

Atmospheric Cooling in the Far-Infrared

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Outline

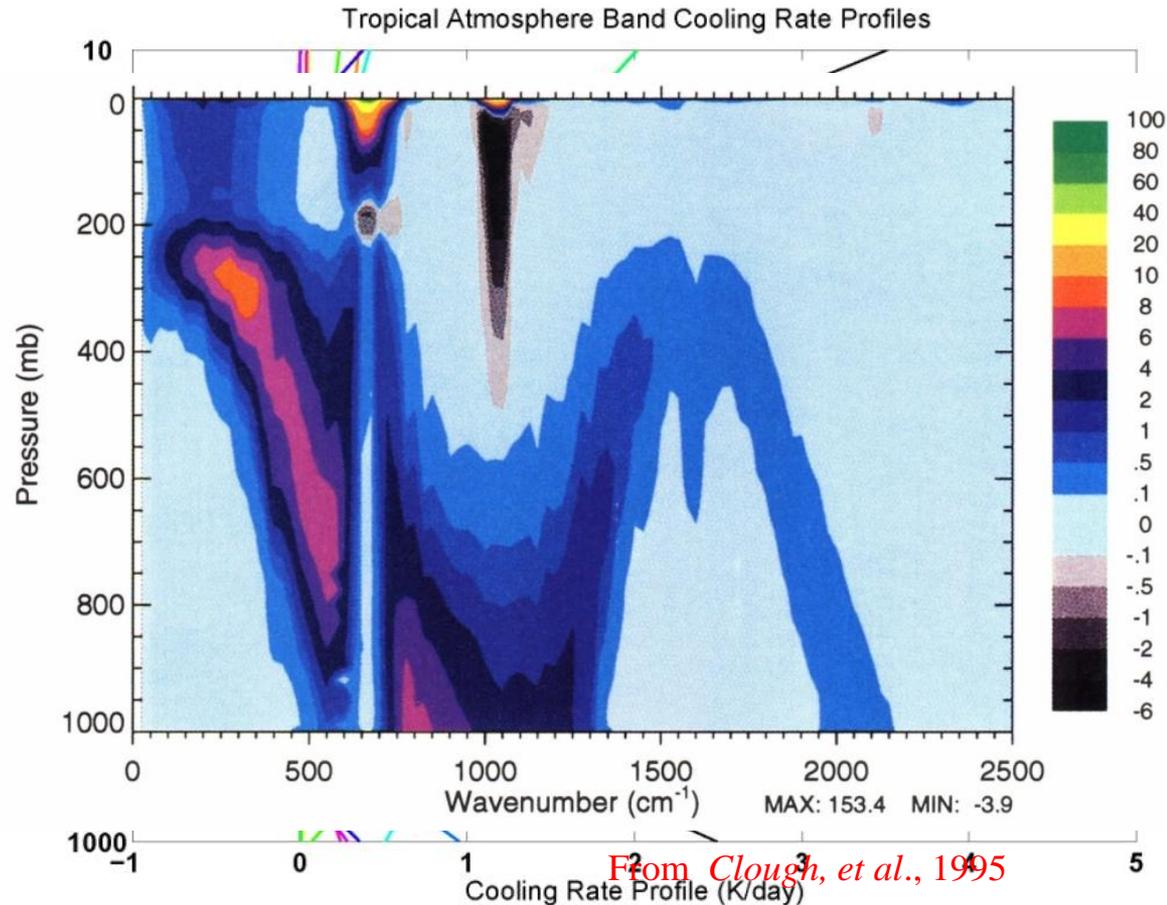
- Importance of atmospheric cooling and the role of the far-infrared.
- How measurements impart information on cooling rates (either directly or indirectly).
 - Comparison of distinct methods for determining cooling rates from measurements.
- How the current measurements from AIRS and CERES may be used to gauge processes that affect far-IR cooling.

Cooling Rate Profiles

- Cooling arises from net radiative flux divergence from absorption by gases including H₂O, CO₂, and O₃ and condensed species.

