

→ **ADVANCED ATMOSPHERIC TRAINING COURSE 2014**

MID-IR Nadir/Limb Retrieval

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Thank's to:

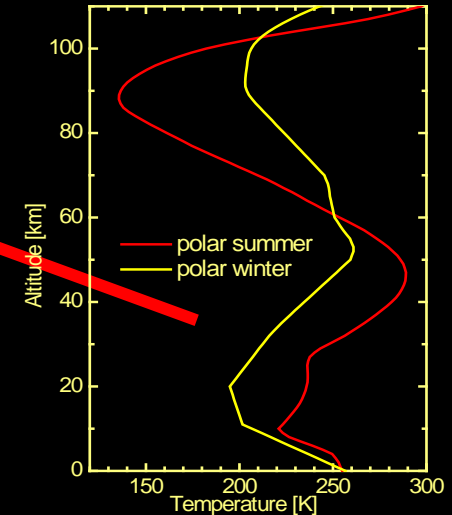
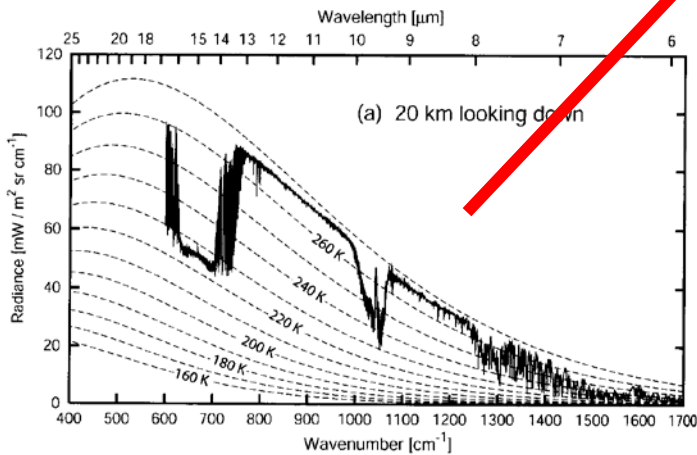
- André Butz
- Frank Hase
- Thomas v. Clarmann

Inversion of remote sensing observations

Forward:

$$\hat{y} = \hat{F}(\hat{x})$$

Signal (e.g. nadir spectrum)



Radiative transfer

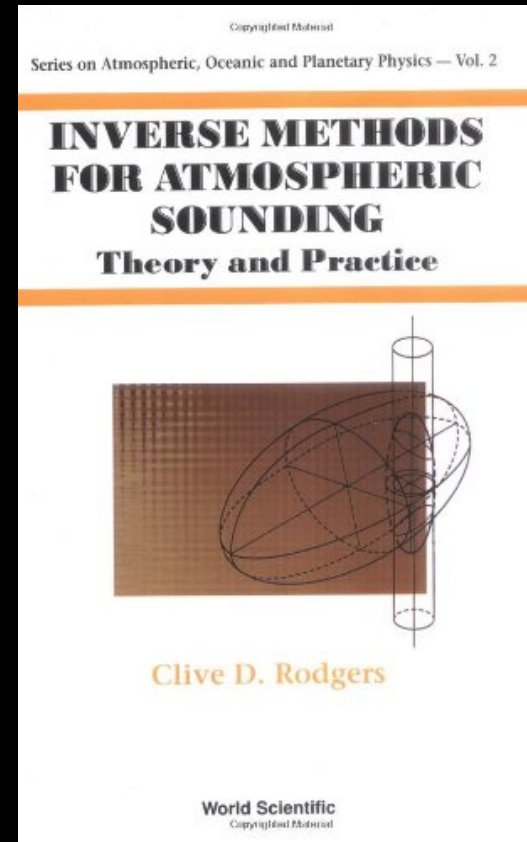
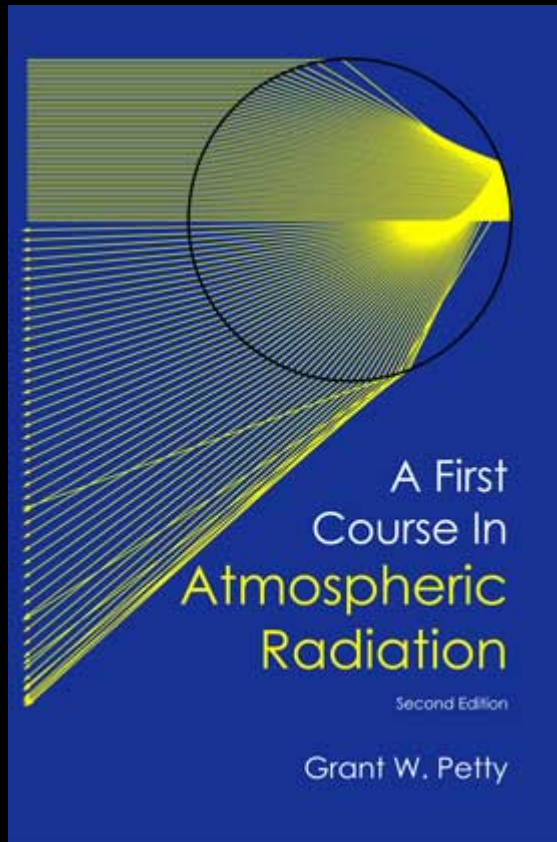
$$L_{\lambda}^{\uparrow}(\tau_{\lambda}, \mu, \varphi) = L_{\lambda}^{\uparrow}(\tau_{\lambda, \text{Boden}}, \mu, \varphi) e^{-\frac{\tau_{\lambda, \text{Boden}} - \tau_{\lambda}}{\mu}} + \int_{\tau_{\lambda}}^{\tau_{\lambda, \text{Boden}}} J_{\lambda}(\tau'_{\lambda}, \mu, \varphi) e^{-\frac{\tau'_{\lambda} - \tau_{\lambda}}{\mu}} \frac{1}{\mu} d\tau'_{\lambda}$$

Inverse:

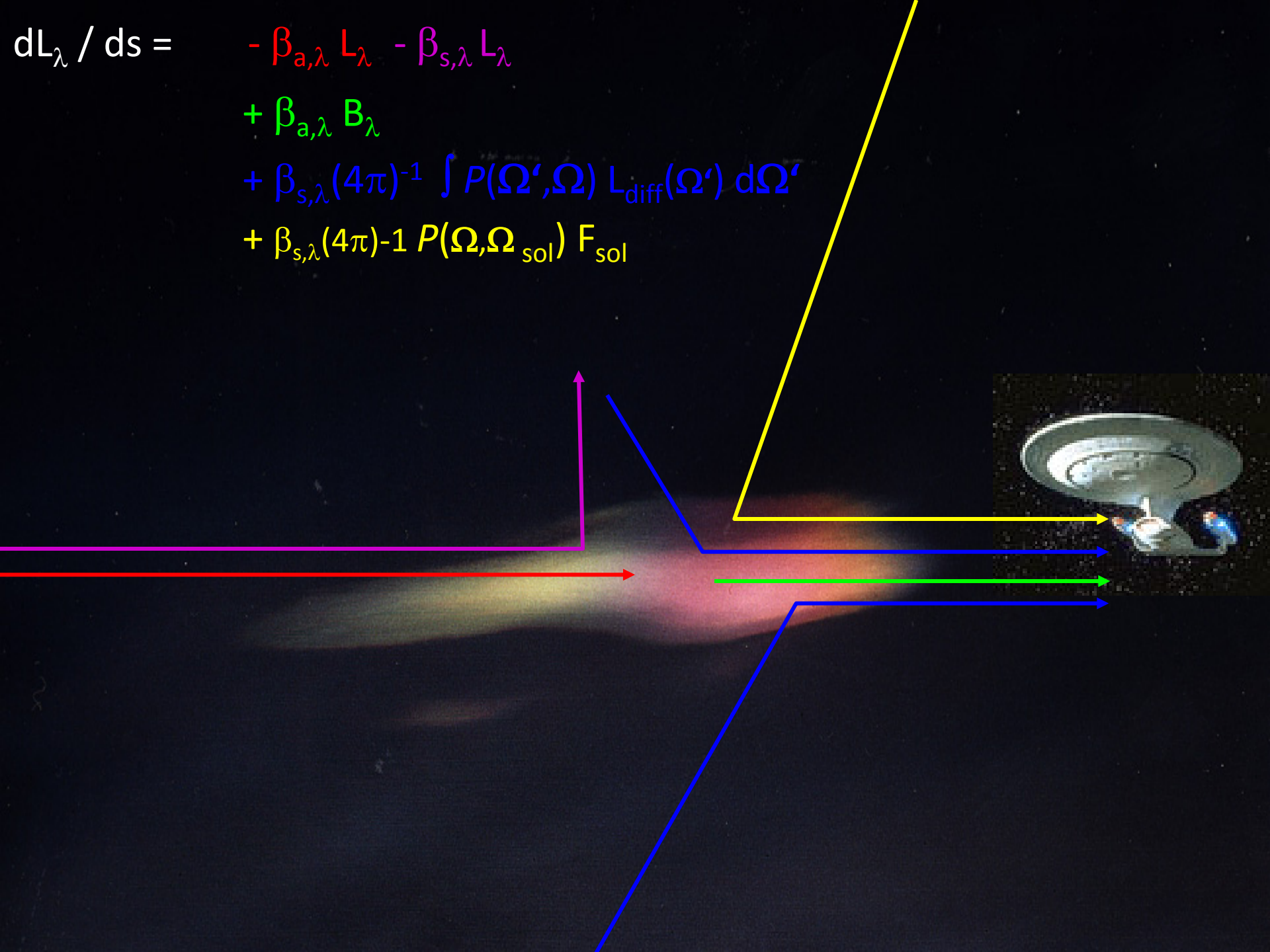
$$\hat{x} = \hat{F}^{-1}(\hat{y}) \quad ???$$

$$\hat{y} = \hat{F}(\hat{x})$$

$$\hat{x} = \hat{F}^{-1}(\hat{y})$$



$$\begin{aligned} dL_\lambda / ds = & -\beta_{a,\lambda} L_\lambda - \beta_{s,\lambda} L_\lambda \\ & + \beta_{a,\lambda} B_\lambda \\ & + \beta_{s,\lambda} (4\pi)^{-1} \int P(\Omega', \Omega) L_{\text{diff}}(\Omega') d\Omega' \\ & + \beta_{s,\lambda} (4\pi)^{-1} P(\Omega, \Omega_{\text{sol}}) F_{\text{sol}} \end{aligned}$$



Emission and Absorption

- Only emission and absorption of radiation, no scattering
→ most important for radiative transfer in MW and IR
- Local thermodynamic equilibrium
→ Planckfunction as source function
→ Kirchhoff's law: emission = absorption

$$\begin{aligned}
 dL_\lambda &= L_\lambda(s + ds) - L_\lambda(s) \\
 &= dL_{\lambda,abs} + dL_{\lambda,emi} \\
 &= -\beta_{a,\lambda}(s)L_\lambda(s)ds + \beta_{a,\lambda}(s)B_\lambda(s)ds
 \end{aligned}$$

