



Physical Simultaneous Retrieval of Emissivity Spectrum and Thermodynamical parameters: A case study for desert soils

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Basic methodological steps to retrieve surface emissivity

Step 1: Represent emissivity with a Fourier cosine truncated series to lower it dimensionality below that of the IASI spectrum.

Step 2: Constrain the retrieval with Laboratory measurements

Step 3: Balance between Atmospheric and Emissivity Constraints with a 2-Dimensional L-curve criterion Basic methodological steps to retrieve surface emissivity

Step 2: Constrain the retrieval with Laboratory measurements

Guido Masiello, Carmine Serio, and Vincenzo Cuomo, "Exploiting Quartz Spectral Signature for the Detection of Cloud-Affected Satellite Infrared Observations over African Desert Areas," Appl. Opt. **43**, 2305-2315 (2004)



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Background Retrieval Methodology: the retrieval problem is formulated within the context of optimal estimation

$$(R - F(v))^{T} S_{\varepsilon}^{-1} (R - F(v)) + (v - v_{a})^{T} S_{a}^{-1} (v - v_{a})$$
Linearize
$$(y - Kx)^{T} S_{\varepsilon}^{-1} (y - Kx) + (x - x_{a})^{T} S_{a}^{-1} (x - x_{a})$$

$$y = R - F(v_{o})$$

$$x = v - v_{o}$$

$$x_{a} = v_{a} - v_{o}$$
How we modify the
Physics to include
surface emissivity
$$K = \left(\frac{\partial F}{\partial v}\right)_{v = v_{o}}$$

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How we modify the Physics to include surface emissivity

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□ Use the *logit* transform to map emissivity from the interval [0,1] to the range [-∞, +∞]

$$y(i) = \text{logit}(\varepsilon(i)) = \log\left(\frac{\varepsilon(i)}{1 - \varepsilon(i)}\right)$$
 $i = 1, ..., N_{ch}; N_{ch} = \text{IASI channels}$

which has the inverse

$$\varepsilon(i) = \frac{\exp(y(i))}{1 + \exp(y(i))} \quad i = 1, \dots, N_{ch}(i)$$

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To retrieve for emissivity, the given radiance R(i), with *i* the channel, is first linearized also with respect to the function y(i), that is we consider in the inverse problem also a linear term of the type

$$\frac{\partial R(i)}{\partial y(i)}(y(i) - y_o(i)) \tag{3}$$

with $y_o(i)$ a suitable first guess. The derivative term can be easily computed when we consider the dependence of the radiance on the surface term,

 $\varepsilon(i)\tau_o(i)B(T_g)$

where $\tau_o(i)$ is the total transmittance at channel *i*, and $B(T_g)$ is the Planck function computed at the ground-surface temperature T_g . We have for the derivative,

$$\frac{\partial R(i)}{\partial y(i)} = \frac{\partial R(i)}{\partial \varepsilon(i)} \frac{\partial \varepsilon(i)}{\partial y(i)} = \frac{\partial R(i)}{\partial \varepsilon(i)} \left(\frac{\partial y(i)}{\partial \varepsilon(i)}\right)^{-1} = \frac{\partial R(i)}{\partial \varepsilon(i)} \varepsilon(i) \left(1 - \varepsilon(i)\right) = \tau_o(i) B(T_g) \varepsilon(i) \left(1 - \varepsilon(i)\right)$$

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Continue.....

Second, we develop the function in a truncated cosine series

$$y(i) = \sum_{k=1}^{N_{cut}} w(k)c(k)\cos\frac{\pi(2i-1)(k-1)}{2N_{ch}}$$
$$w(k) = \begin{cases} \sqrt{\frac{1}{N_{ch}}} & k = 1\\ \sqrt{\frac{2}{N_{ch}}} & k = 2, \dots, N_{cut} \end{cases}$$

where, $N_{cut} \leq N_{ch}$. The Fourier coefficients, c(k) can be obtained by,

$$c(k) = w(k) \sum_{i=1}^{N_{ch}} y(k) \cos \frac{\pi (2i-1)(k-1)}{2N_{ch}} \quad k = 1, \dots, N_{ch}$$
$$w(k) = \begin{cases} \sqrt{\frac{1}{N_{ch}}} & k = 1\\ \sqrt{\frac{2}{N_{ch}}} & k = 2, \dots, N_{ch} \end{cases}$$

