{y}}^m) + \mathbf{K}_n \delta \mathbf{x}_n \right]$$

Retrieval performance – constant noise



- Need strategy for handling input dependent noise
 - Scene temperature dependence (clear/cloudy)
 - worse in SW band
 - Cloud clearing noise amplification
- H-transformation not overly sensitive to noise
 - For clear retrievals: sufficient to update noise covariance regionally

Example of IR sounder noise characteristics (clear sky)



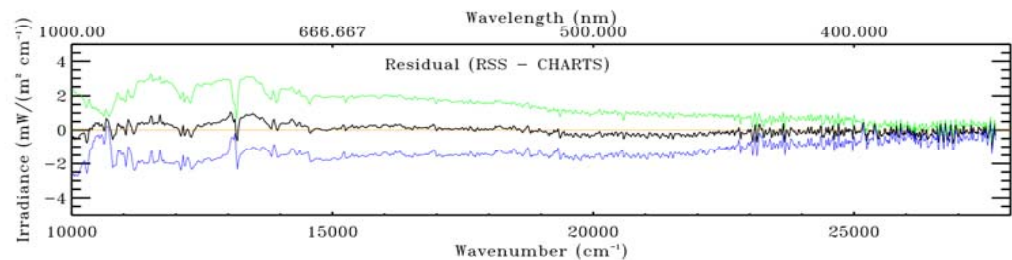
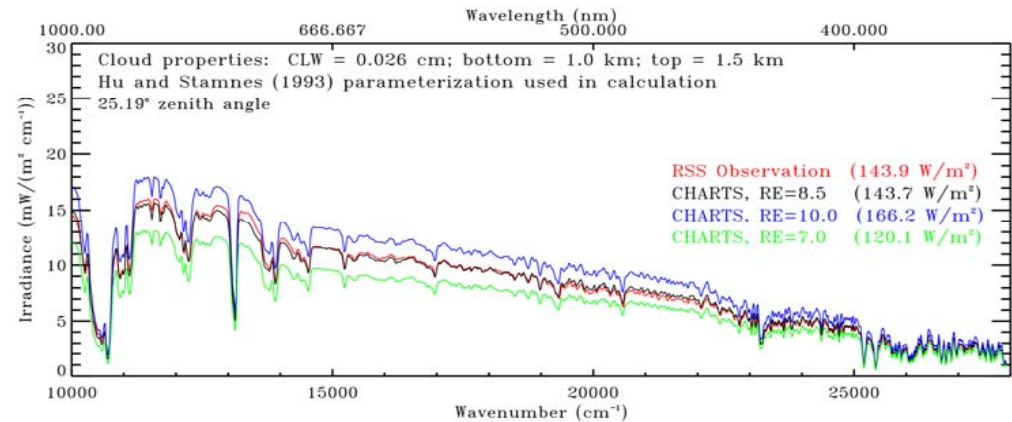
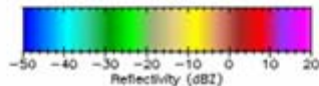
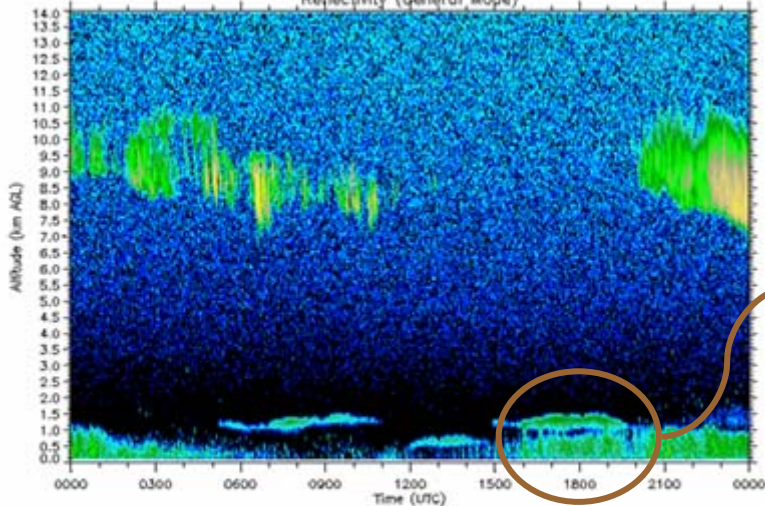
Application to Scattering Atmospheres

CHARTS (Moncet and Clough, 1997):

- Fast adding-doubling scheme for use with LBLRTM
 - Uses tables of layer reflection/transmittance as a function of total absorption computed at run time
- Validation against measurements from Rotating Shadowband Spectroradiometer (RSS) spectra at the ARM/SGP site