Absorption coefficient:

$$\beta_{a,\lambda} = \sigma_{a,\lambda}N$$

Absorption cross-section:

$$\sigma_{a,\lambda}$$

Number density:

$$N$$

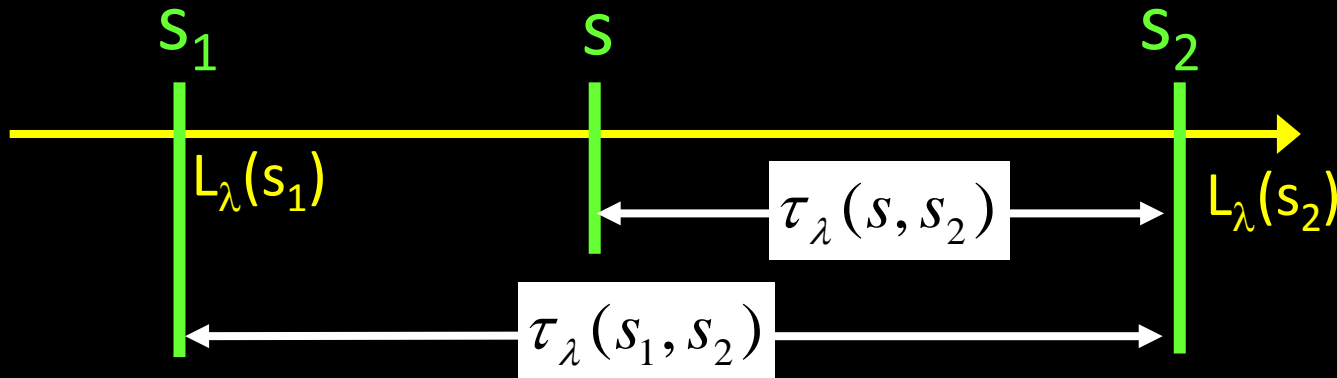
Planck function:

$$B_\lambda(s)$$

Schwarzschild-equation:

$$\frac{dL_\lambda}{ds} = \beta_{a,\lambda}(s)(B_\lambda(s) - L_\lambda(s))$$

Solution of Schwarzschild's equation



Optical depth:

$$\tau_{\lambda}(s_1, s_2) = \int_{s_1}^{s_2} \beta_{a,\lambda}(s) ds$$

$$L_{\lambda}(s_2) = L_{\lambda}(s_1) e^{-\tau_{\lambda}(s_1, s_2)} + \int_{s_1}^{s_2} \beta_{a,\lambda}(s) B_{\lambda}(s) e^{-\tau_{\lambda}(s, s_2)} ds$$

Emission of a layer with constant T

$$L_{\lambda}(s_2) = \int_0^{\tau_{\lambda}(s_1, s_2)} B_{\lambda}(s) e^{-\tau_{\lambda}(s, s_2)} d\tau_{\lambda}(s, s_2)$$

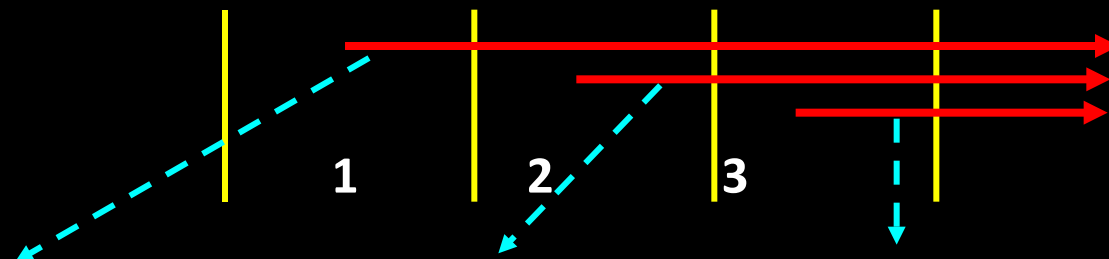
$$B_{\lambda}(s) = \text{const} = B_{\lambda}$$



$$L_{\lambda}(s_2) = B_{\lambda} \int_0^{\tau_{\lambda}(s_1, s_2)} e^{-\tau_{\lambda}(s, s_2)} d\tau_{\lambda}(s, s_2) = B_{\lambda} (1 - e^{-\tau_{\lambda}(s_1, s_2)}) = B_{\lambda} \left(1 - \underbrace{t_{\lambda}(s_1, s_2)}_{\text{Transmission of the layer}} \right)$$

Emissivity of the layer

For three layers:



$$L_{\lambda} = B_{\lambda,1}(1-t_{\lambda,1})t_{\lambda,2}t_{\lambda,3} + B_{\lambda,2}(1-t_{\lambda,2})t_{\lambda,3} + B_{\lambda,3}(1-t_{\lambda,3})$$

$$= B_{\lambda,1}(t_{\lambda,2}t_{\lambda,3} - t_{\lambda,1}t_{\lambda,2}t_{\lambda,3}) + B_{\lambda,2}(t_{\lambda,3} - t_{\lambda,2}t_{\lambda,3}) + B_{\lambda,3}(1-t_{\lambda,3})$$

Thermal contrast in nadir sounding: the lowest layer

$$L_{\lambda} = B_{\lambda, \text{surface}} t_{\lambda, 1} + B_{\lambda, 1} (1 - t_{\lambda, 1})$$

$$B_{\lambda, \text{surface}} = B_{\lambda, 1}$$

→

$$L_{\lambda} = B_{\lambda, 1}$$

→ No information about transmission of lowest layer

$$B_{\lambda, \text{surface}} \neq B_{\lambda, 1}$$

→

$$t_{\lambda, 1} = \frac{L_{\lambda} - B_{\lambda, 1}}{B_{\lambda, \text{surface}} - B_{\lambda, 1}}$$



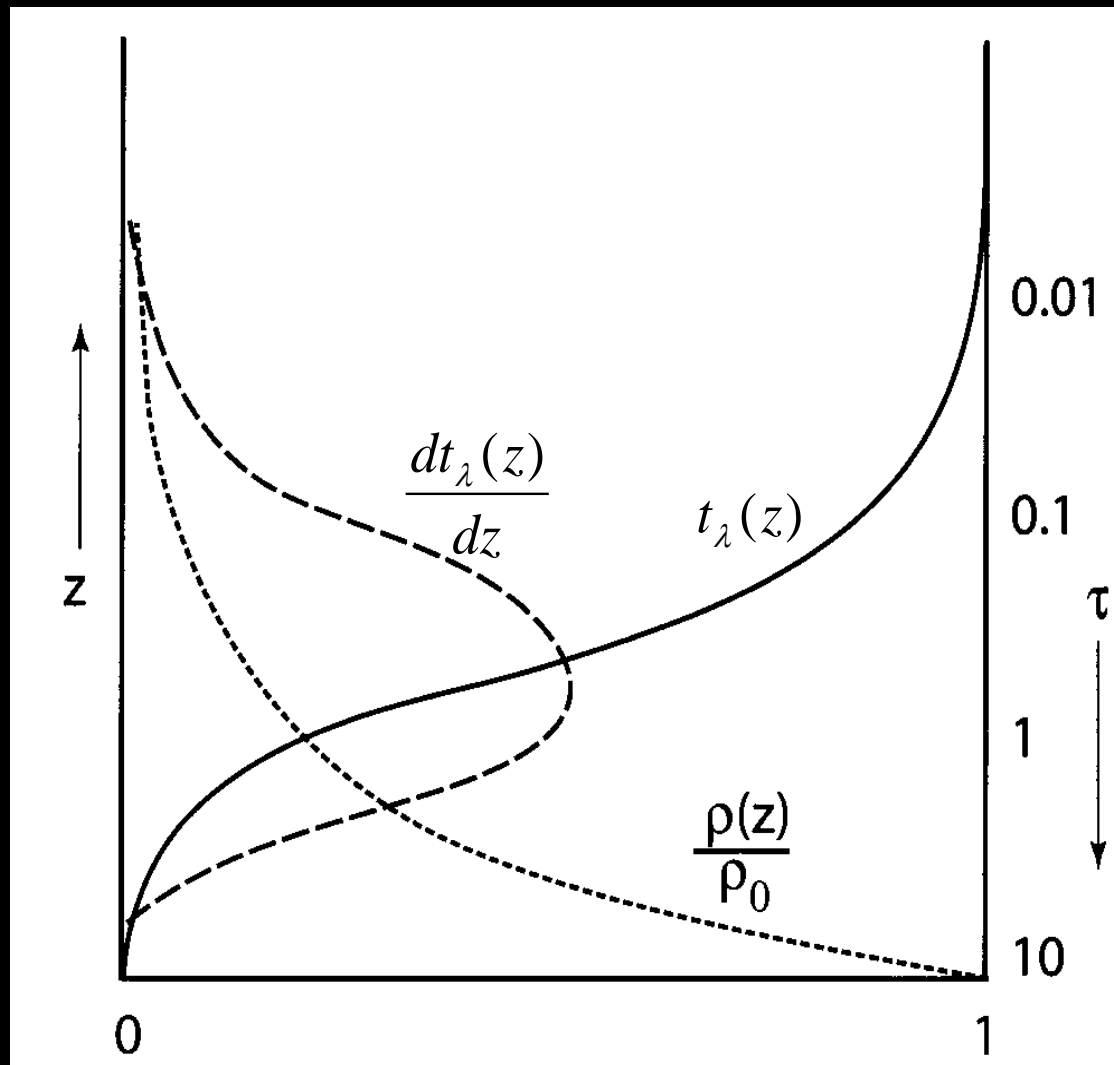
Schwarzschild's equation as function of transmission and weighting function

$$L_{\lambda}(s_2) = L_{\lambda}(s_1)e^{-\tau_{\lambda}(s_1, s_2)} + \int_{s_1}^{s_2} \beta_{a, \lambda}(s) B_{\lambda}(s) e^{-\tau_{\lambda}(s, s_2)} ds$$

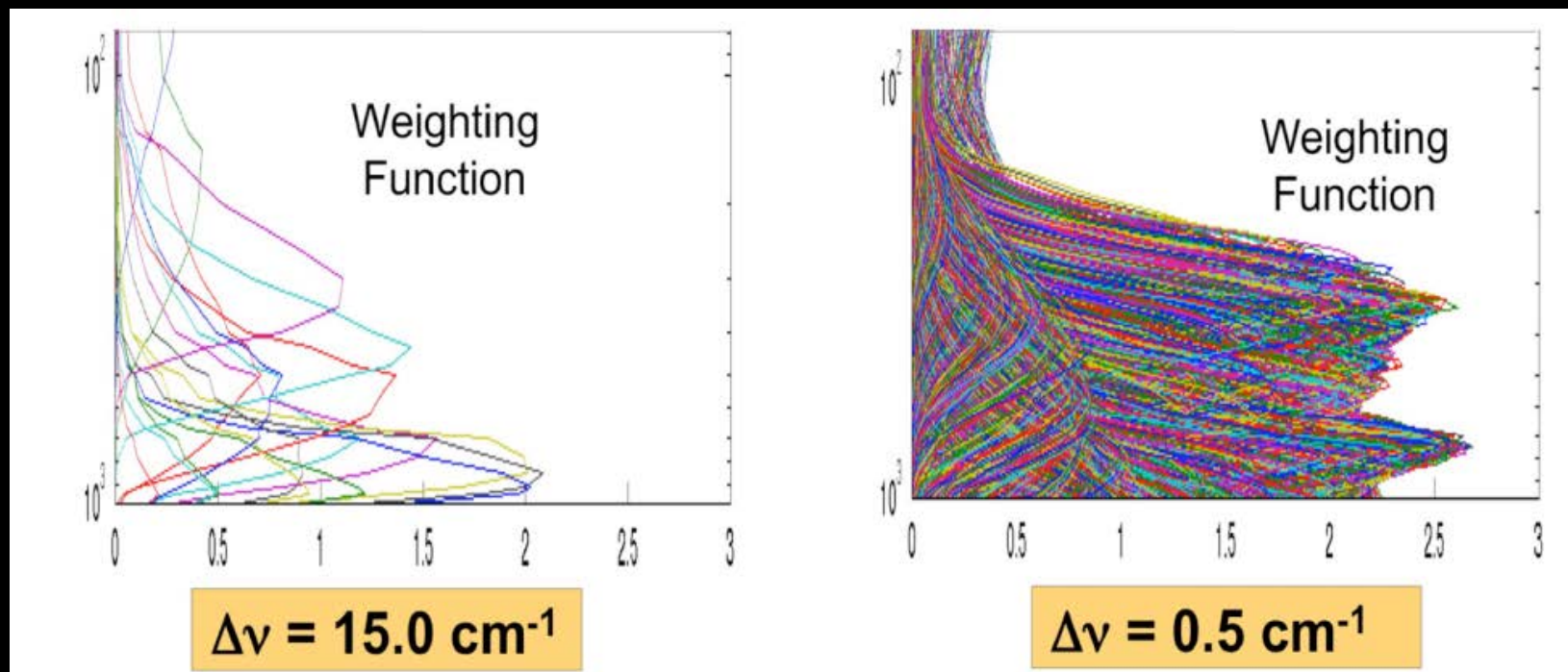
$$L_{\lambda}(s_2) = L_{\lambda}(s_1)t_{\lambda}(s_1, s_2) + \int_{s_1}^{s_2} B_{\lambda}(s)W_{\lambda}(s)ds$$