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Continue...... and Inserting the truncated cosine transform within the linear term (3), we get

$$\frac{\partial R(i)}{\partial y(i)} \left(\sum_{k=1}^{N_{cut}} w(k) c(k) \cos \frac{\pi (2i-1)(k-1)}{2N_{ch}} - \sum_{k=1}^{N_{cut}} w(k) c_o(k) \cos \frac{\pi (2i-1)(k-1)}{2N_{ch}} \right)$$

which defining the jacobian matrix,

$$A_{ik} = \frac{\partial R(i)}{\partial y(i)} w(k) \cos \frac{\pi (2i-1)(k-1)}{2N_{ch}}, \quad i = 1, \dots, N_{ch}; k = 1, \dots, N_{cut}$$

allows us to rewrite the linear term (2) in a matrix form,

$$\mathbf{A}(\mathbf{c}-\mathbf{c}_{o}) \tag{4}$$

Which is suitable for inversion. Note that in (4) the cosine coefficients vectors have size N_{cut} , and the matrix **A** has size $N_{ch} \times N_{cut}$.

The Fourier transform allows us to work in terms of spectral resolution, exactly the way we deal with this concept. For sea emissvity, 60 Fourier Coefficients are enough



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...but, if we want to resolve the Quartz Resthralen bands in desert soil we need more than 200 Fourier Coefficients



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$$y = R - F(v_{o})$$

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$$x_{a} = v_{a} - v_{o}$$

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From the size of background constraint we introduce information from laboratory measurements

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Laboratory emissivity is used for the background, mean and covariance matrix assumed to be diagonal (ASTER-Salisbury data base)



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The whole covariance matrix for atmospheric parameters and emissivity is built up in block-diagonal matrix



 γ_1 and γ_2 can be optimized to balance between the two terms. Balancing is obtained with an original and fully analytical implementation of a 2-D L-curve criterion

Fully 2-D L-curve method, outline of the mathematics involved

$$L(\gamma_1, \gamma_2) =_{def} \begin{cases} z = \varphi(\gamma_1, \gamma_2) = (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) \\ x = \psi(\gamma_1, \gamma_2) = \hat{\mathbf{u}}^t \hat{\mathbf{I}}_{\gamma_1} \hat{\mathbf{u}} \\ y = \omega(\gamma_1, \gamma_2) = \hat{\mathbf{u}}^t \hat{\mathbf{I}}_{\gamma_2} \mathbf{u} \end{cases}$$
(1)

$$\kappa(\gamma_1, \gamma_2) = \frac{|\det P|}{w^4} \tag{2}$$

with

$$w^{2} = 1 + \left(\frac{\partial z}{\partial x}\right)^{2} + \left(\frac{\partial z}{\partial x}\right)^{2}$$
(3)

and where $|\det P|$ is the absolute value of the determinant of the matrix, **P**, whose elements, P_{ij} are

$$P_{ij} = \frac{\partial^2 z}{\partial x_i \partial x_j} \tag{4}$$

with i, j = 1, 2 and $x_1 = x, x_2 = y$. The derivatives above can be obtained by a transformation of partial derivatives of φ, ψ, ω with respect the two regularization parameters, γ_1, γ_2 . Introducing the notation,

$$\begin{cases} X_i = \frac{\partial X}{\partial \gamma_i}; & i = 1, 2\\ X_{ij} = \frac{\partial^2 X}{\partial \gamma_i \partial \gamma_j}; & i, j = 1, 2 \end{cases}$$
(5)

to indicate the derivative of a given function X with respect to γ_1 and γ_2 , we have

$$\begin{cases} \frac{\partial z}{\partial x} = g = \frac{\varphi_1}{\psi_1} + \frac{\varphi_2}{\psi_2}\\ \frac{\partial z}{\partial y} = g^* = \frac{\varphi_1}{\omega_1} + \frac{\varphi_2}{\omega_2} \end{cases}$$
(6)