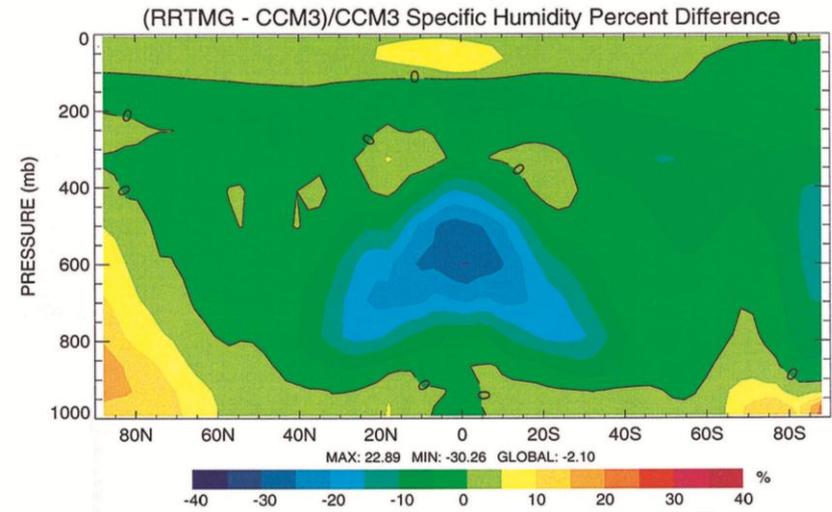
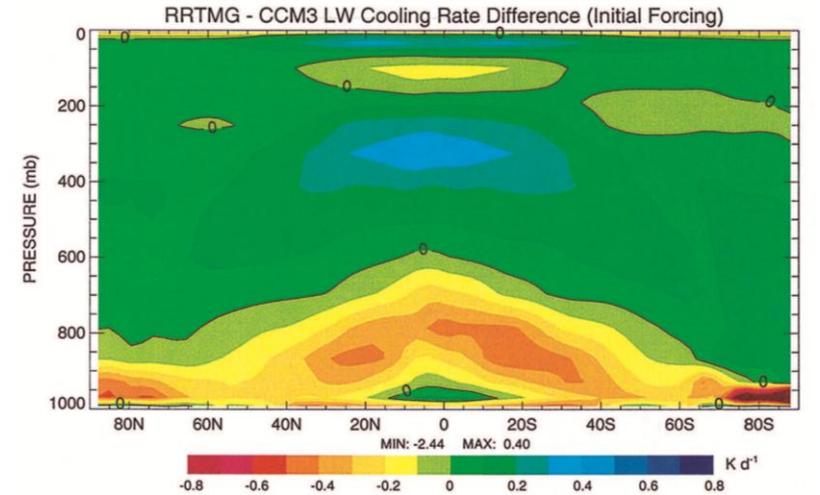
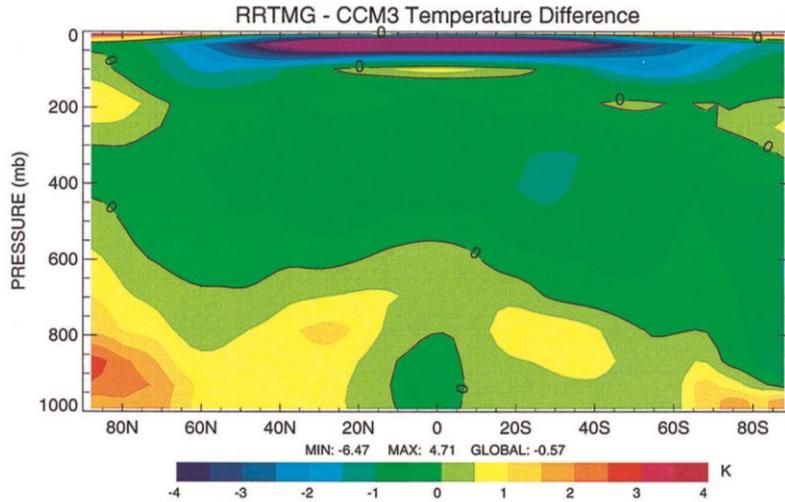
$$\dot{\theta}_v(z) = \frac{1}{C_p \rho(z)} \frac{dF_v^{NET}(z)}{dz}$$

- Models perform band radiation calculations to calculate heating/cooling rates to integrate the primitive equations.
 - Radiation can account for 30% of computational expense.
 - Radiation impacts circulation, especially vertical velocity, controls TTL and convection