Transmission: $t_{\lambda}(s_1, s_2) = \exp[-\tau_{\lambda}(s_1, s_2)]$

Weighting function: $W_{\lambda}(s) = \frac{dt_{\lambda}(s, s_2)}{ds}$



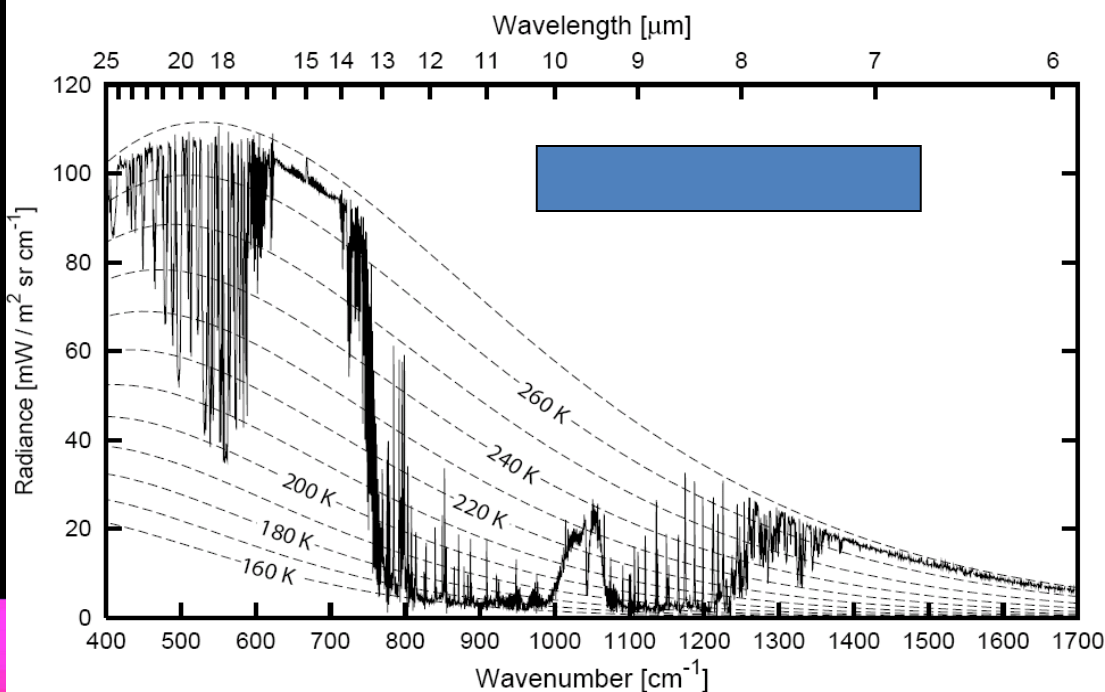
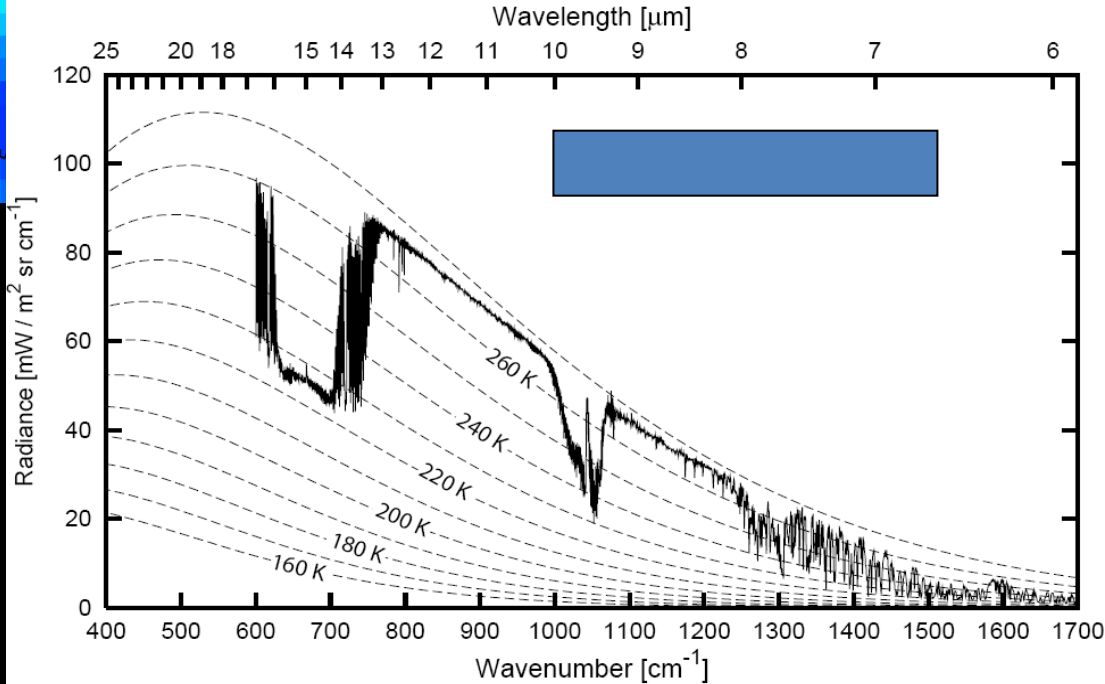
Nadir sounding weighting functions



Smith et al., 2009

grün

Looking up or down?



Looking up: where?