And for the second derivatives,

$$\begin{cases} \frac{\partial^2 z}{\partial x^2} = \frac{g_1}{\psi_1} + \frac{g_2}{\psi_2}; & \frac{\partial^2 z}{\partial x \partial y} = \frac{g_1^*}{\psi_1} + \frac{g_2^*}{\psi_2} \\ \frac{\partial^2 z}{\partial y^2} = \frac{g_1^*}{\omega_1} + \frac{g_2^*}{\omega_2}; & \frac{\partial^2 z}{\partial y \partial x} = \frac{g_1}{\omega_1} + \frac{g_2}{\omega_2} \end{cases}$$
(7)



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$$\begin{cases} g_1 = \frac{\varphi_{11}}{\psi_1} - \frac{\varphi_1\psi_{11}}{\psi_1^2} + \frac{\varphi_{12}}{\psi_2} - \frac{\varphi_2\psi_{12}}{\psi_2^2} \\ g_2 = \frac{\varphi_{22}}{\psi_2} - \frac{\varphi_2\psi_{22}}{\psi_2^2} + \frac{\varphi_{21}}{\psi_1} - \frac{\varphi_1\psi_{21}}{\psi_1^2} \\ g_1^* = \frac{\varphi_{11}}{\omega_1} - \frac{\varphi_1\omega_{11}}{\omega_1^2} + \frac{\varphi_{12}}{\omega_2} - \frac{\varphi_2\omega_{12}}{\omega_2^2} \\ g_2^* = \frac{\varphi_{22}}{\omega_2} - \frac{\varphi_2\omega_{22}}{\omega_2^2} + \frac{\varphi_{21}}{\omega_1} - \frac{\varphi_1\omega_{21}}{\omega_1^4} \end{cases}$$
(8)

The notation X_i and X_{ij} for the first and second derivatives allows us to deal with these quantities as component of a vector and a matrix, respectively. This greatly simplifies the software implementation of an algorithm to compute the Gaussian curvature for each given couples (γ_1, γ_2).

At this point we need a scheme to compute the derivatives of φ , ψ , ω . These can be obtained by a direct differentiation of the parametric surface of Eq. 2. Continuing to use the simplified notation also for the derivatives of the vector solution, $\hat{\mathbf{u}}$, we have for thr derivatives of φ ,

$$\begin{cases} \varphi_1 = (\mathbf{G}\hat{\mathbf{u}}_1)^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) + (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}}_1) \\ \varphi_2 = (\mathbf{G}\hat{\mathbf{u}}_2)^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) + (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}}_2) \\ \varphi_{11} = (\mathbf{G}\hat{\mathbf{u}}_{11})^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) + 2(\mathbf{G}\hat{\mathbf{u}}_1)^t (\mathbf{G}\hat{\mathbf{u}}_1) + (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}}_{11}) \\ \varphi_{22} = (\mathbf{G}\hat{\mathbf{u}}_{22})^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) + 2(\mathbf{G}\hat{\mathbf{u}}_2)^t (\mathbf{G}\hat{\mathbf{u}}_2) + (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}}_{22}) \\ \varphi_{12} = (\mathbf{G}\hat{\mathbf{u}}_{12})^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) + (\mathbf{G}\hat{\mathbf{u}}_2)^t (\mathbf{G}\hat{\mathbf{u}}_1) + (\mathbf{G}\hat{\mathbf{u}}_1)^t (\mathbf{G}\hat{\mathbf{u}}_2) + (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}}_{12}) \\ \varphi_{21} = (\mathbf{G}\hat{\mathbf{u}}_{21})^t (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y}) + (\mathbf{G}\hat{\mathbf{u}}_1)^t (\mathbf{G}\hat{\mathbf{u}}_2) + (\mathbf{G}\hat{\mathbf{u}}_2)^t (\mathbf{G}\hat{\mathbf{u}}_{11}) + (\mathbf{G}\hat{\mathbf{u}} - \mathbf{y})^t (\mathbf{G}\hat{\mathbf{u}}_{21}) \end{cases}$$