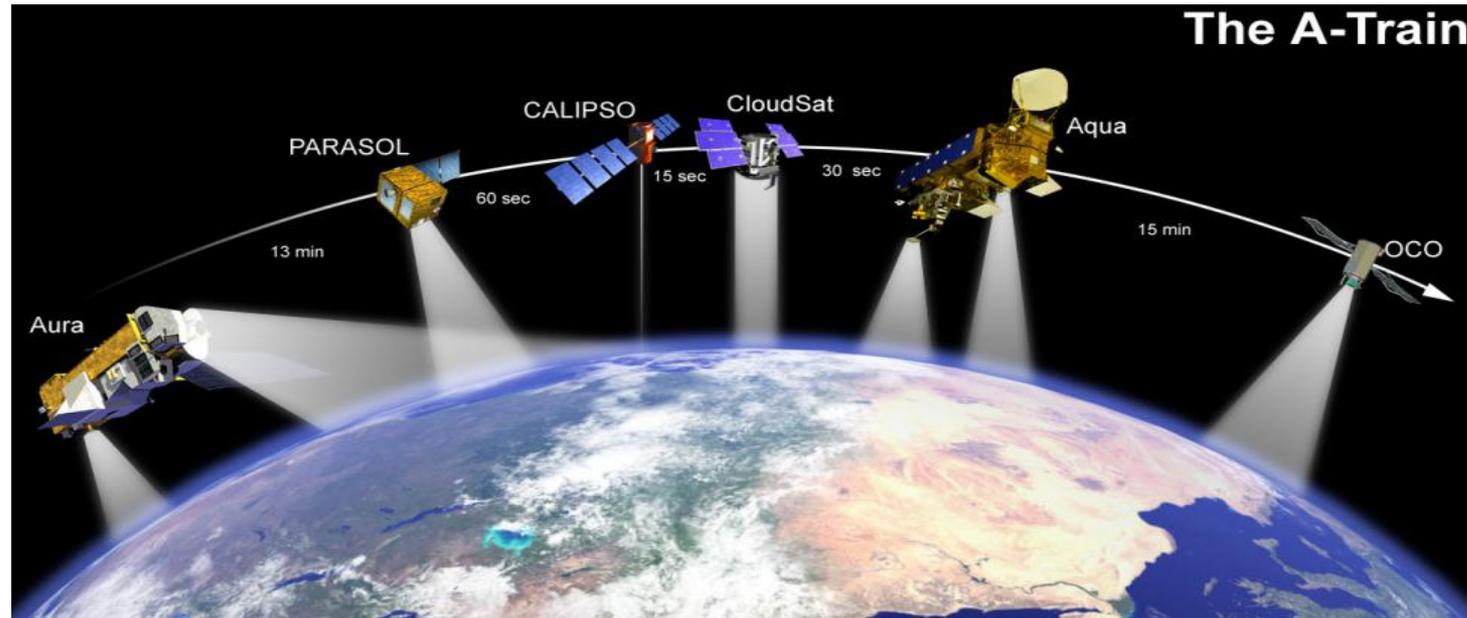
Heating/Cooling Rates Matter

- *Iacono, et al., [2000]* investigated the results of an RTM change in CCM3.
 - Significant cooling rate changes from revised H₂O continuum model.
- A comparison of model integrations shows changes in T, H₂O profiles due to altered latent, radiative energy distribution.