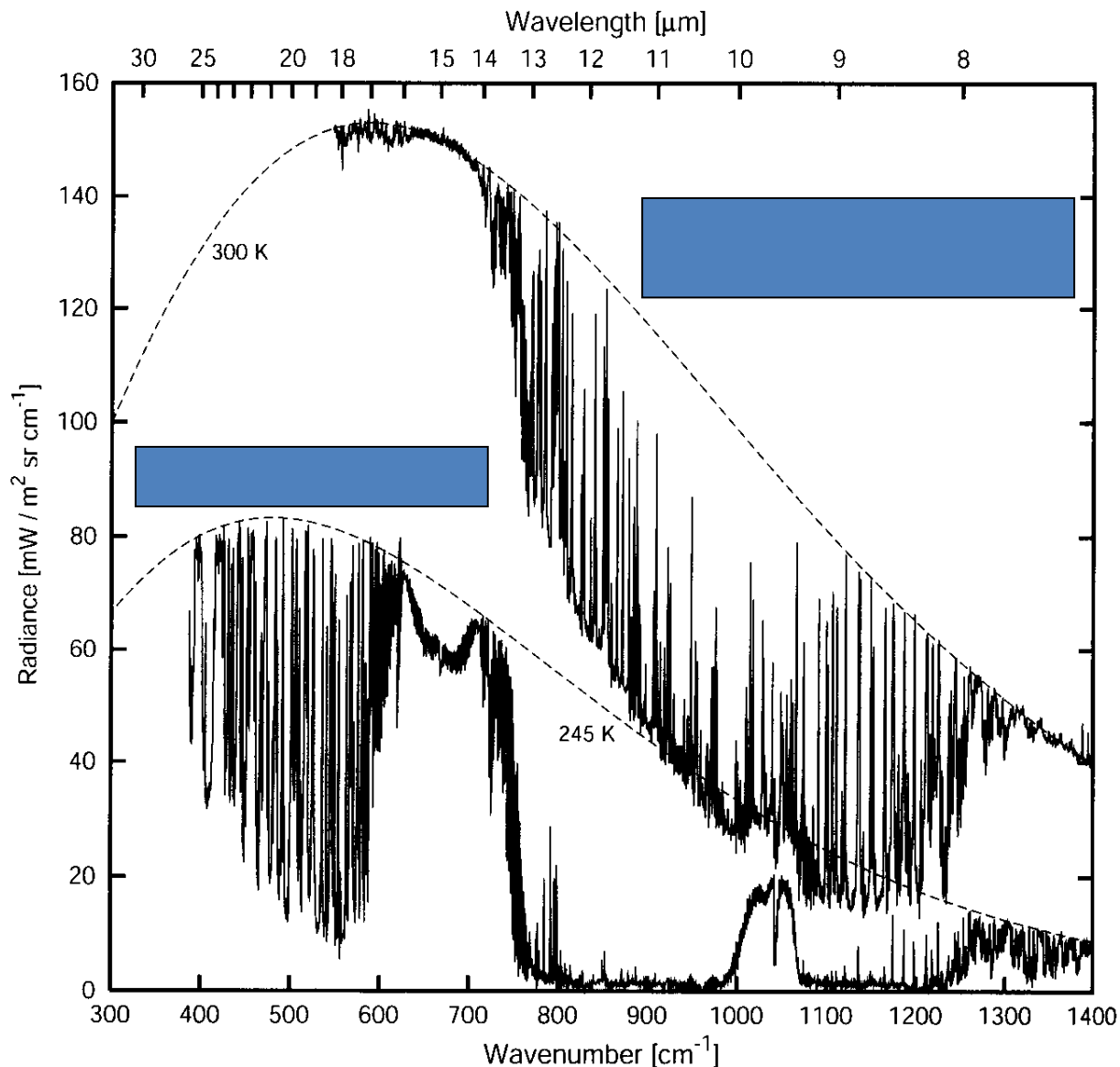
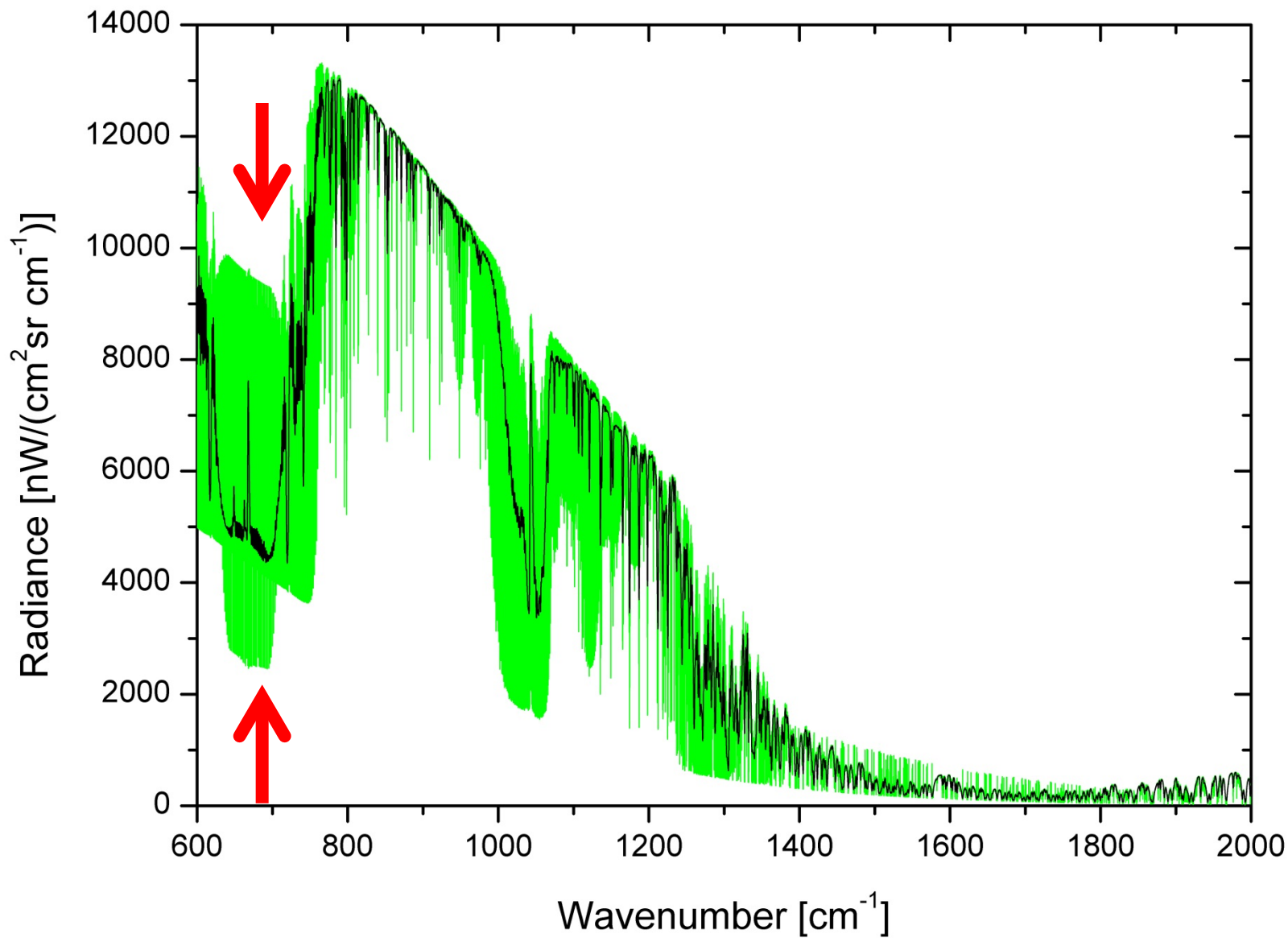
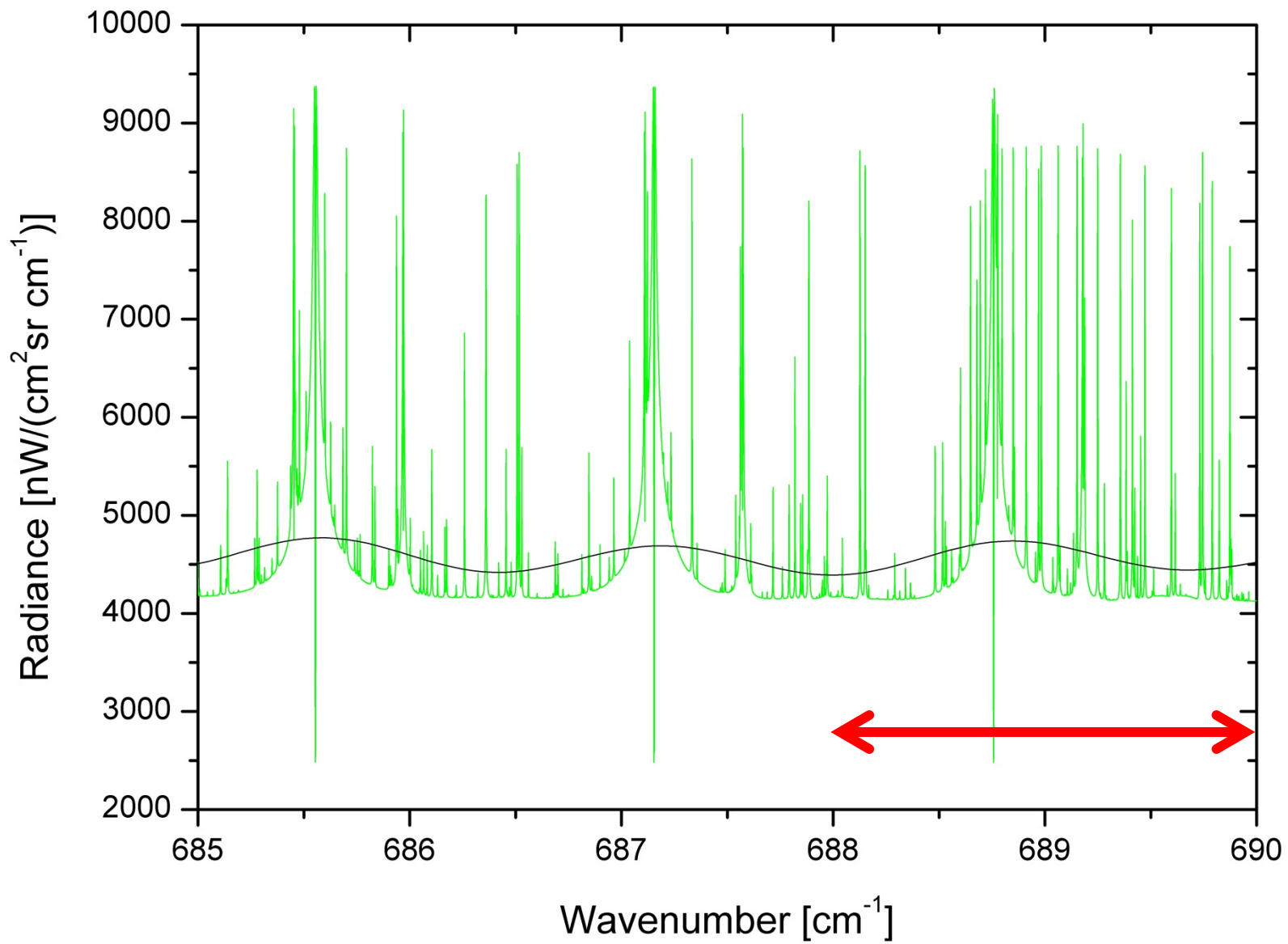
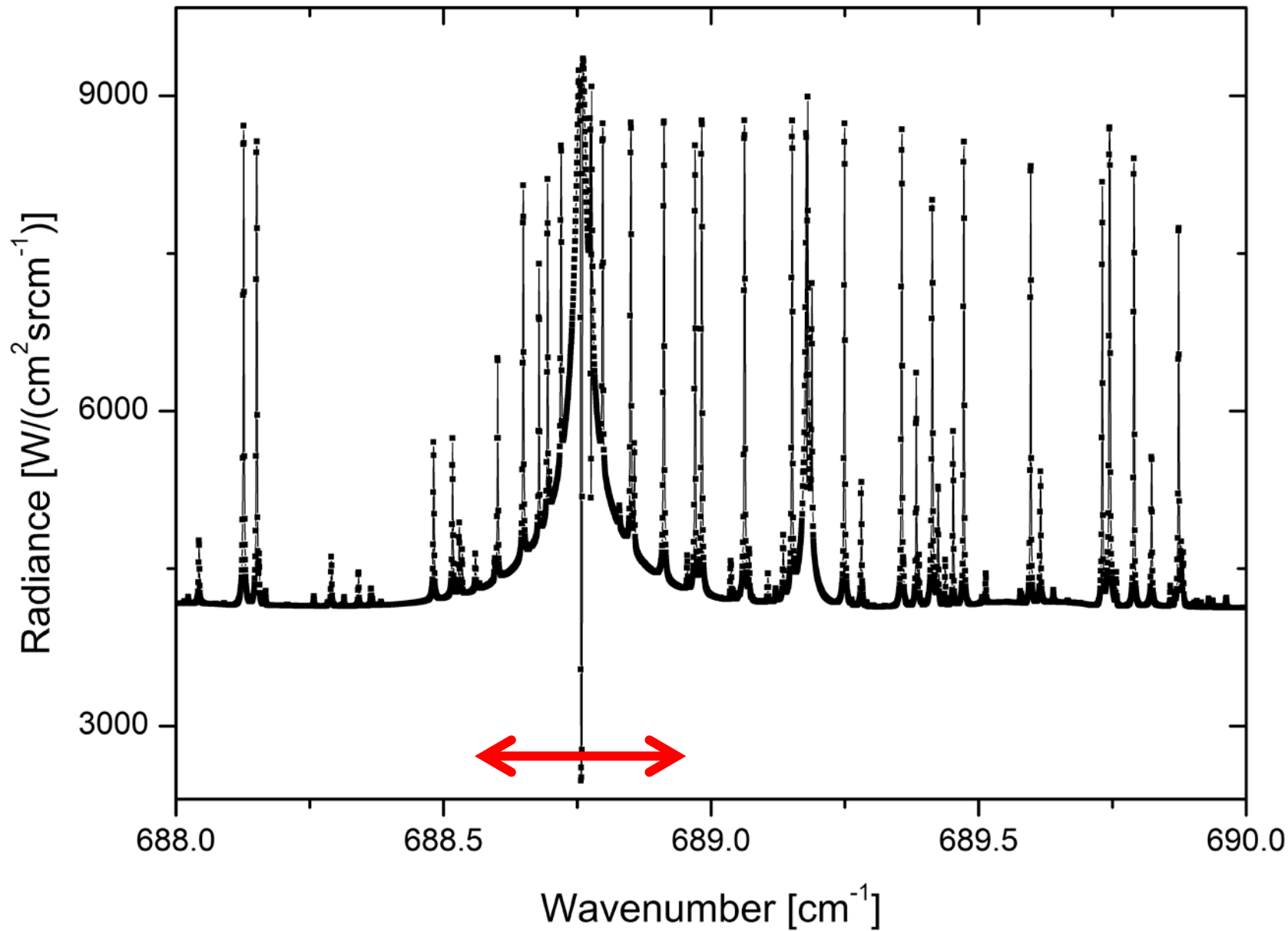