For the derivatives of ψ , we have

$$\begin{cases}
\psi_{1} = \hat{\mathbf{u}}_{1}^{t}\mathbf{L}_{\gamma_{1}}\hat{\mathbf{u}} + \hat{\mathbf{u}}^{t}\mathbf{L}_{\gamma_{1}}\hat{\mathbf{u}}_{1} \\
\psi_{2} = \hat{\mathbf{u}}_{2}^{t}\mathbf{L}_{\gamma_{1}}\hat{\mathbf{u}} + \hat{\mathbf{u}}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{2} \\
\psi_{11} = \hat{\mathbf{u}}_{11}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}} + 2\hat{\mathbf{u}}_{1}^{t}\mathbf{L}_{\gamma_{1}}\hat{\mathbf{u}}_{1} + \hat{\mathbf{u}}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{1} \\
\psi_{22} = \hat{\mathbf{u}}_{22}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}} + 2\hat{\mathbf{u}}_{2}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{2} + \hat{\mathbf{u}}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{22} \\
\psi_{12} = \hat{\mathbf{u}}_{12}^{t}\mathbf{L}_{\gamma_{1}}\hat{\mathbf{u}} + \hat{\mathbf{u}}_{2}^{t}\mathbf{L}_{\gamma_{1}}\hat{\mathbf{u}}_{1} + \hat{\mathbf{u}}_{1}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{2} + \hat{\mathbf{u}}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{12} \\
\psi_{21} = \hat{\mathbf{u}}_{21}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}} + \hat{\mathbf{u}}_{1}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{2} + \hat{\mathbf{u}}_{2}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{2} + \hat{\mathbf{u}}_{1}^{t}\mathbf{I}_{\gamma_{1}}\hat{\mathbf{u}}_{21}
\end{cases}$$
(10)

Because of the symmetry of two terms ψ and ω , the derivatives of ω are obtained by replacing \mathbf{I}_{γ_1} with \mathbf{I}_{γ_2} in Eq. 10.

Finally, it is seen that all the computations above depend on the partial derivatives of the $\hat{\mathbf{u}}$. These can be easily obtained by differentiation of Eq. ??. We have

$$\begin{aligned} \mathbf{A} \hat{\mathbf{u}}_{1} &= & -\mathbf{I}_{\gamma_{1}} \hat{\mathbf{u}} \\ \mathbf{A} \hat{\mathbf{u}}_{2} &= & -\mathbf{I}_{\gamma_{2}} \hat{\mathbf{u}} \\ \mathbf{A} \hat{\mathbf{u}}_{11} &= & -2\mathbf{I}_{\gamma_{1}} \hat{\mathbf{u}}_{1} \\ \mathbf{A} \hat{\mathbf{u}}_{22} &= & -2\mathbf{I}_{\gamma_{2}} \hat{\mathbf{u}}_{2} \\ \mathbf{A} \hat{\mathbf{u}}_{12} &= & -\mathbf{I}_{\gamma_{1}} \hat{\mathbf{u}}_{2} - \mathbf{I}_{\gamma_{2}} \hat{\mathbf{u}}_{1} \\ \mathbf{A} \hat{\mathbf{u}}_{21} &= & -\mathbf{I}_{\gamma_{2}} \hat{\mathbf{u}}_{1} - \mathbf{I}_{\gamma_{1}} \hat{\mathbf{u}}_{2} \end{aligned}$$
(11)



Retrieval exercise over desert areas





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Simultaneously retrieved with $\epsilon(\sigma)$: T_s, T, H₂O, O₃









Sahara desert

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Kalahari Savanna







Conclusions

- We have developed a physical inverse methodology to retrieve the emissivity spectrum simultaneously with Surface Temperature and Atmospheric parameters: Temperature, water vapour and ozone. The methodology relies mainly on three basic ideas
 - Develop the emissivity spectrum in a truncated Fourier cosine series
 - Constrain the solution with Laboratory measurements
 - Balance the optimal estimation final product with a 2-Dimensional L-curve criterion
- A test retrieval exercise with IASI observations over desert area shows that the retrieved emissivity spectrum is capable to capture the fine details of the surface emission, even with a non committal background covariance matrix for emissivity
- The methodology will be soon applied to derive maps at global scale of the emissivity spectrum.

Set of parameters retrieved with φ -IASI

Simultaneously

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- Emissivity spectrum
- Skin Temperature
- Temperature profile
- Water vapour profile
- Ozone profile

CO
 CO₂
 CH₄
 N₂O

Sequentially, column amount