All figs. from *Iacono, et al., 2000*

Remote Sensing Measurements

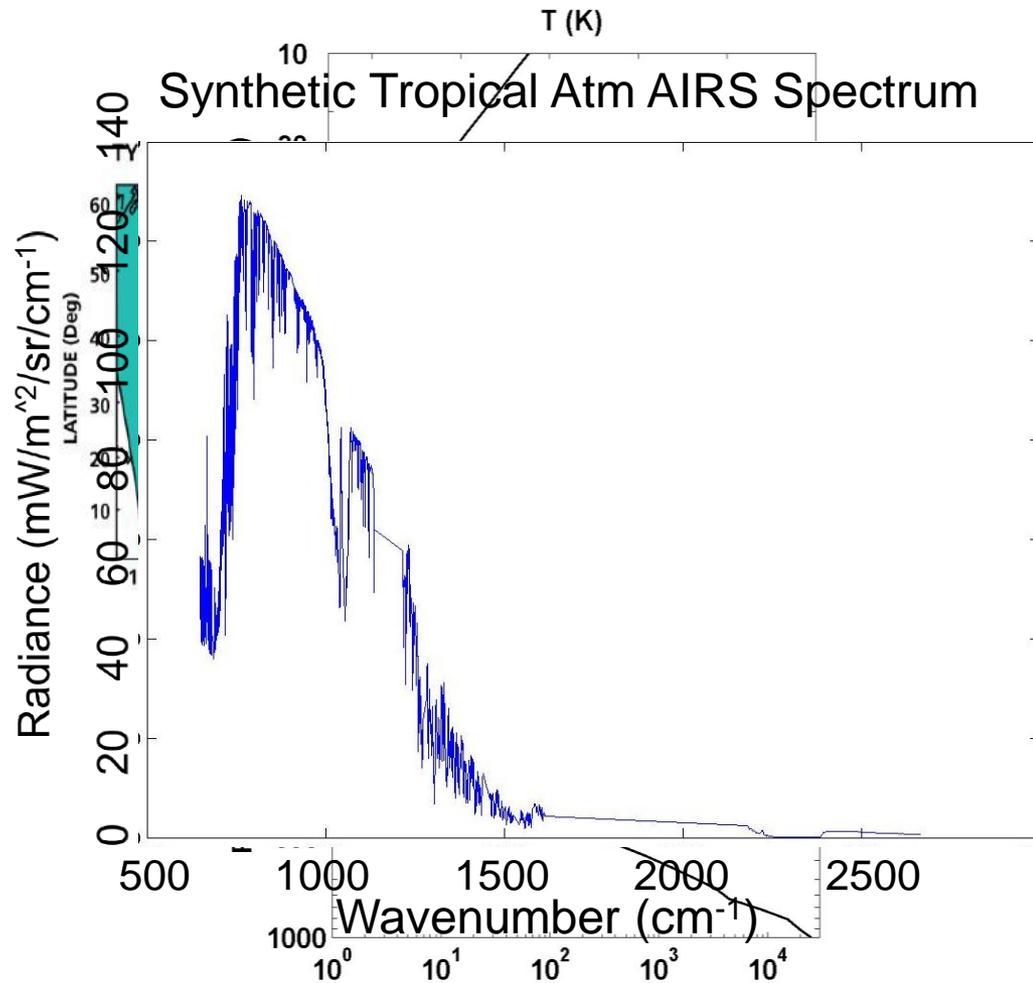


Artist's
rendition of
the A-Train
courtesy of
NASA

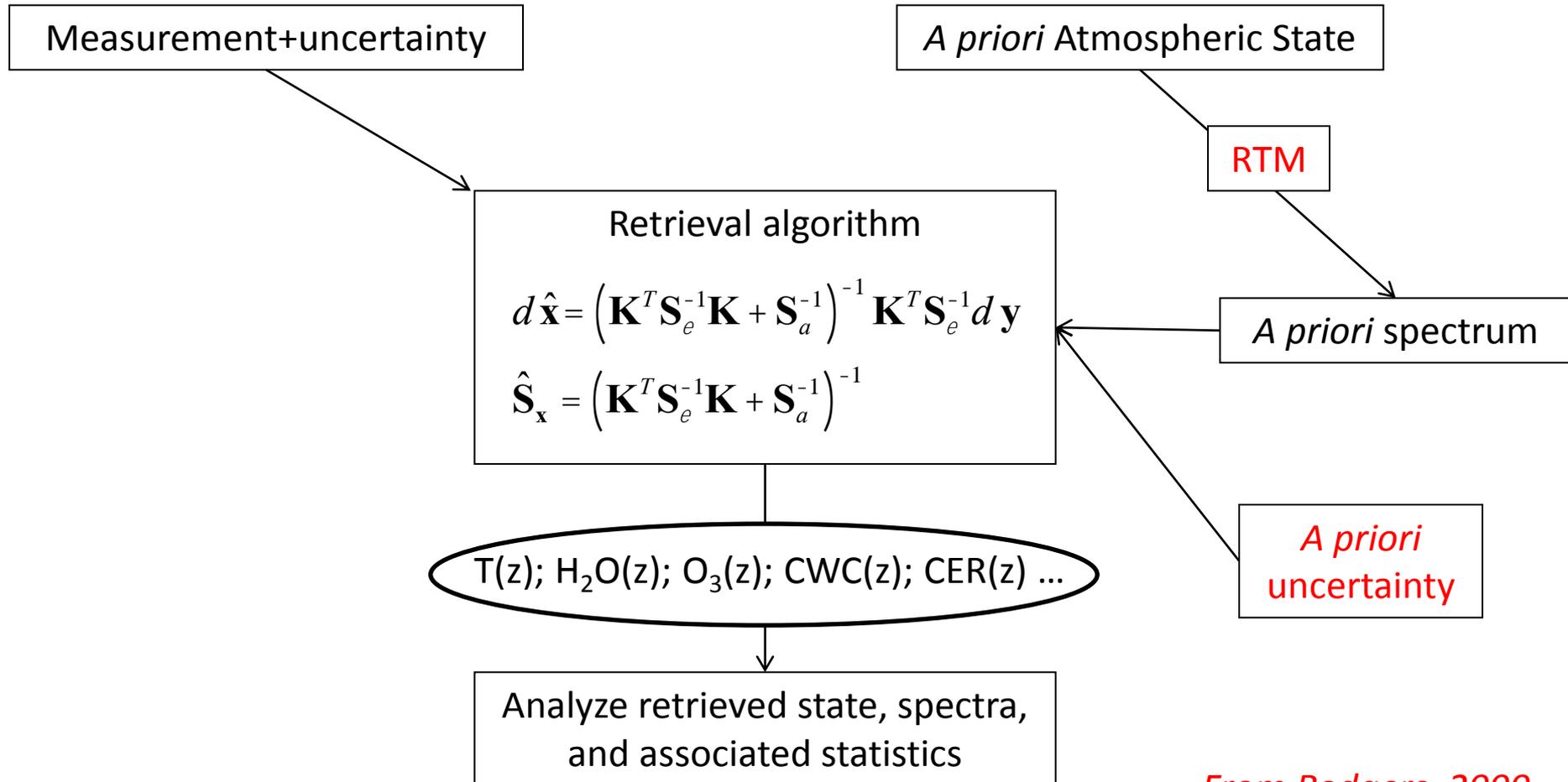
- There is a critical need for having accurate, computationally-inexpensive cooling rate data and remote sensing systems may be able to provide this.
- The polar-orbiting EOS A-Train flotilla presents a voluminous dataset describing the earth's lower atmosphere including T, H₂O, O₃, and clouds
 - AIRS has been operational for 9+ years.
 - CloudSat and CALIPSO platforms operational for 5+ years.

Information in AIRS measurements

- Passive IR spectra provide information on the T, H₂O, and O₃ profiles through differential absorption.
 - 2378 channels
 - 3.7 to 15.4 μm (650-2700 cm⁻¹)
 - No far-IR coverage mostly due to detector limitations.