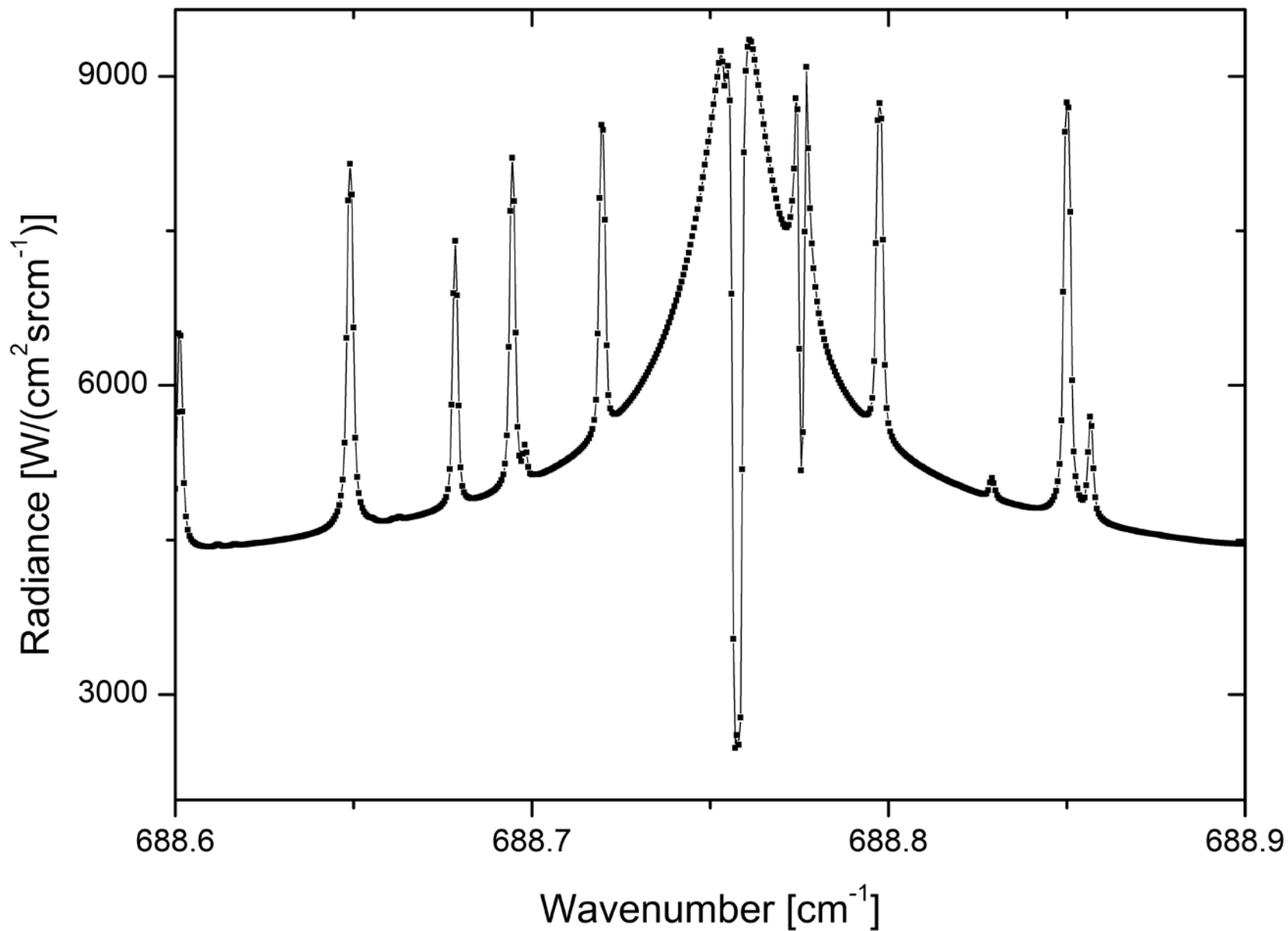
Fig. 8.1: Two examples of measured atmospheric emission spectra as seen from ground level looking up. Planck function curves corresponding to the approximate surface temperature in each case are superimposed (dashed lines). (Data courtesy of Robert Knuteson, Space Science and Engineering Center, University of Wisconsin-Madison.)