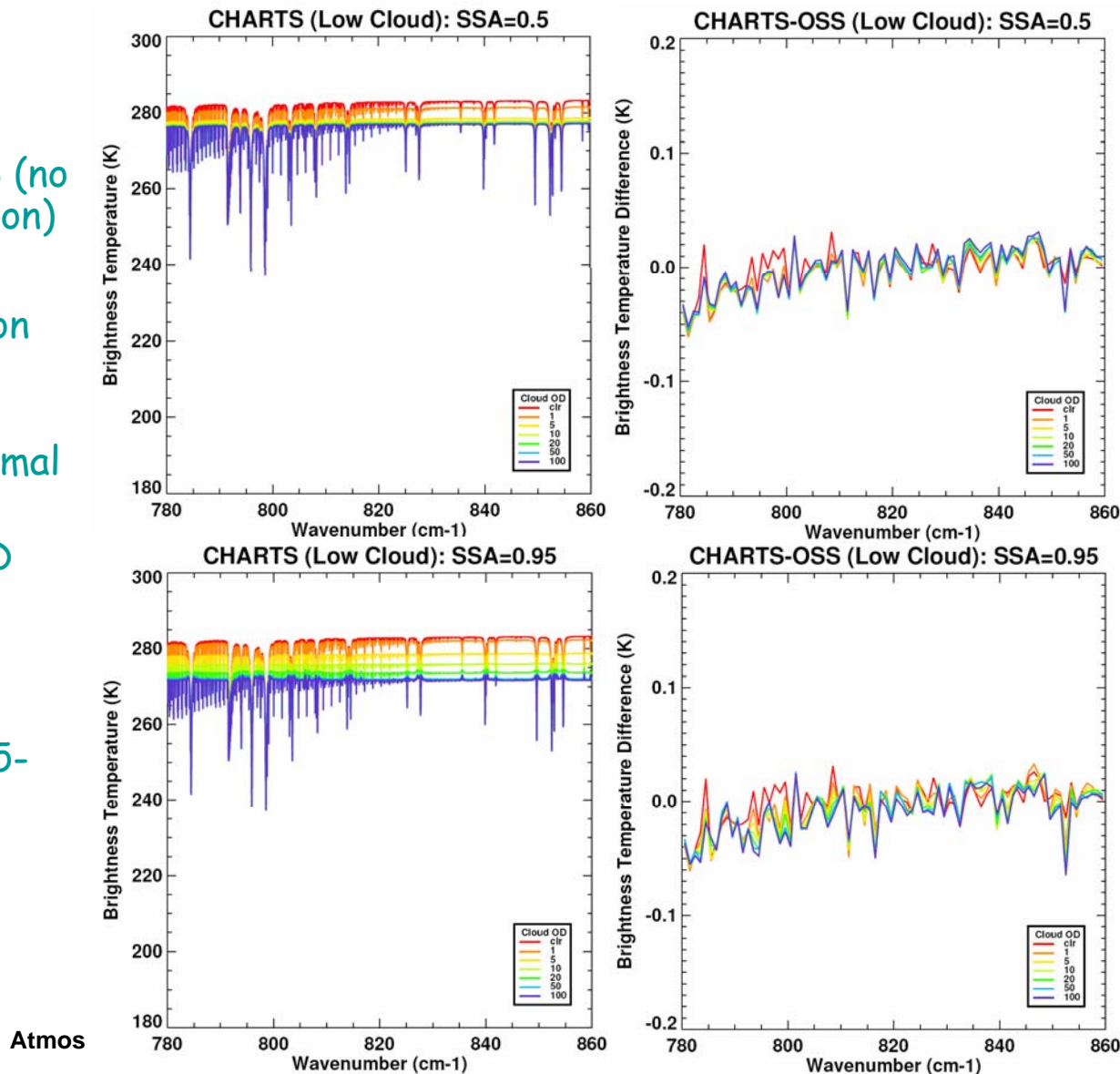
MMCR Reflectivity Data
19 May 2000

Reflectivity (General Mode)



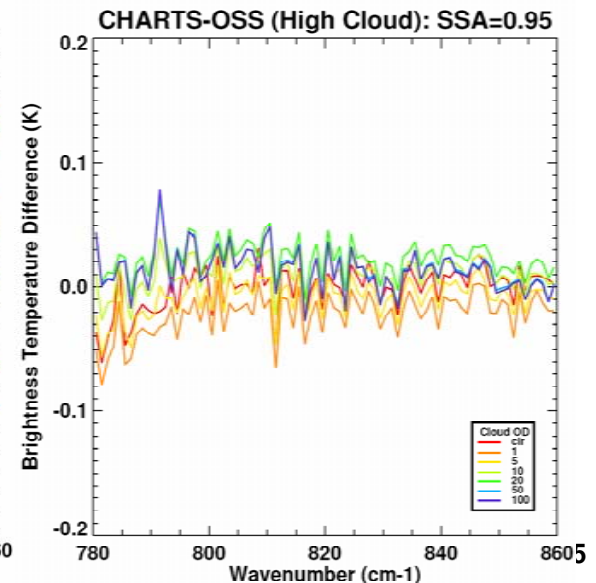
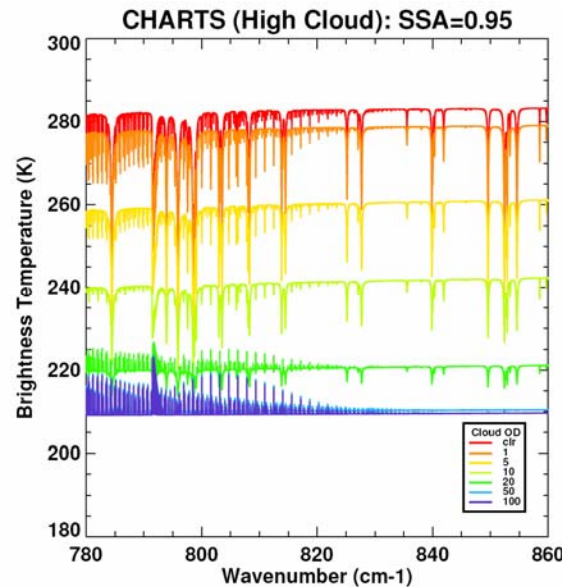
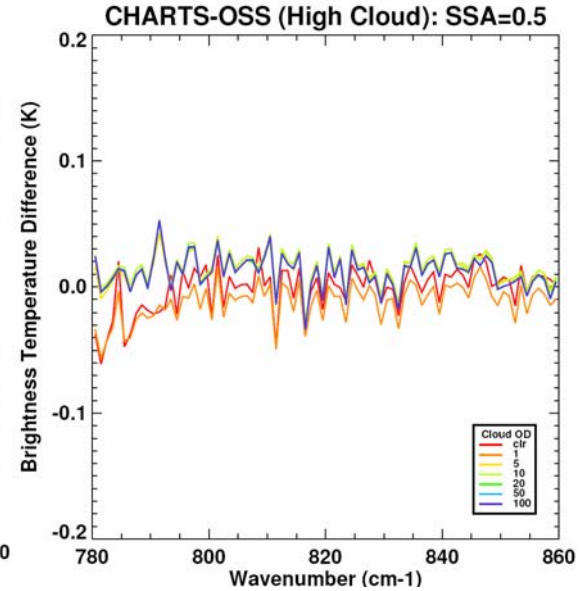
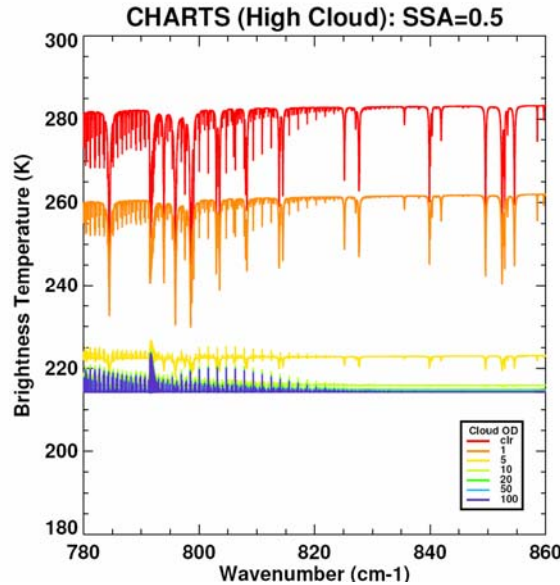
OSS/CHARTS Comparison (2)