Deriving Information from Retrieval Flow Chart

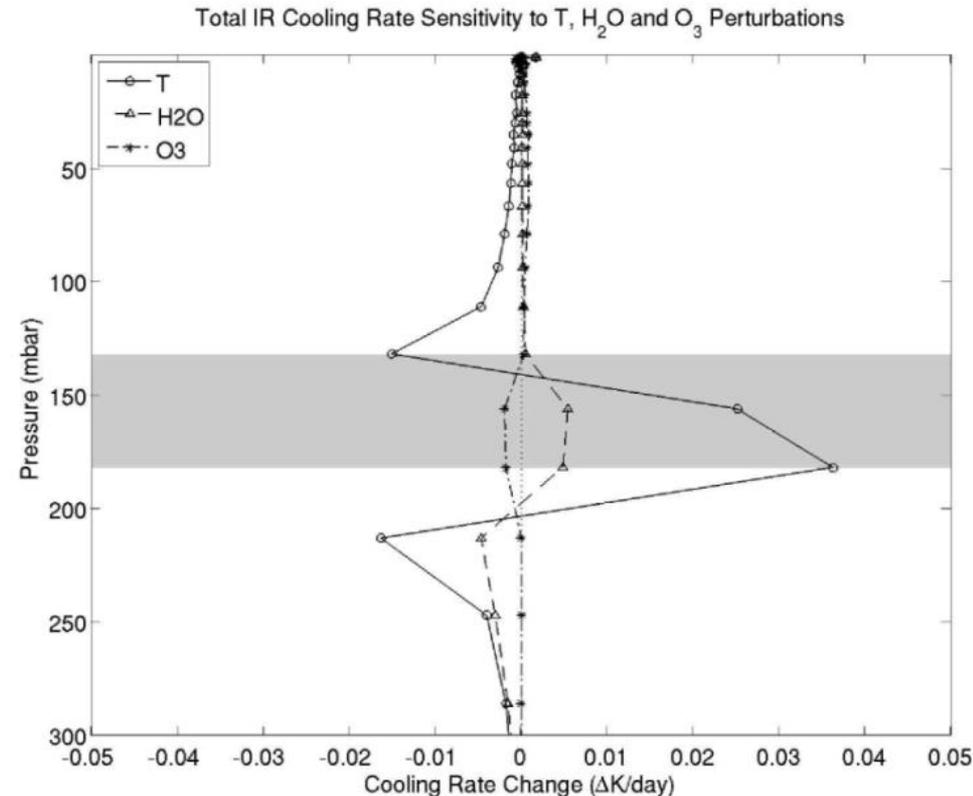


From Rodgers, 2000

Cooling Rate Profile Uncertainty

- Perturbations in T, H₂O, O₃ profiles lead to θ' changes that propagate across layers.
- Calculation of θ' uncertainty requires formal error propagation analysis.
 - Covariance counts!

$$[D\dot{q}(z)]^2 = \sum_{i=1}^n \sum_{j=1}^n \frac{\partial \dot{q}(z)}{\partial x_i} \frac{\partial \dot{q}(z)}{\partial x_j} \text{cov}(x_i, x_j)$$



From *Feldman, et al., 2008.*

Cooling Rate Error Propagation

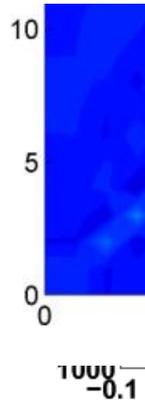
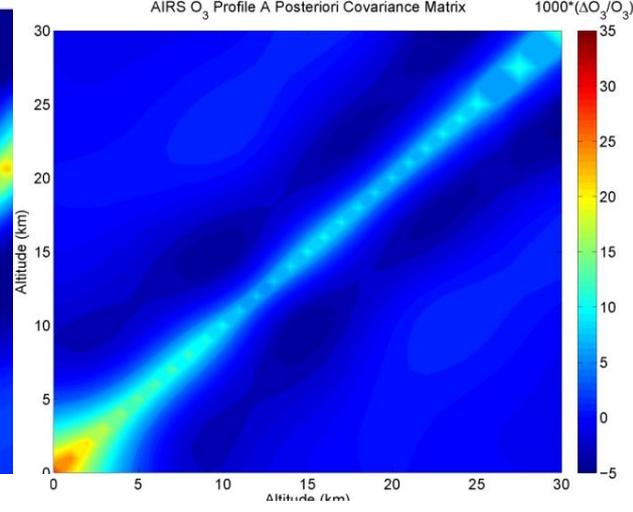
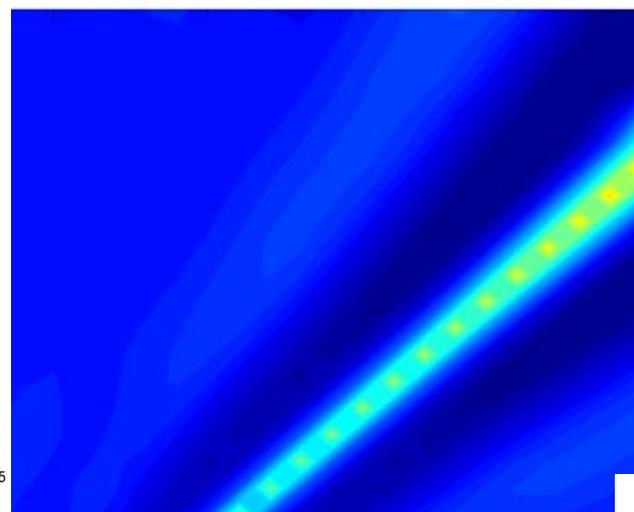
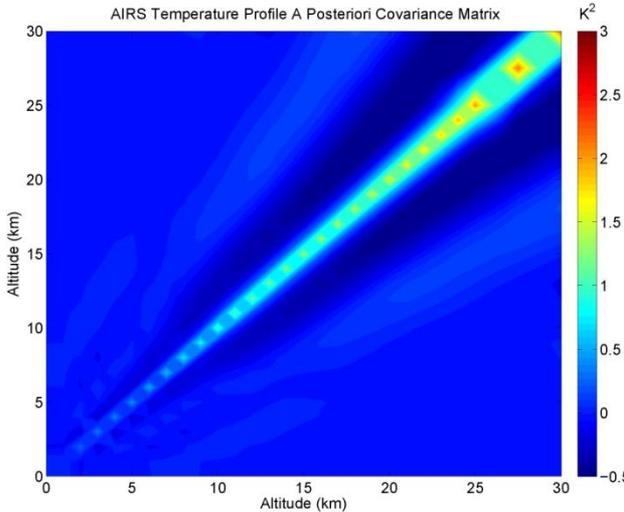
AIRS Temperature Profile A Posteriori Covariance Matrix