'IASI' and 'monochromatic' spectrum

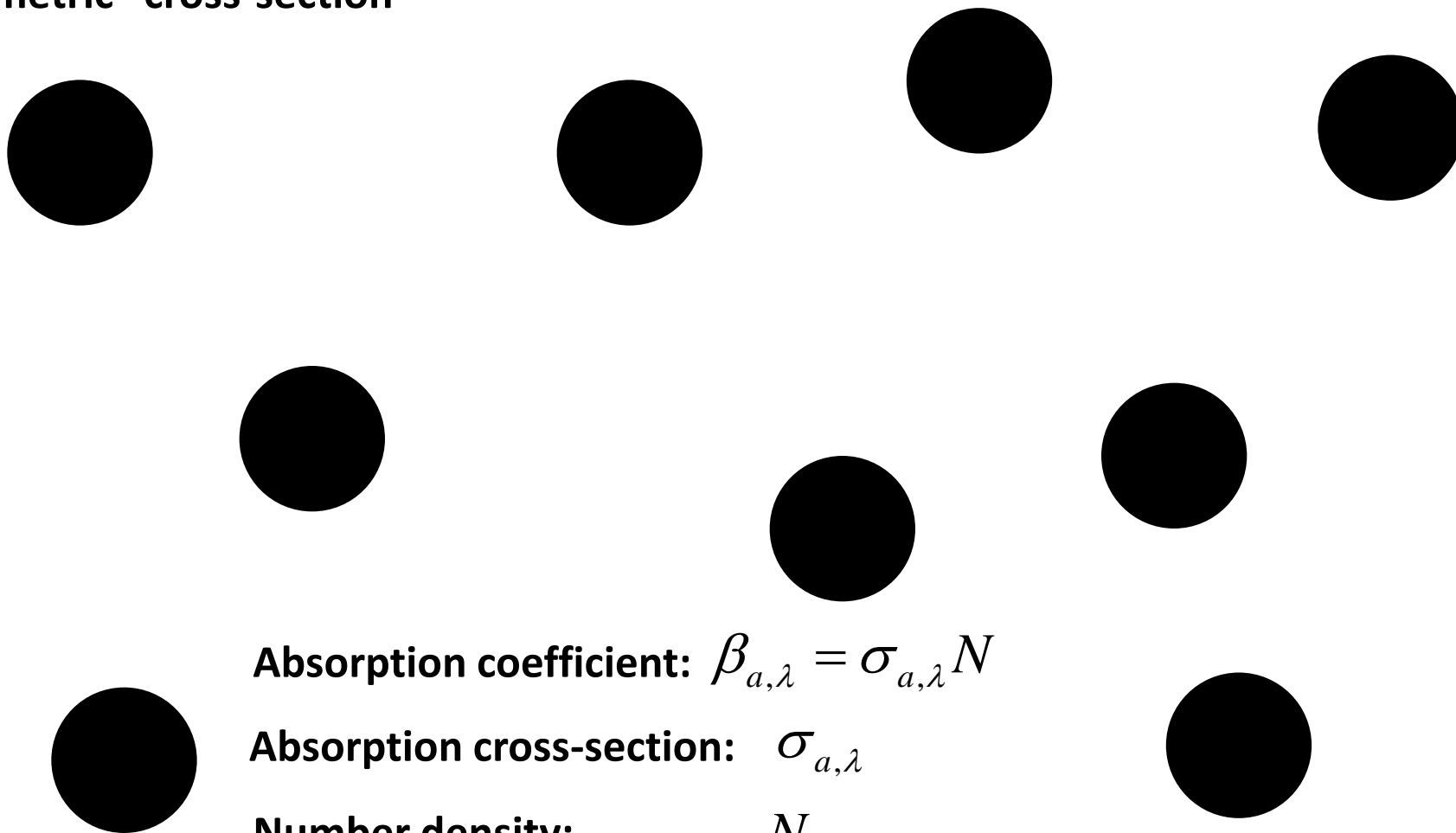








“Geometric” cross-section

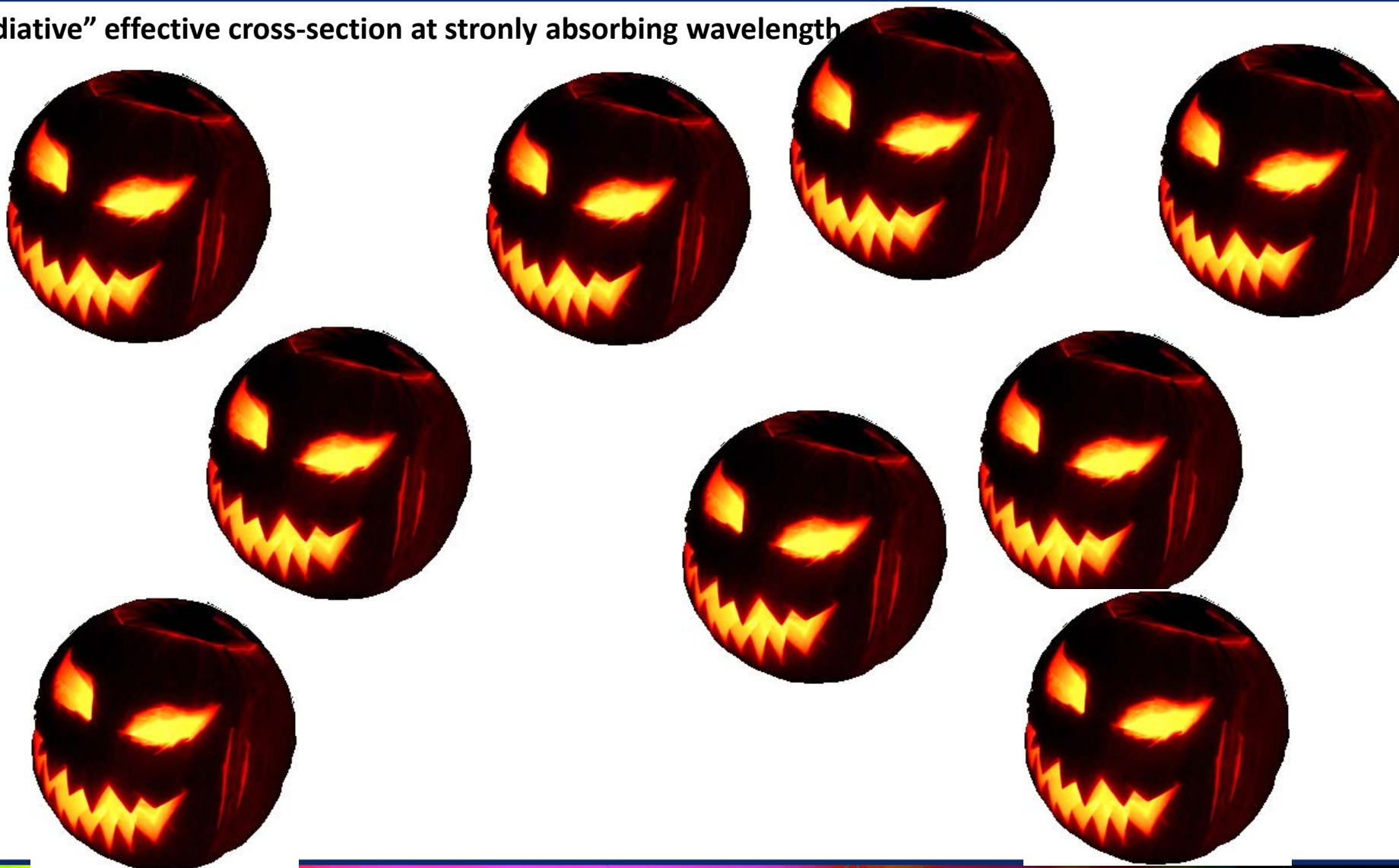


Absorption coefficient: $\beta_{a,\lambda} = \sigma_{a,\lambda} N$

Absorption cross-section: $\sigma_{a,\lambda}$

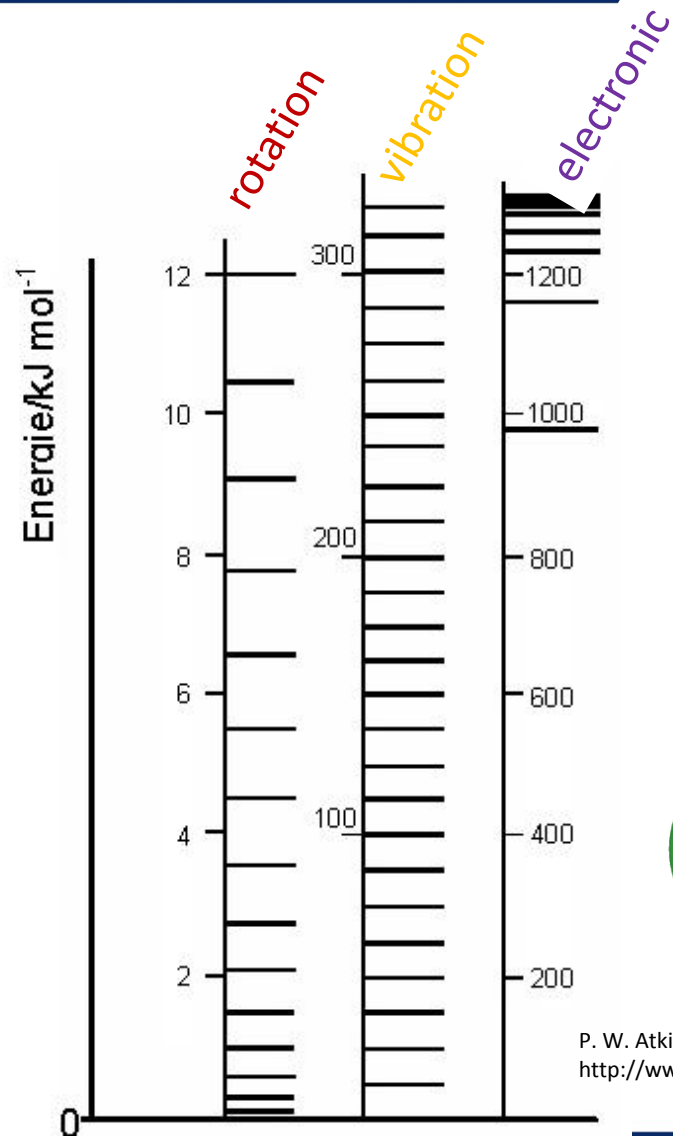
Number density: N

“Radiative” effective cross-section at strongly absorbing wavelength

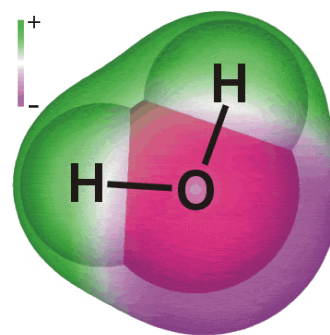


“Radiative” effective cross-section at weakly absorbing wavelength

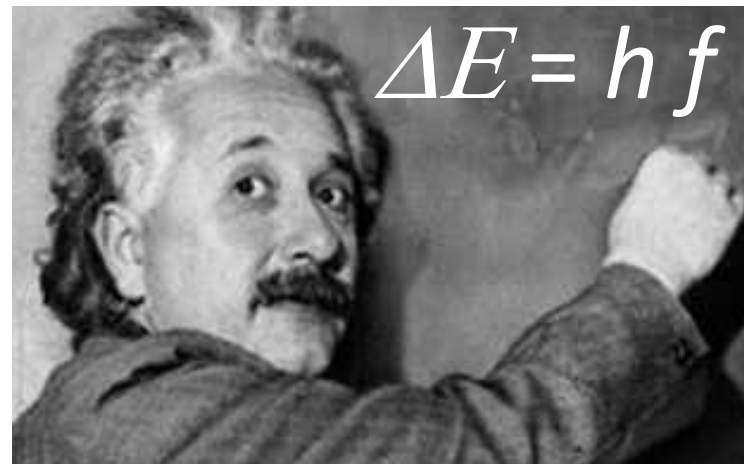




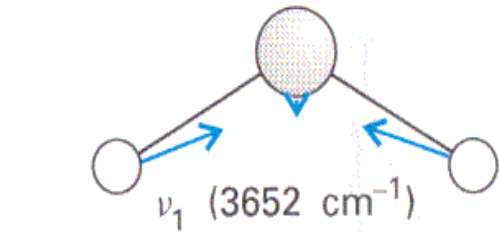
- Quantum mechanics: a bound microscopic system can only be in distinct rotational/vibrational/electronic states.
- Transfer from one to the other state can occur through emission/absorption of electromagnetic radiation (photon)
- **Microwave: rotation of a molecule with static dipole moment**
- **IR: vibration of a molecule; changing dipole moment**
- **UV-VIS: electronic transitions**



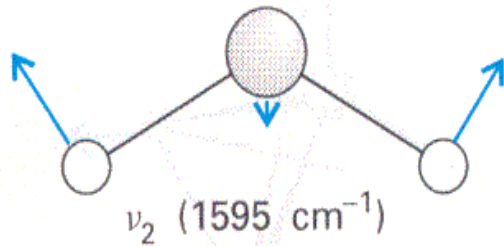
P. W. Atkins, Physikalische Chemie, 1996
<http://www.polymere.uni-koeln.de/11572.html>



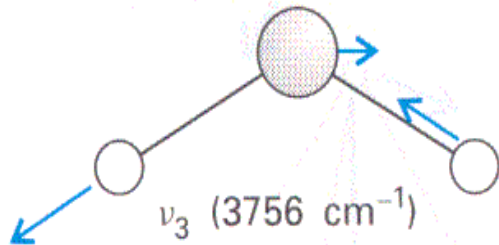
Beispiel: Schwingung und Rotation des H₂O-Moleküls



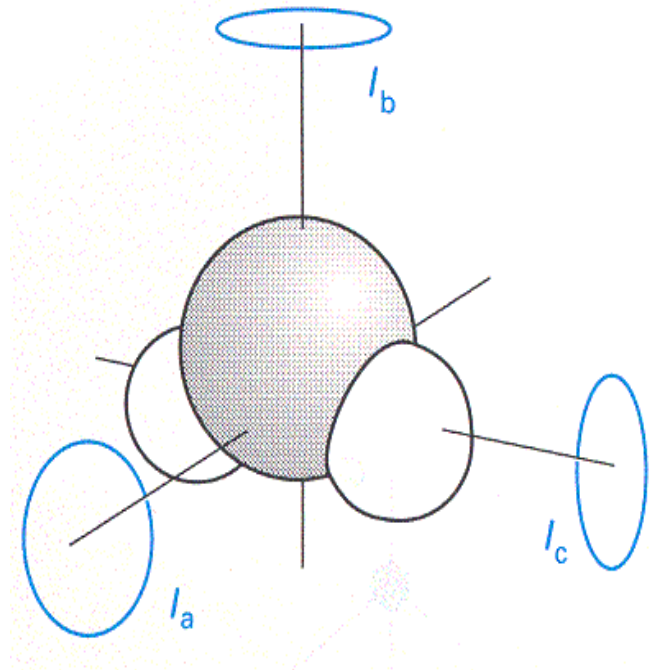
2.74 μm



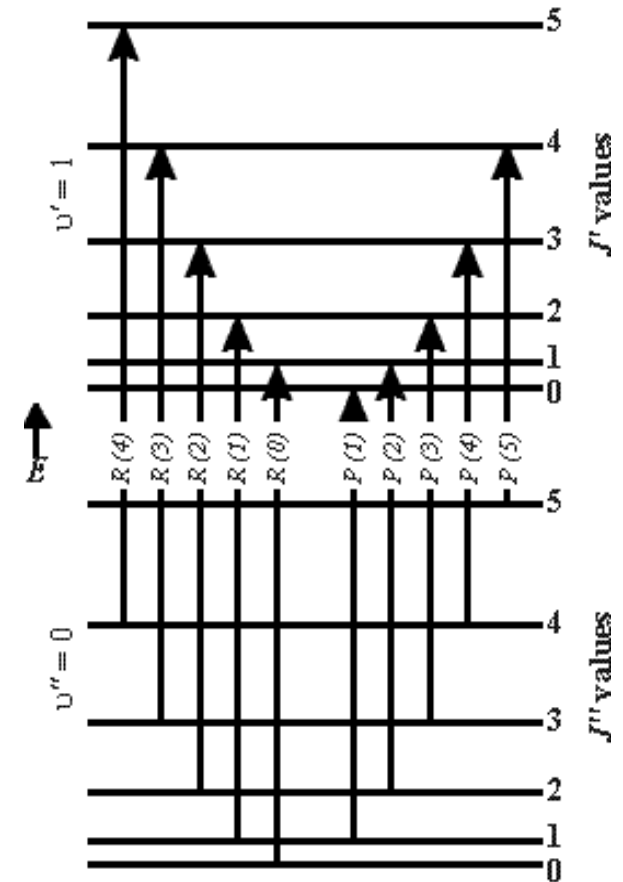
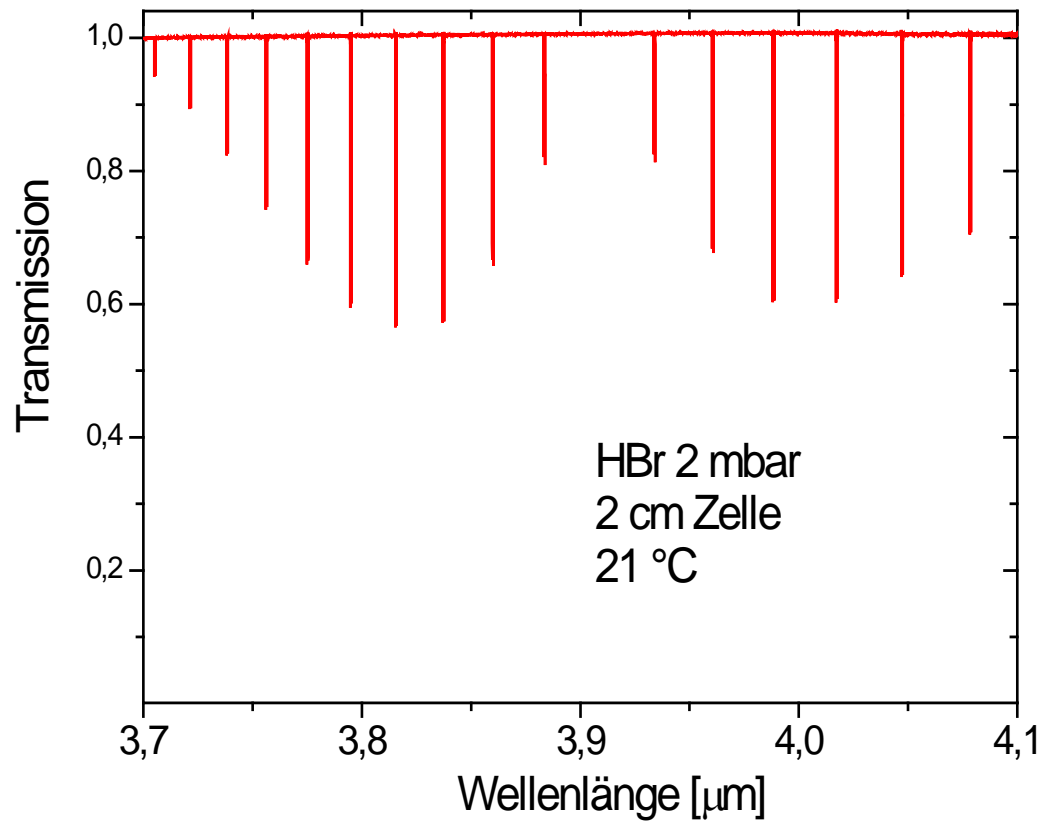
6.27 μm