- OSSSCAT:
 - Single wavelength version of CHARTS (no spectral interpolation)
- Cloudy validation:
 - Molecular absorption from 740-900 cm^{-1} domain
 - 1cm^{-1} boxcars, thermal only
 - Cloud extinction OD range: 0-100
- Example:
 - 780-860 cm^{-1}
 - Low cloud case (925-825 mb)



OSS/CHARTS Comparison (3)

- Same as previous
- High cloud case (300-200 mb)



Clear sky training adequate in thermal regime

Refinement in training needed for thick clouds (OD > 50) when SSA approaches 1 and high scan angles

Atmos

Cirrus cloud microphysics parameterization

- Size distribution is strongly bi-modal

- Mid-latitude cirrus

- Small mode:

- fixed shape recipe (16% bullet-rosettes, 31% planar polycrystal / irregular – 53% quasi-spherical)

- Large mode:

- Temperature-dependent shape recipe

- Tropical cirrus

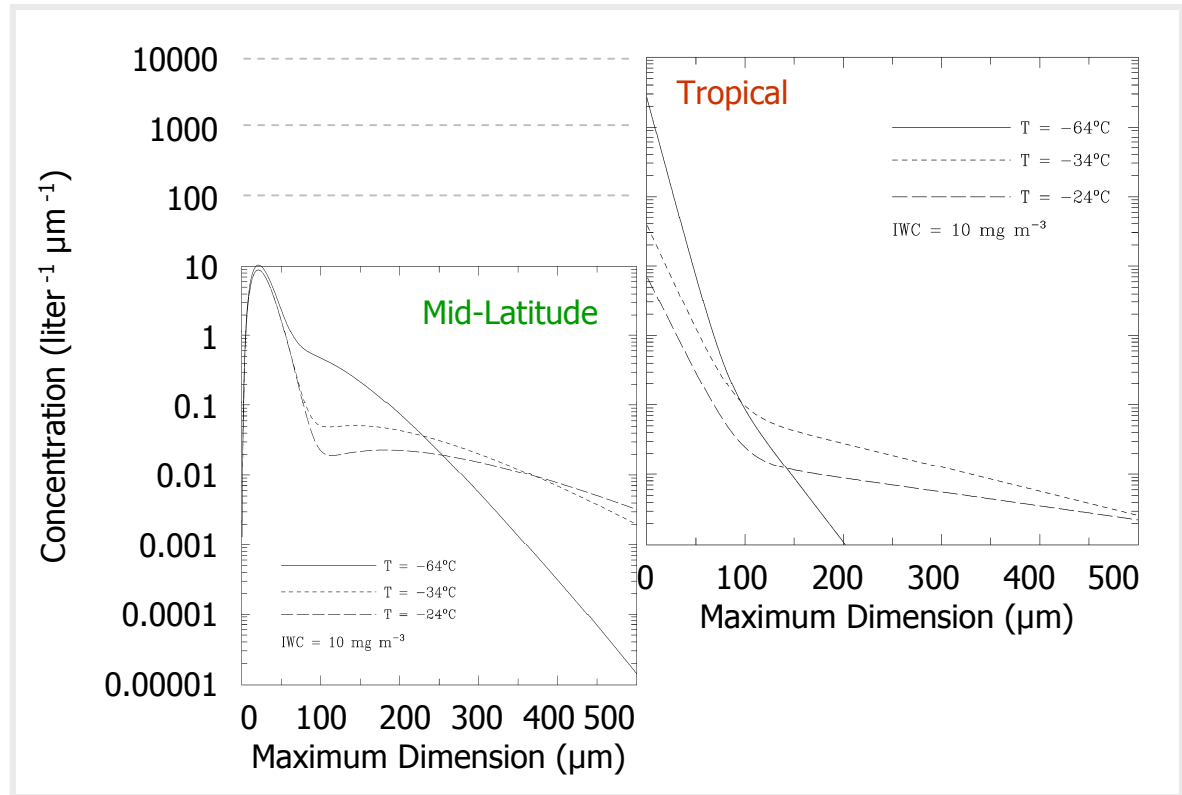
- Small mode:

- 40% planar polycrystals, 60% quasi-spherical

- Large mode:

- 30% hexagonal, 65% planar poly-crystals, 5% hexagonal plates

- Strong temperature dependence of size distribution shape

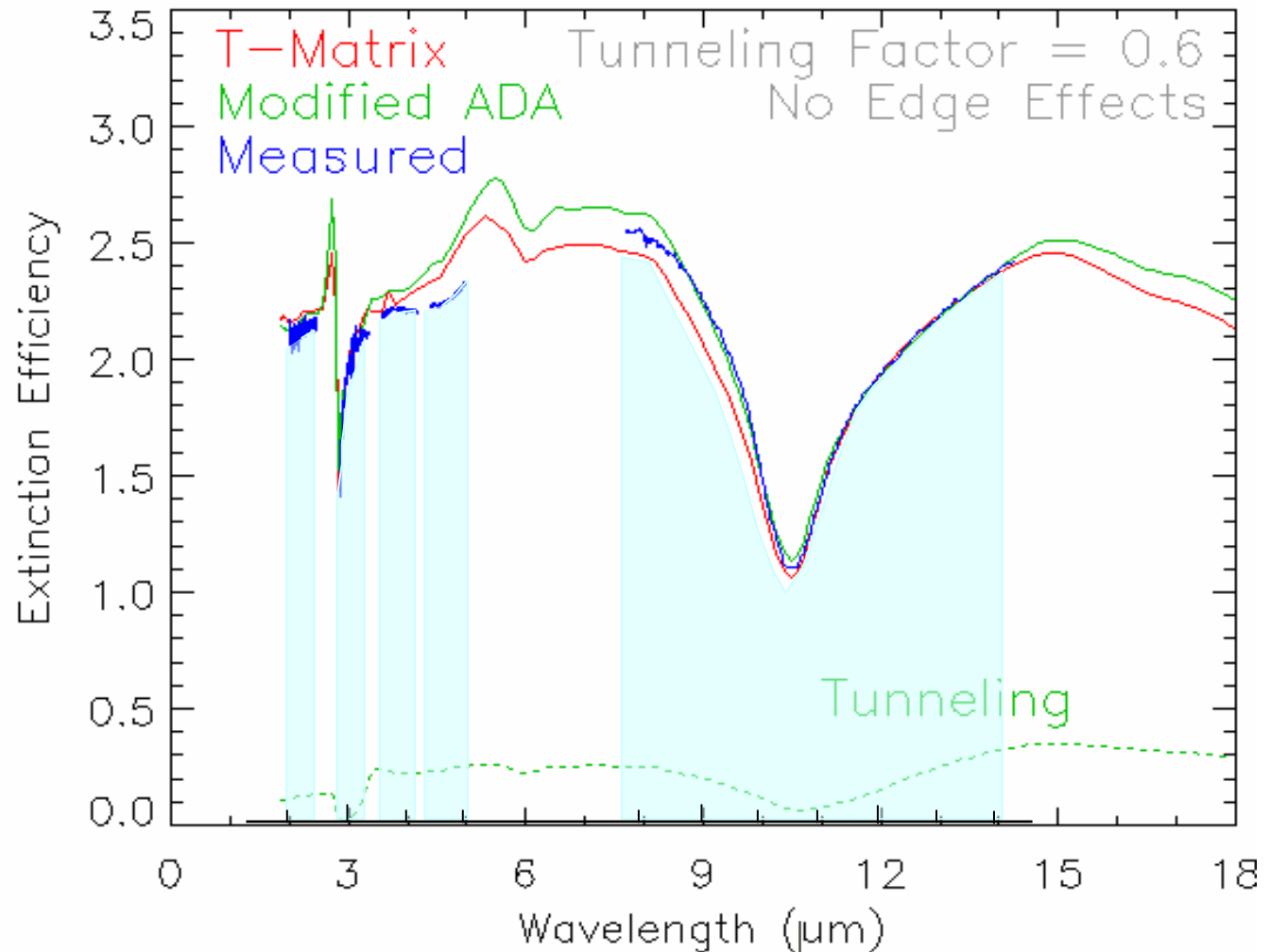


Optical properties from Modified Anomalous Diffraction Approximation (MADA)

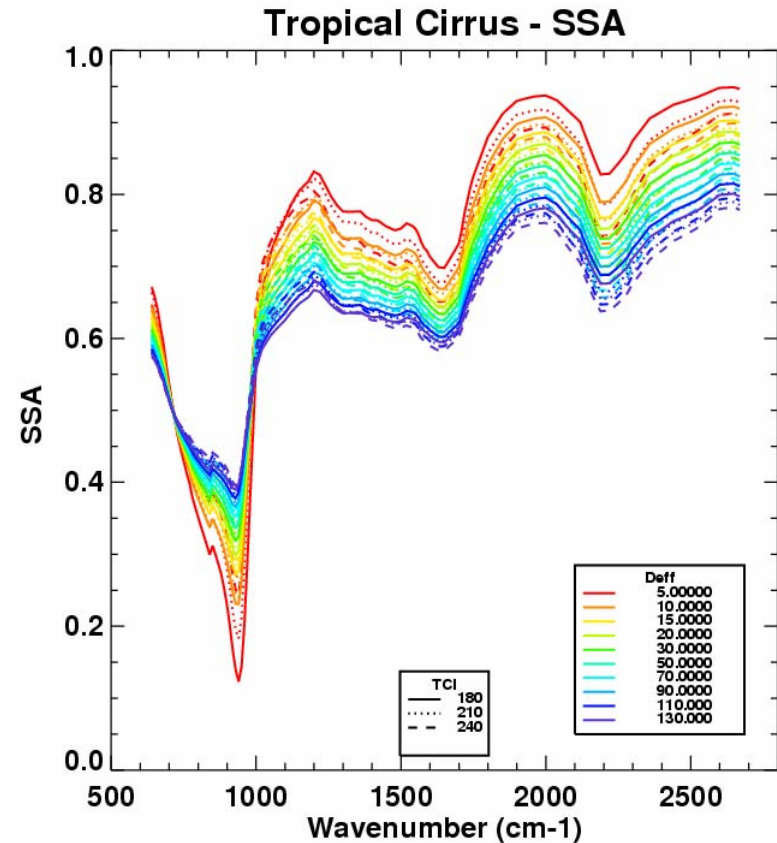
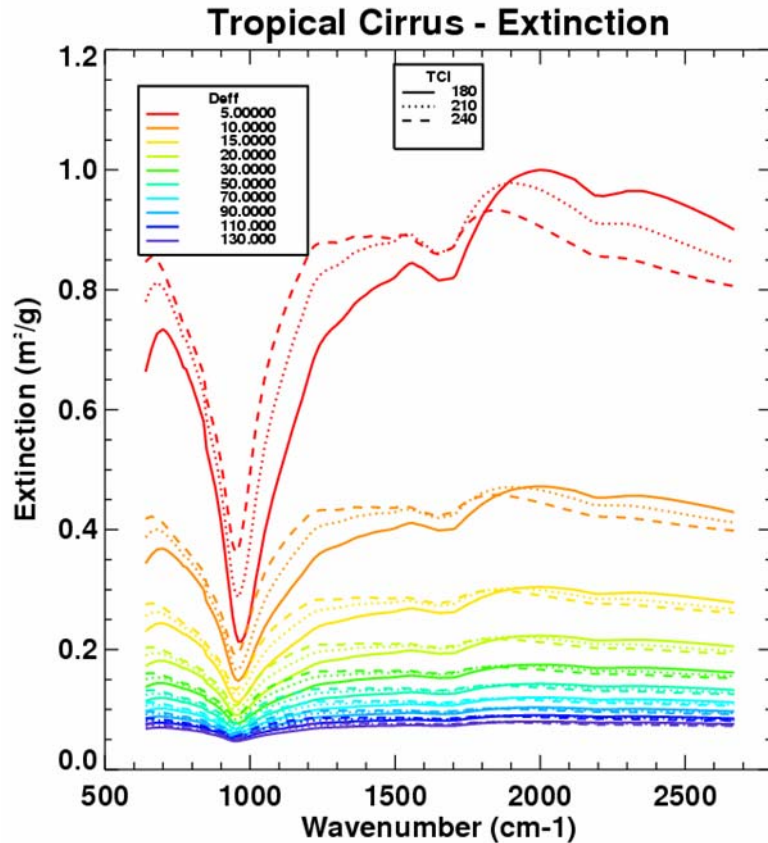
Comparing T-Matrix, MADA, and Measured Qext