K^2

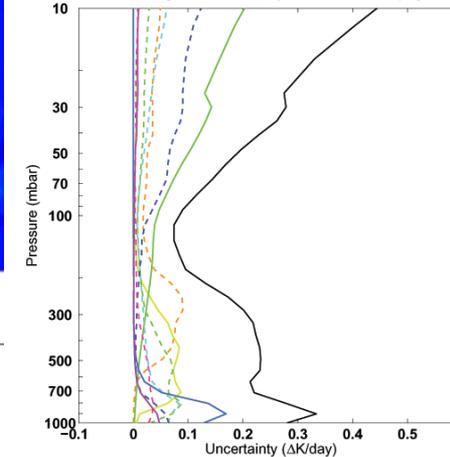
AIRS Temperature Profile A Posteriori Covariance Matrix

AIRS O₃ Profile A Posteriori Covariance Matrix

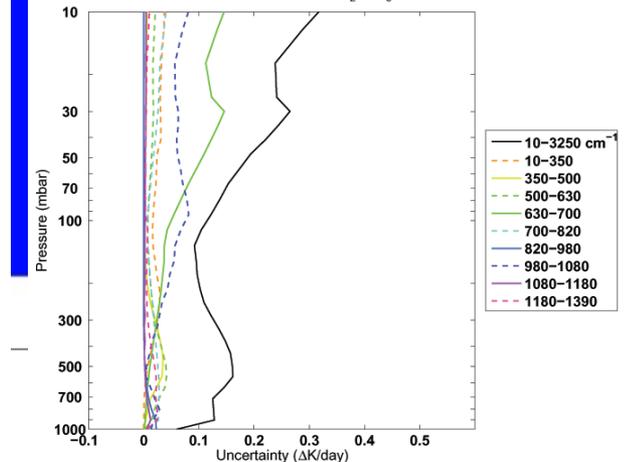
$1000 \cdot (\Delta O_3 / O_3)^2$



Band Cooling Rate Uncertainty: Formal Error Propagation



Band Cooling Rate Uncertainty: No T, H₂O, O₃ covariance



- T, H₂O, O₃ profile covariance matrix derived from kernel spread.
- Exclusion of off-diagonal elements leads to underestimation of uncertainty.