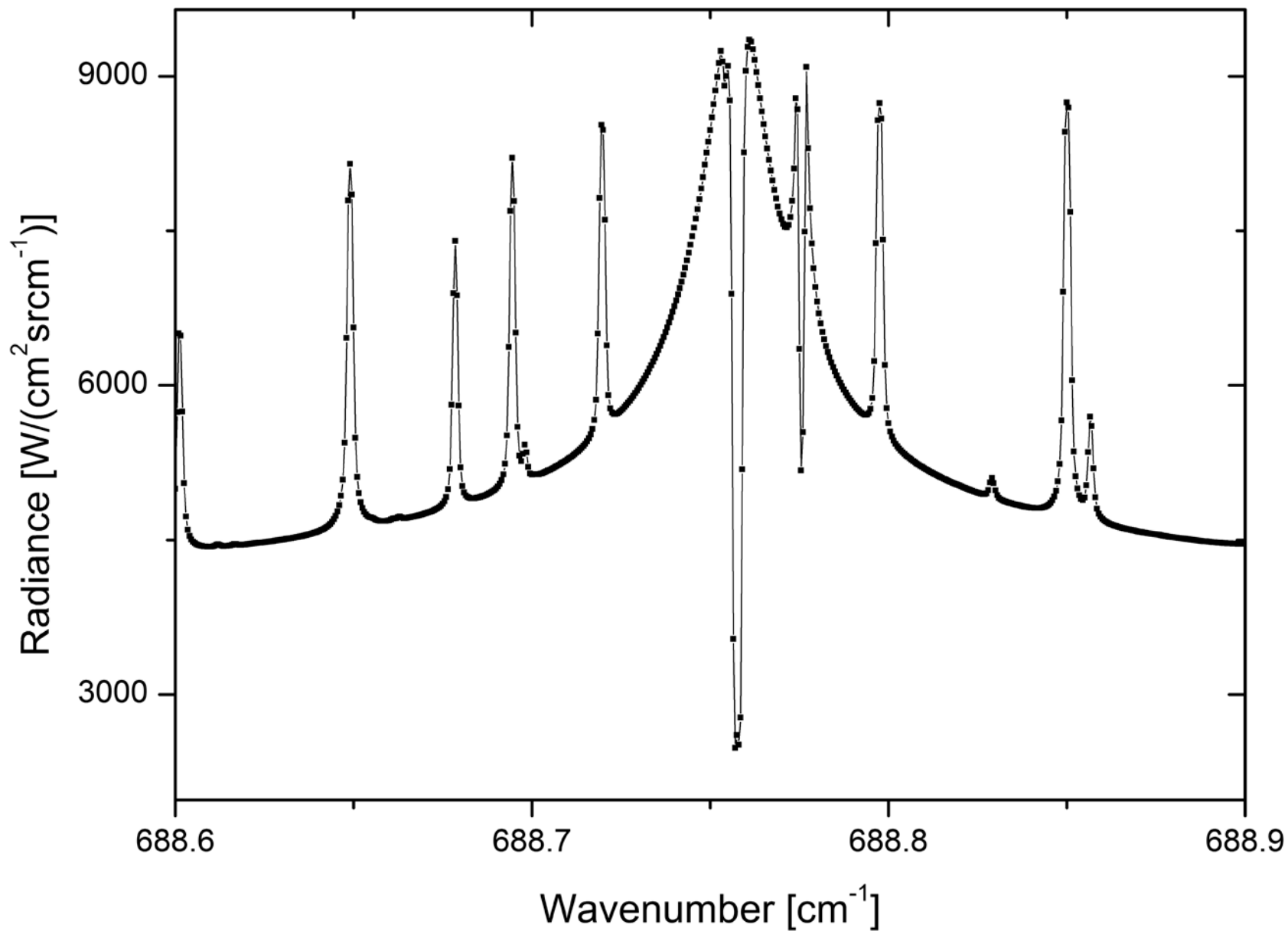


2.66 μm



Ro-vibrational transition in the mid-IR





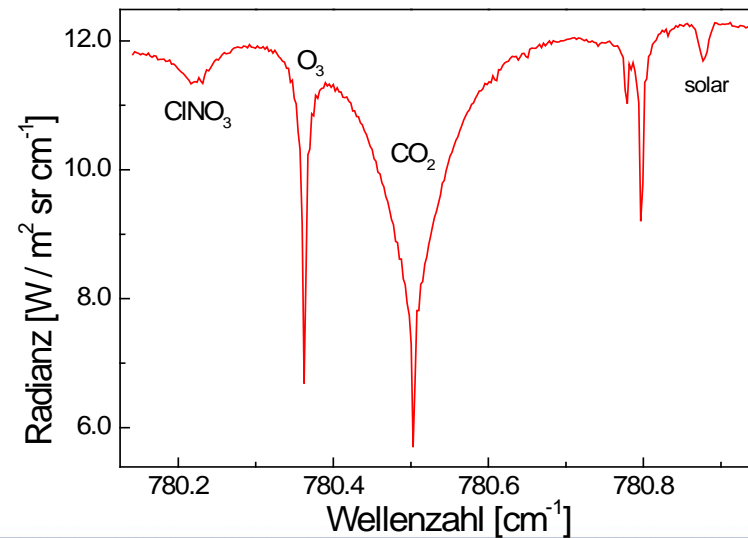
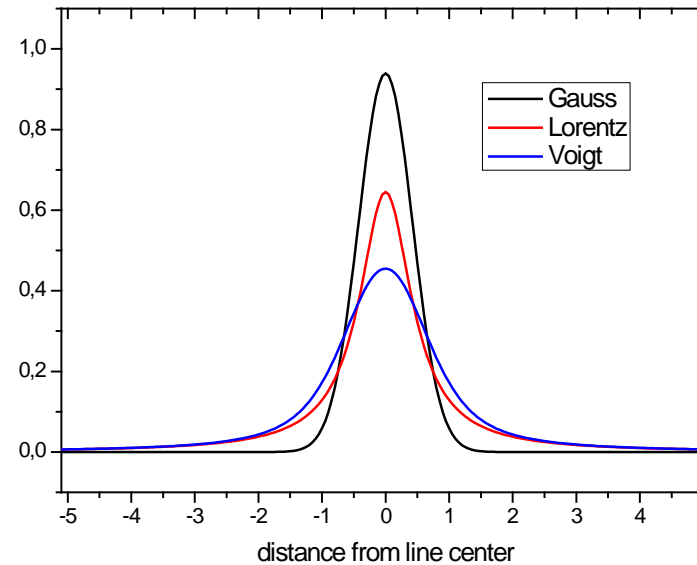
The finite width of spectral lines

Doppler-broadening:

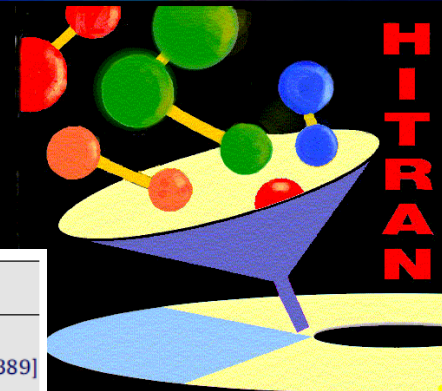
- Thermal movement of molecules along the line-of-sight
- Gaussian shape
- $\sim 0.001 \text{ cm}^{-1}$ @ 1000 cm^{-1}
- Proportional to the frequency

Pressure-broadening:

- Collisions with other molecules disturb their oscillation
- Transition frequency becomes 'blurred'
- Lorentz shape
- $\sim 0.05 \text{ cm}^{-1}$ @ 1000 hPa
- Proportional to the collision rate (pressure)



<http://www.cfa.harvard.edu/hitran/>



Variable	Definition	Units	Comments
<i>Mol</i>	Molecule number	Unitless	Chronological assignment
<i>I_a</i>	Isotopologue number	Unitless	Ordering based on terrestrial values of atoms given in Ref. [389]
<i>v</i>	Transition wavenumber	cm ⁻¹	Line position in vacuum
<i>S</i>	Intensity	cm ⁻¹ /(molecule cm ⁻²)	At 296 K
<i>A</i>	Einstein A-coefficient	s ⁻¹	See Ref. [387]
<i>γ_{air}</i>	Air-broadened half-width	cm ⁻¹ atm ⁻¹	HWHM at 296 K
<i>γ_{self}</i>	Self-broadened half-width	cm ⁻¹ atm ⁻¹	HWHM at 296 K
<i>E'</i>	Lower-state energy	cm ⁻¹	Referenced to zero for lowest possible level
<i>n_{air}</i>	Temperature-dependence exponent of <i>γ_{air}</i>	Unitless	
<i>δ_{air}</i>	Air pressure-induced shift	cm ⁻¹ atm ⁻¹	At 296 K
<i>v', v''</i>	Upper- and lower-state "global" quanta	Unitless	See Table 3 of Ref. [1]
<i>q', q''</i>	Upper- and lower-state "local" quanta	Unitless	See Table 4 of Ref. [1]
<i>i_{err}</i>	Uncertainty indices	Unitless	See Table 5 of Ref. [1]
<i>i_{ref}</i>	Reference indices	Unitless	Pointers to sources in <i>HITRAN</i>
<i>g', g''</i>	Upper- and lower-state statistical weights	Unitless	Includes state-independent factors in <i>HITRAN</i> , see Ref. [387]
<i>Other properties or constants</i>			
<i>Q</i>	Partition sum	Unitless	Function of temperature
<i>h</i>	Planck constant	erg s	$6.62606896(33) \times 10^{-27}$
<i>c</i>	Speed of light	cm s ⁻¹	$2.99792458 \times 10^{10}$
<i>k_B</i>	Boltzmann constant	erg K ⁻¹	$1.3806504(24) \times 10^{-16}$
<i>T</i>	Temperature	K	

Discretisation and measurement error

$$\hat{y} = \hat{F}(\hat{x})$$



$$\vec{y} = \vec{F}(\vec{x}) + \vec{\varepsilon}$$

Error
(spectral noise)

Measurement vector
(spectral channels)

Vector with altitude profiles
(n points)

Radiative transfer model

$$\vec{y} = \vec{F}(\vec{x}) + \vec{\varepsilon}$$

Linearisation

$$\vec{y} = \vec{F}(\vec{x}_0) + \left. \frac{\partial \vec{F}(\vec{x})}{\partial \vec{x}} \right|_{\vec{x}_0} (\vec{x} - \vec{x}_0) + \vec{\varepsilon} = \vec{F}(\vec{x}_0) + \mathbf{K}(\vec{x} - \vec{x}_0) + \vec{\varepsilon}$$

First guess

Jacobi-Matrix (~weighting functions)

$$\mathbf{K} = \begin{pmatrix} \frac{\partial F_1}{\partial x_1} & \cdot & \cdot & \cdot & \frac{\partial F_1}{\partial x_n} \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \frac{\partial F_m}{\partial x_1} & \cdot & \cdot & \cdot & \frac{\partial F_m}{\partial x_n} \\ \frac{\partial F_n}{\partial x_1} & \cdot & \cdot & \cdot & \frac{\partial F_n}{\partial x_n} \end{pmatrix}$$

$$\vec{y} = \vec{F}(\vec{x}_0) + \mathbf{K}(\vec{x} - \vec{x}_0) + \vec{\varepsilon}$$