- Testing MADA against cloud chamber Q_{ext} measurements and against T-matrix theory using observed size distribution
- Effective diameter was 14 microns

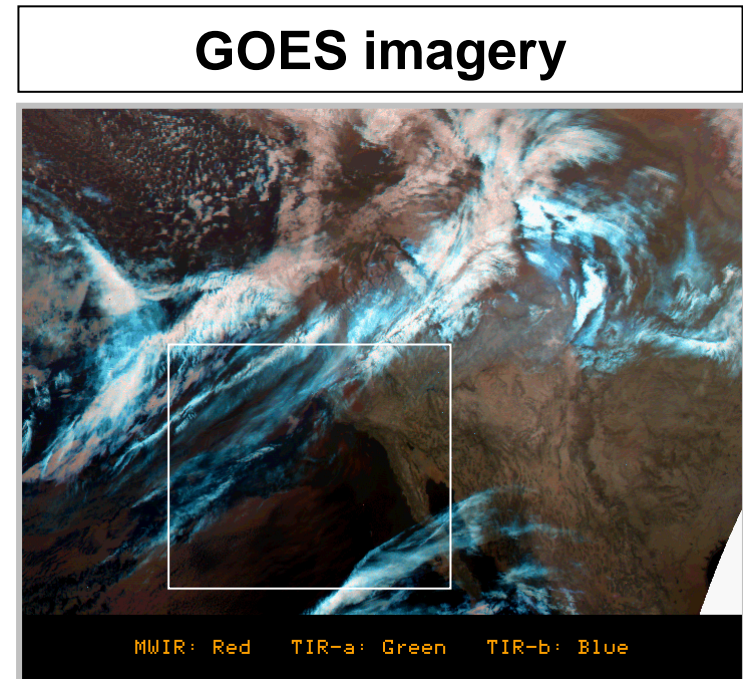
References:
(Mitchell, 2000,
2002)



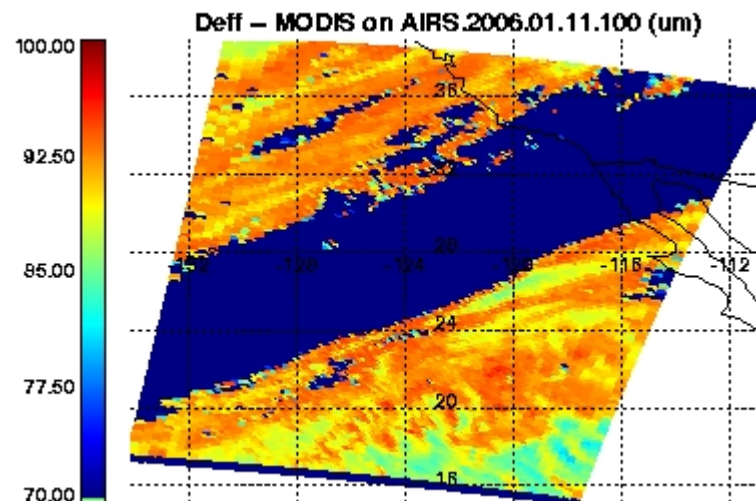
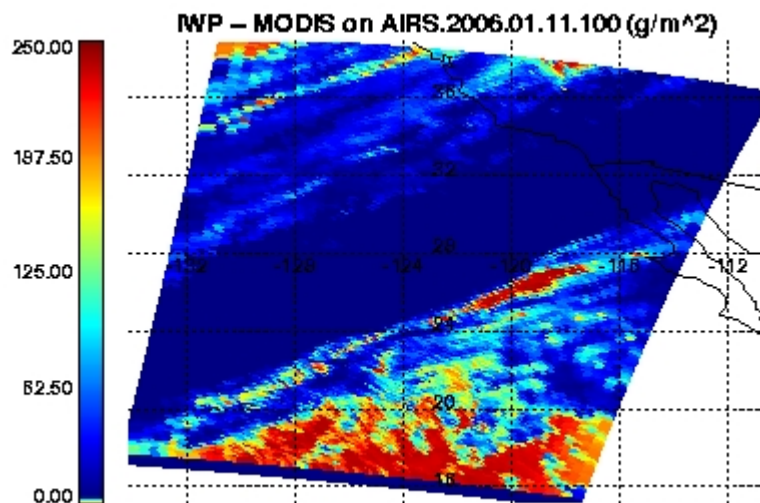
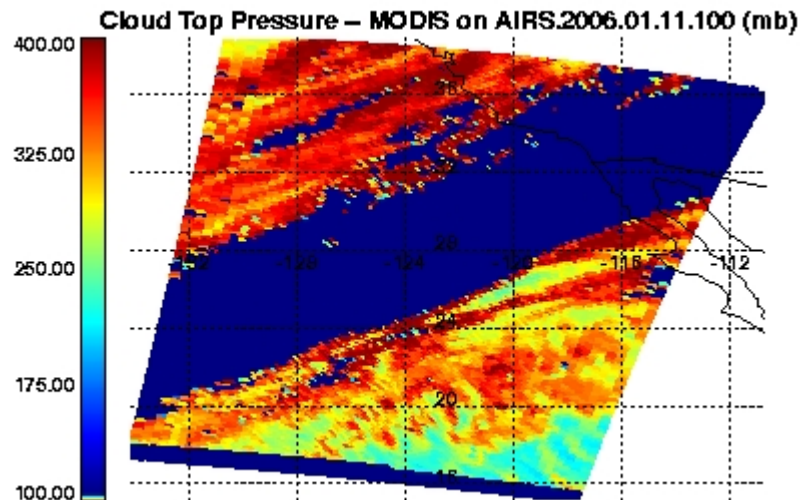
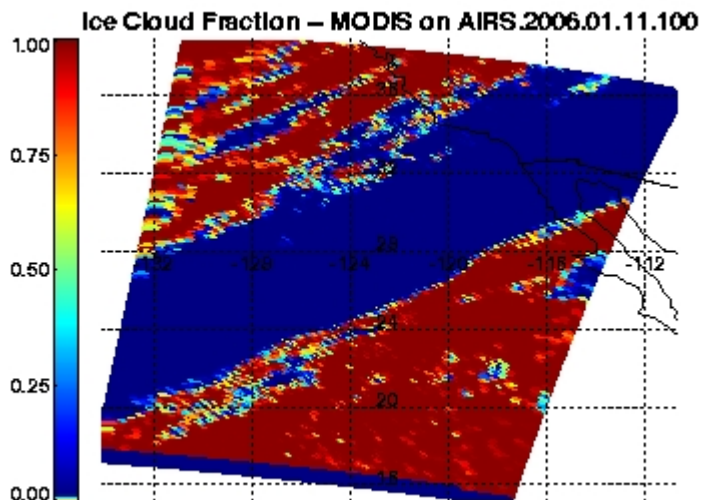
MADA optical properties (tropical cirrus)