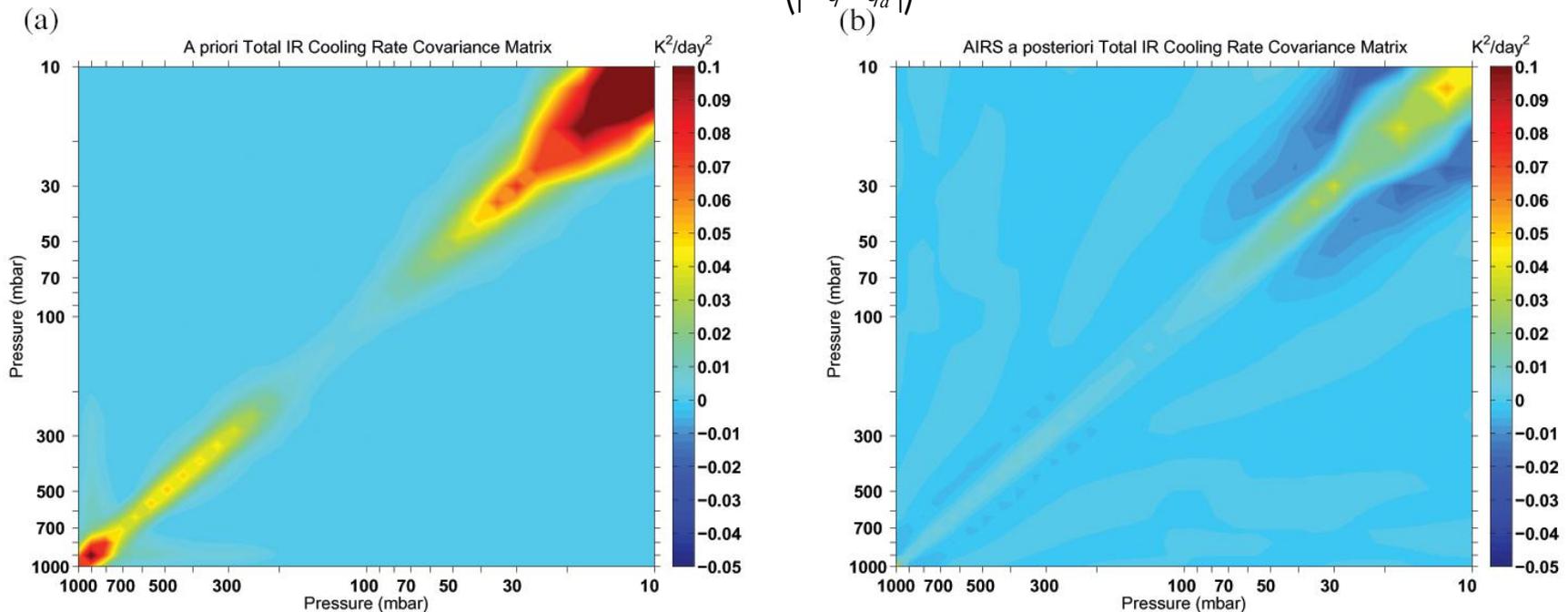
Cooling Rate Covariance Matrix

- Covariance matrices assess how prior and posterior errors are correlated

$$\hat{S}_q = \frac{\mathbb{J}_q \hat{S}_x \mathbb{J}_q^T}{\mathbb{J}_x \hat{S}_e \mathbb{J}_x^T}$$

- Information content of a measurement vis-avis the cooling rate profile can be assessed.

$$h \mu - \log \left(\left| \hat{S}_q \hat{S}_{q_a}^{-1} \right| \right)$$



From *Feldman, et al., 2008*

Cooling Rate Information Content

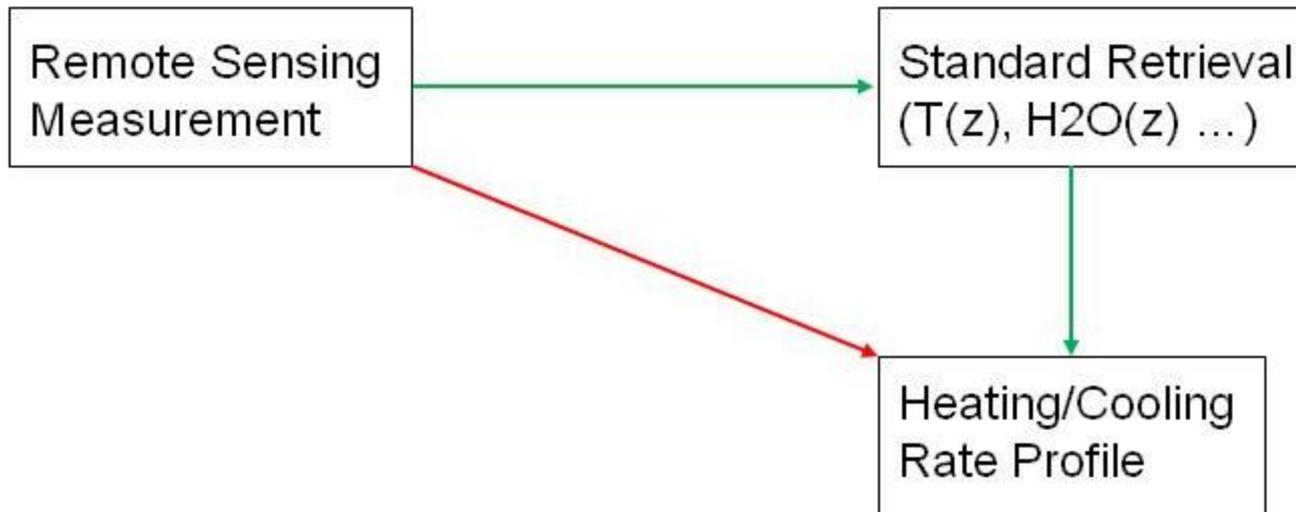
- Information content is related to the change in understanding of a set of correlated variables as a result of the measurement.

From *Feldman, et al. 2008*

Instrument	Time period	Spectral range (cm ⁻¹) ¹⁾	NeDT (K)	Spectral resolution (cm ⁻¹)	h _{TRP} (bits)	h _{MLS} (bits)	h _{SAW} (bits)
IRIS-D	1970-71	400-1600	2-4	2.8	9.8	8.4	6.4
AIRS	2002-Present	650-1400, 2100-2700	0.1-0.6	1-2	17.1	11.5	12.6
TES	2004-Present	650-1325, 1900-2250	1-4	0.12	13.2	10.5	8.0
IASI	2006-Present	650-2700	0.3-0.5	0.5	21.8	19.9	18.3
Far-IR	Proposed	200-2000	1.1	0.6	17.5	18.3	11.4