Case 1: m = n = 1



$$y = f(x_0) + \left. \frac{df(x)}{dx} \right|_{x_0} (x - x_0) + \varepsilon$$

Newton-Iteration:

$$x_{i+1} = x_i + \frac{1}{\left. \frac{df(x)}{dx} \right|_{x_i}} (y - f(x_i))$$

Error estimation:

$$\sigma_x^2 = \frac{1}{\left(\frac{df}{dx} \right)^2} \sigma_y^2$$

$$\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \underbrace{\begin{pmatrix} \left. \frac{dF_1(x)}{dx} \right|_{x_0} \\ \left. \frac{dF_2(x)}{dx} \right|_{x_0} \end{pmatrix}}_{\mathbf{K}} (x - x_0) + \begin{pmatrix} \varepsilon_1 \\ \varepsilon_2 \end{pmatrix}$$

Case 2: m = 2 n = 1

\mathbf{K} is vector and cannot be inverted

Way 1:

1. Calculate x separately for each y as in example 1
2. Calculate error for each x: as in example 1
3. Calculate the result x as weighted mean
4. And the error of x as:

$$\sigma_{x_1}, \sigma_{x_2}$$

$$x = \frac{\frac{x_1}{\sigma_{x_1}^2} + \frac{x_2}{\sigma_{x_2}^2}}{\frac{1}{\sigma_{x_1}^2} + \frac{1}{\sigma_{x_2}^2}}$$

$$\sigma_x^2 = \left(\frac{1}{\sigma_{x_1}^2} + \frac{1}{\sigma_{x_2}^2} \right)^{-1}$$

Way2:

Case 2: $m = 2$ $n = 1$

Least-squares inversion

Minimise the weighted quadratic differences:

$$[\vec{y} - \vec{F}(x)]^T \mathbf{S}_y^{-1} [\vec{y} - \vec{F}(x)] = [\vec{y} - (\vec{F}(x_0) + \mathbf{K}(x - x_0))]^T \mathbf{S}_y^{-1} [\vec{y} - (\vec{F}(x_0) + \mathbf{K}(x - x_0))]$$

Covariance matrix of observations:

$$\mathbf{S}_y = \begin{pmatrix} \sigma_{y_1}^2 & 0 \\ 0 & \sigma_{y_2}^2 \end{pmatrix}$$

Result of minimisation

$$x = x_0 + \frac{1}{\frac{F_1'(x_0)^2}{\sigma_{y_1}^2} + \frac{F_2'(x_0)^2}{\sigma_{y_2}^2}} \left(\frac{F_1'(x_0)}{\sigma_{y_1}^2} (y_1 - F_1(x_0)) + \frac{F_2'(x_0)}{\sigma_{y_2}^2} (y_2 - F_2(x_0)) \right)$$

Same result as in way 1

Case 2: m = 2 n = 1

$$x = x_0 + \frac{1}{\frac{F_1'(x_0)^2}{\sigma_{y_1}^2} + \frac{F_2'(x_0)^2}{\sigma_{y_2}^2}} \left(\frac{F_1'(x_0)}{\sigma_{y_1}^2} (y_1 - F_1(x_0)) + \frac{F_2'(x_0)}{\sigma_{y_2}^2} (y_2 - F_2(x_0)) \right)$$

For non-linear problems this can be written as iteration:

$$x_{i+1} = x_i + \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} \right)^{-1} \mathbf{K}^T \mathbf{S}_y^{-1} \left(\vec{y} - \vec{F}(x_i) \right)$$

Variance

$$\sigma_x^2 = \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} \right)^{-1}$$

General case for m,n

$$\vec{x}_{i+1} = \vec{x}_i + \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} \right)^{-1} \mathbf{K}^T \mathbf{S}_y^{-1} \left(\vec{y} - \vec{F}(\vec{x}_i) \right)$$

$$\mathbf{S}_x = \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} \right)^{-1}$$

Regularization

- More unknowns than observations $n > m$
- $m \geq n$ but linear dependent



Introduction of constraints
(Regularization)

Tikhonov-Phillips regularization

The altitude profile should be “smooth”

Instead of:

$$\left[\vec{y} - \vec{F}(\vec{x}) \right]^T \mathbf{S}_y^{-1} \left[\vec{y} - \vec{F}(\vec{x}) \right]$$

Minimize:

$$\left[\vec{y} - \vec{F}(\vec{x}) \right]^T \mathbf{S}_y^{-1} \left[\vec{y} - \vec{F}(\vec{x}) \right] + \gamma \vec{x}^T \mathbf{L}^T \mathbf{L} \vec{x}$$

$$\mathbf{L} = \begin{pmatrix} -1 & 1 & 0 & 0 & \dots \\ 0 & -1 & 1 & 0 & \dots \\ \cdot & \cdot & \cdot & \cdot & \dots \\ \cdot & \cdot & \cdot & \cdot & \dots \\ \dots & \dots & 0 & -1 & 1 \end{pmatrix}$$

Solution:

$$\vec{x}_{i+1} = \vec{x}_i + \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} + \gamma \mathbf{L}^T \mathbf{L} \right)^{-1} \left[\mathbf{K}^T \mathbf{S}_y^{-1} \left(\vec{y} - \vec{F}(\vec{x}_i) \right) + \gamma \mathbf{L}^T \mathbf{L} \left(\vec{x}_a - \vec{x}_i \right) \right]$$

Linear statistical regularization “optimal estimation”

Use statistical knowledge about atmospheric parameter x

$$\left[\vec{y} - \vec{F}(\vec{x}) \right]^T \mathbf{S}_y^{-1} \left[\vec{y} - \vec{F}(\vec{x}) \right]$$



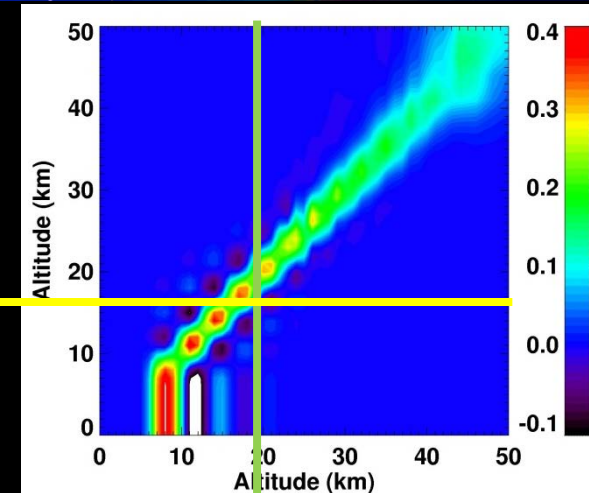
$$\left[\vec{y} - \vec{F}(\vec{x}) \right]^T \mathbf{S}_y^{-1} \left[\vec{y} - \vec{F}(\vec{x}) \right] + (\vec{x} - \vec{x}_a)^T \mathbf{S}_a^{-1} (\vec{x} - \vec{x}_a)$$

Solution:

a-priori covariance matrix

$$\vec{x}_{i+1} = \vec{x}_i + \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} + \mathbf{S}_a^{-1} \right)^{-1} \left[\mathbf{K}^T \mathbf{S}_y^{-1} \left(\vec{y} - \vec{F}(\vec{x}_i) \right) + \mathbf{S}_a^{-1} \left(\vec{x}_a - \vec{x}_i \right) \right]$$

Averaging kernel matrix



MIPAS
CIONO2
Tikhonov
Regul.

Tikhonov-Phillips

$$\mathbf{A} = \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} + \gamma \mathbf{L}^T \mathbf{L} \right)^{-1} \mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K}$$

Optimal estimation

$$\mathbf{A} = \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} + \mathbf{S}_a^{-1} \right)^{-1} \mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K}$$

Column k : answer of the retrieval to a delta-function at altitude k

Row j : contribution of different altitudes to the results in altitude j

Smoothing described by

$$\vec{x}_{inv} = \vec{x}_0 + \mathbf{A}(\vec{x}_{true} - \vec{x}_0)$$

Ozone inversion from ground-based FTIR

$$\vec{y} = \vec{F}(\vec{x})$$

$$K_{ij} = \partial y_i / \partial x_j$$

$$\Delta \vec{y} = \mathbf{K} \Delta \vec{x}$$

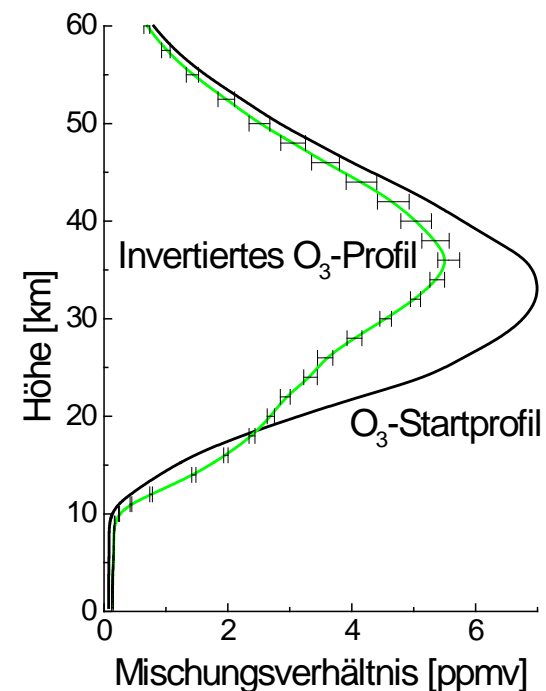
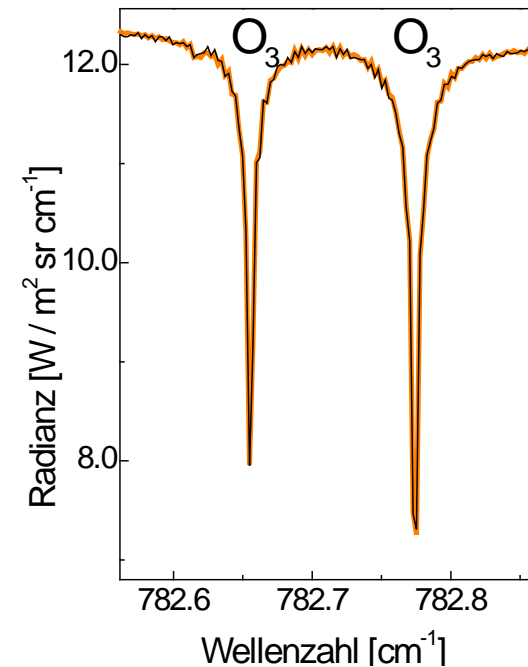
Iterative inversion of linearized problem:

Minimize:

$$\left[\vec{y} - \vec{F}(x) \right]^T \mathbf{S}_y^{-1} \left[\vec{y} - \vec{F}(x) \right] + \gamma \vec{x}^T \mathbf{L}^T \mathbf{L} \vec{x}$$



$$\vec{x}_{i+1} = \vec{x}_i + \left(\mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} + \gamma \mathbf{L}^T \mathbf{L} \right)^{-1} \left[\mathbf{K}^T \mathbf{S}_y^{-1} \left(\vec{y} - \vec{F}(\vec{x}_i) \right) + \gamma \mathbf{L}^T \mathbf{L} \left(\vec{x}_a - \vec{x}_i \right) \right]$$



Averaging kernel and vertical resolution

Ozone from ground-based FTIR

ClONO₂ from MIPAS