- Single FOV 1DVAR retrieval
 - Atmosphere/SST from NCEP/GDAS
 - Adjusted parameters:
 - Cloud top/thickness
 - Ice particles effective diameter (Deff)
 - IWP
 - Effective temperature
 - MODIS 1st guess
 - AER/SERCAA cloud algorithms
 - RTM:
 - OSSSCAT (100 layers)
 - 4-streams

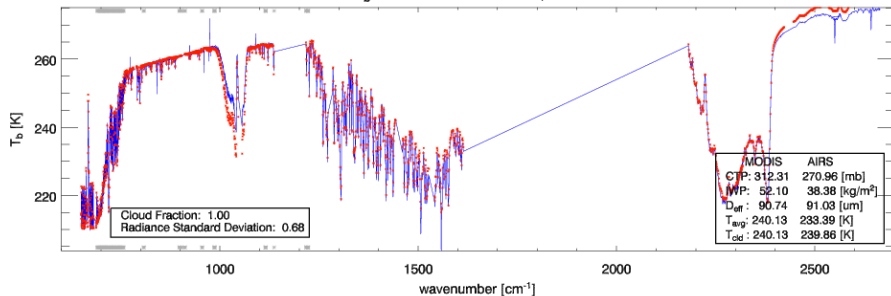
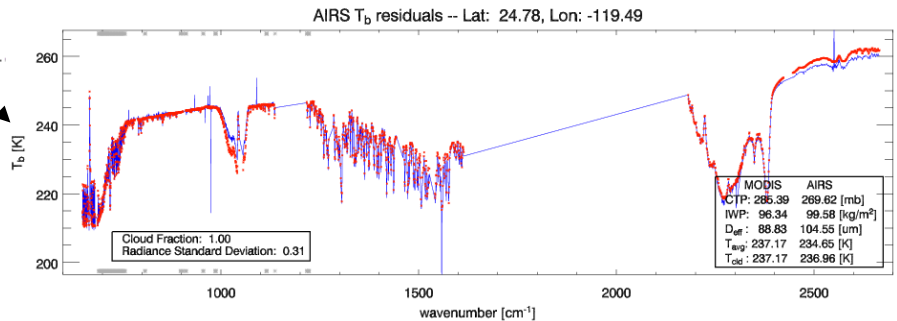
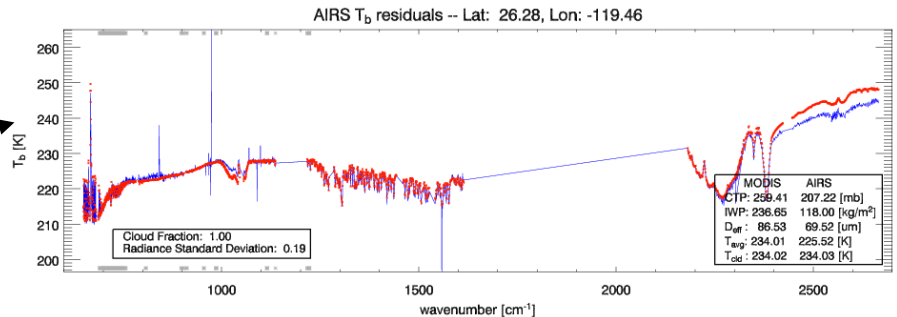
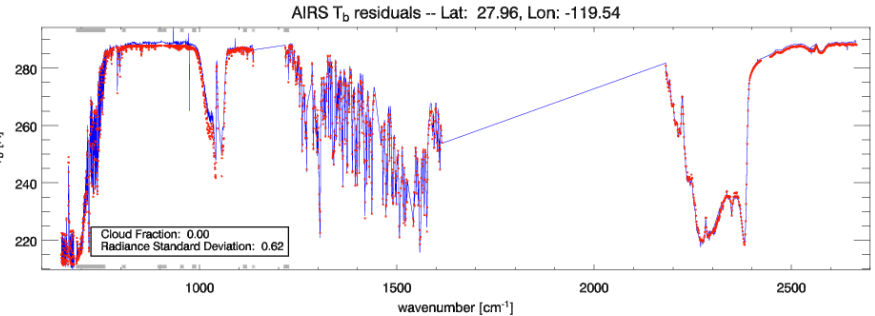
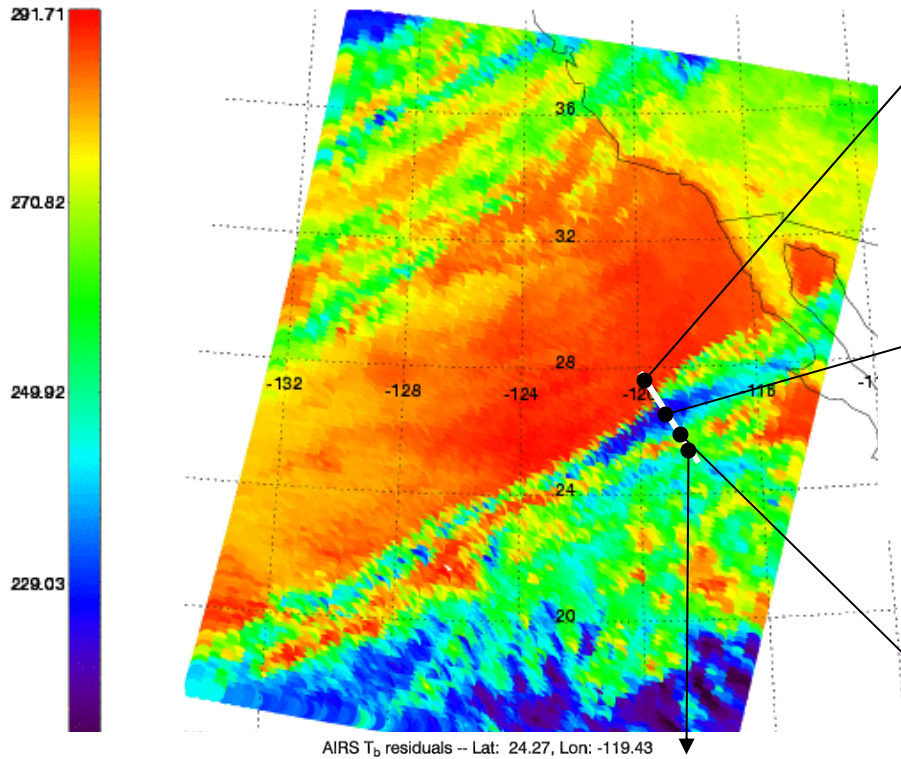