Retrieval of Cooling Rates

- Many products derived from the satellite instrument measurements through retrievals.
- Many different approaches to retrieving quantities from measurements.
 - Cooling rates retrieval proposed by Liou and Xue [1988] and Feldman et al [2006]



Inversion for Infrared Cooling Rate Profile

- Conventionally use T, H₂O, O₃, CH₄, and N₂O profiles
- Remote sensing measurements can be inverted for atmospheric state → calculate cooling rates.
- TOA radiance closely related to TOA flux which is a function of net flux divergence
 - Monotonic kernel but *a priori* constraint guarantees measurement information imparted uniformly across profile

Measurement

$$I_n(+m, z) = B_n(q_{surf}) T_n(z, 0) + \int_0^z B_n(q(z')) \frac{\partial T_n(z, z')}{\partial z'} dz'$$

TOA Flux

$$F^{TOA} = F^{SURF} + \int_0^\infty \frac{q'(z)}{r(z) C_p} dz$$

Flux interpretation

$$dF^{TOA} = \sum_{i=1}^n \frac{\partial F^{SURF}}{\partial x_i} dx_i + \int_0^\infty \frac{1}{r(z) C_p} \sum_{i=1}^n \frac{\partial q'(z)}{\partial x_i} dx_i dz$$

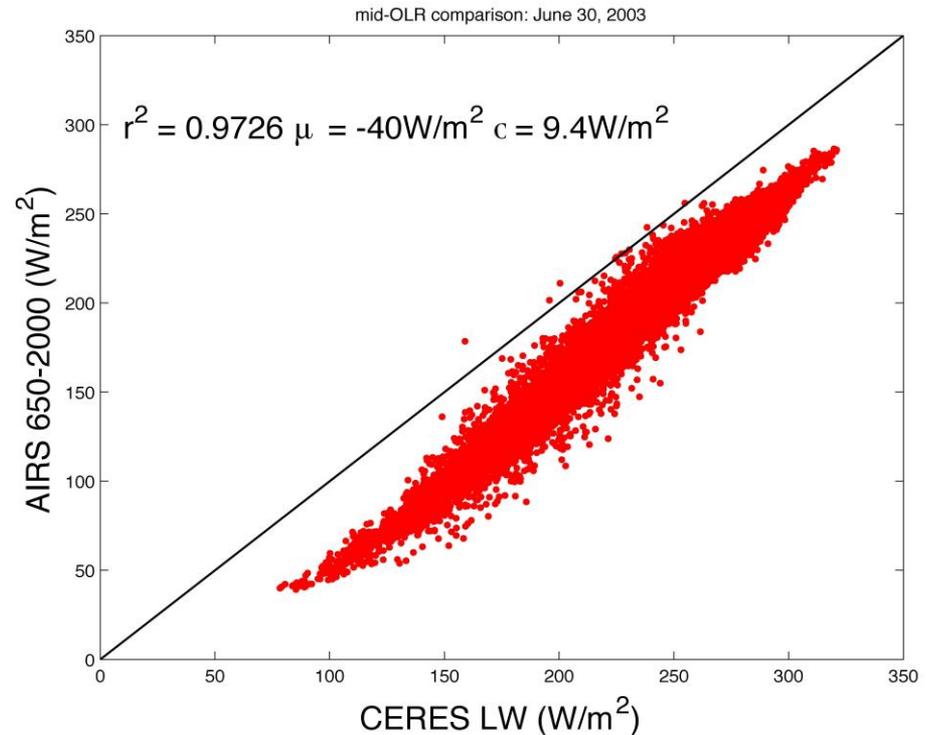
Kernel

In the absence of TOA far-IR measurements

- Extensive datasets of mid-IR measurements and also information from CloudSat and CALIPSO.
 - Standard retrieval products can (and should) be used to gauge estimate far-IR cooling
 - These require covariance matrix estimates!
- In the absence of far-IR measurements, measurements from other sources must be extrapolated
 - Critical requirement for accurate spectroscopy, cloud optical properties, and an appropriate mapping from mid-IR clouds to far-IR clouds.
- The data is out there!
 - Routine calculations are performed by weather models, in climate models and for satellite products. Comparison required.

Using AIRS + CERES to understand far-IR

- AIRS radiance spectra has been converted to spectrally-resolved fluxes collocated with Aqua CERES LW fluxes [Huang et al, 2010].
- A comparison of the principal components AIRS mid-IR flux ($650\text{-}2000\text{ cm}^{-1}$) to CERES broadband flux ($200\text{-}2000\text{ cm}^{-1}$) will indicate the extra information in the far-IR.
- Process studies are likely required to understand how the discrepancies between mid-IR and broadband flux map onto far-IR cooling.



Data courtesy of Xianglei Huang

Discussion

- Radiative cooling rates are important to atmospheric circulation and far-IR cooling from water vapor and clouds represent a significant component of this cooling.
- Remote sensing measurements provide information, either directly or indirectly, about spectral and broadband cooling rates
 - The incorporation of this information requires careful attention to retrieval theory.
- Ongoing efforts to compare mid- and broadband IR measurements from AIRS and CERES respectively may yield extra information about far-IR flux and ultimately far-IR cooling.

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