Retrieved cloud product



Calculated vs. measured cloudy AIRS spectra

AIRS (896 cm⁻¹) brightness temperature



● Localized training (reference):

- already offers higher numerical accuracy (both in clear and cloudy atmospheres) and significant speed gain over current OPTRAN based RT model
- Used for NPOESS/ATMS CrIS (older version) and CMIS EDR algorithms
- Considered for operations at NCEP for processing of current operational sensors (including AIRS)

● Non-localized (generalized) training:

- Potential for high computational gains (over localized training) for high spectral resolution IR sounders
 - Forward model is only one component of inversion algorithm
 - Further work needed to improve overall inversion speed
- Work on going for applications to land (spectrally variable surface emissivity) and cloudy atmosphere (spectrally variable cloud properties)

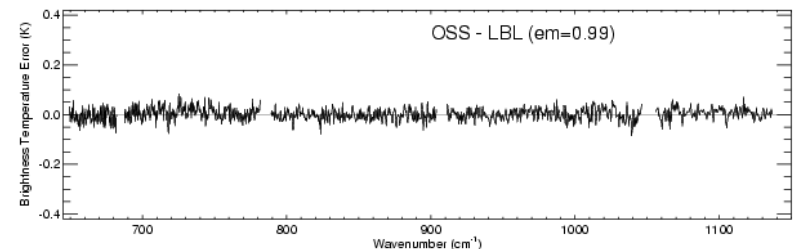
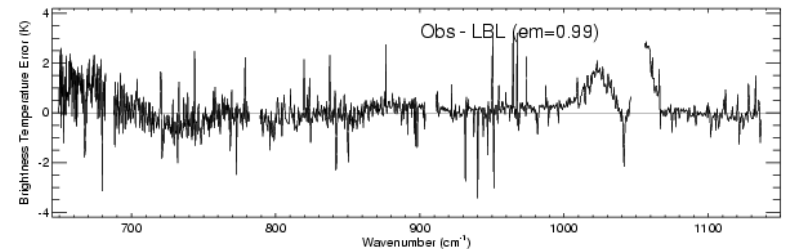
● Cloud modeling/retrieval in cloudy conditions

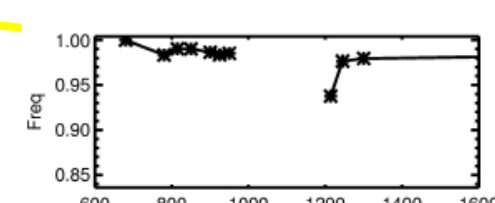
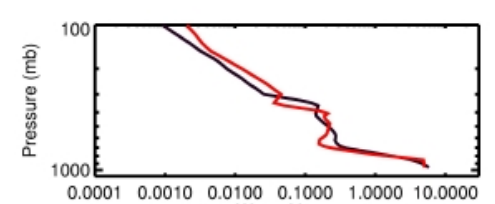
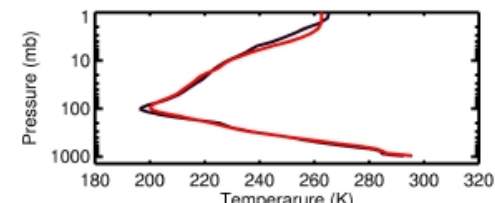
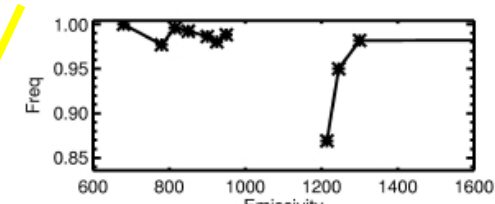
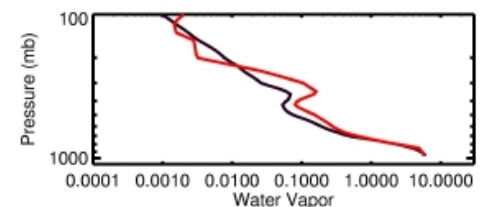
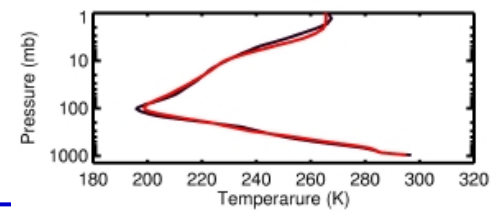
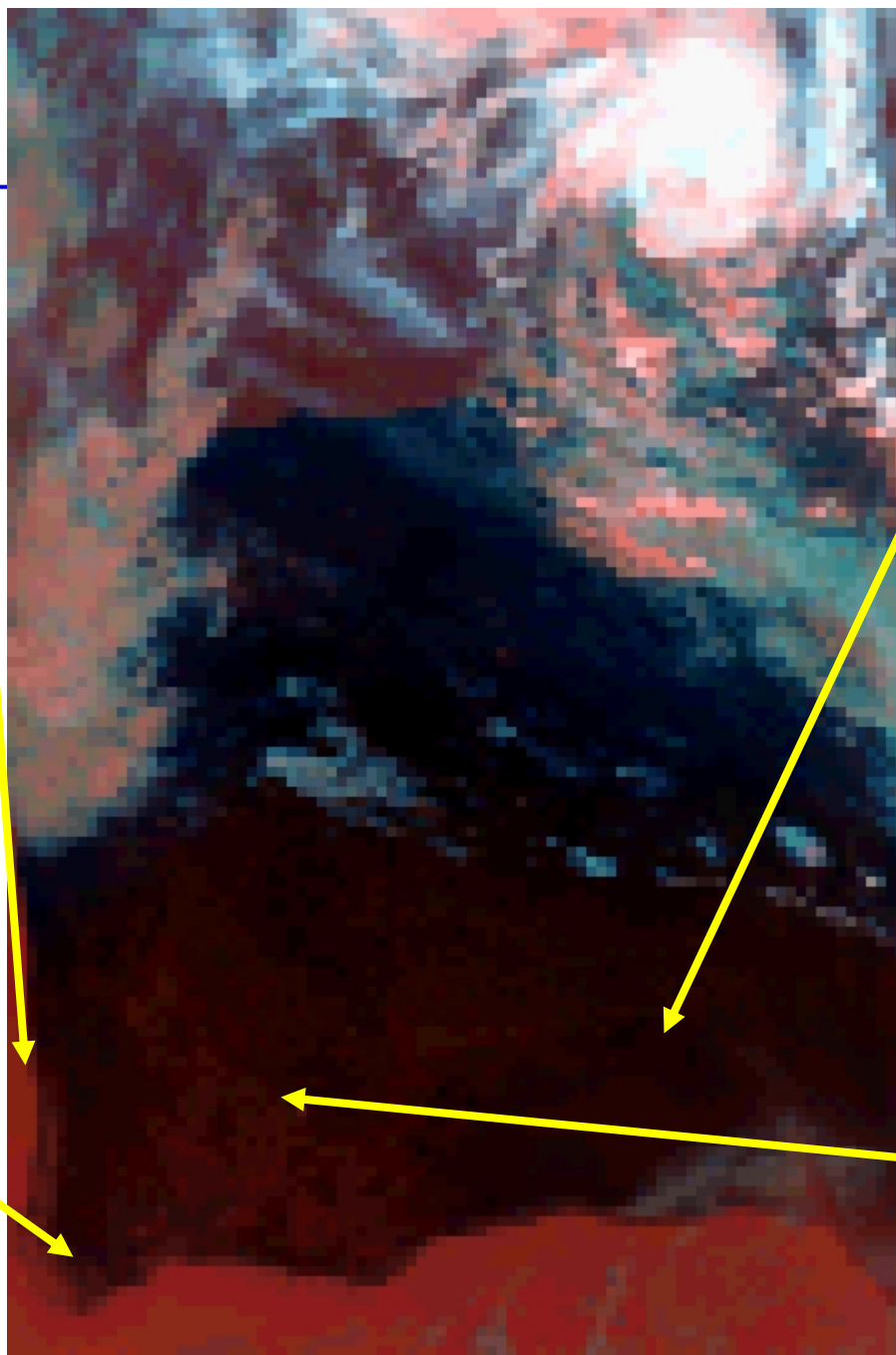
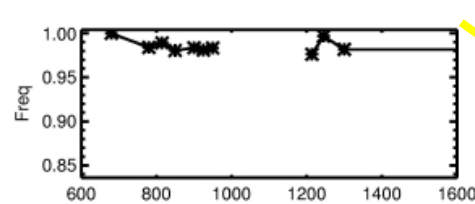
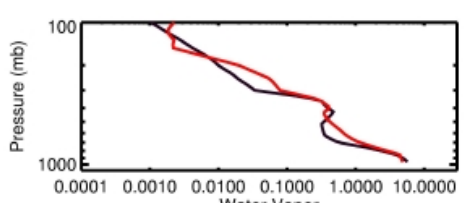
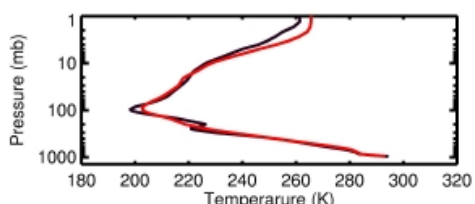
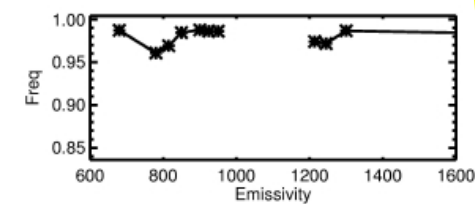
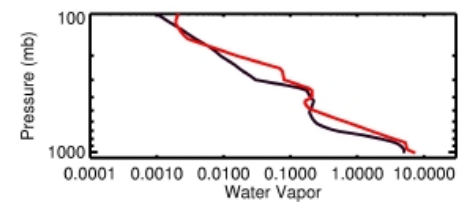
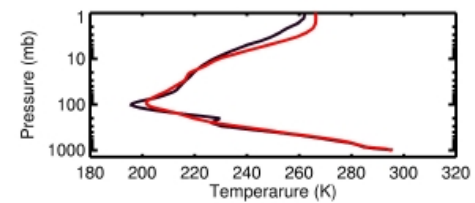
- Applied to AIRS cloud property retrieval
- Develop fast parameterizations for real-time application (already available for EO imaging instruments)
- Extend validation of RT model/cloud property parameterization to microwave (NPOESS/CMIS) and near-IR/visible region (AFRL/MODTRAN and MODIS applications)
- Collaboration with **NOAA-CU Center for Environmental Technology (CET) NOAA Earth System Research Laboratory** to include analytical Jacobians in scattering model
 - Goal: simultaneous retrieval of cloud and atmospheric composition

Summary/future work (cont.)

- Refine handling of solar source (clear/cloudy) in near-IR region
- Validate treatment of surface reflectivity over land
- Other focus areas:
 - improvement in molecular spectroscopy in both microwave and IR
 - Broadband flux/heating rate calculations

OSS vs. LBLRTM - AIRS clear-sky,
ARM TWP site (08/12/08)





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