

living planet symposium BONN 23-27 May 2022

TAKING THE PULSE OF OUR PLANET FROM SPACE

EUMETSAT CECMWF

A novel hybrid machine learning phasor-based approach to retrieve a full set of solar-induced fluorescence metrics and biophysical parameters from model simulations and field data

Prof. Laura Sironi

June 26TH 2022

ESA UNCLASSIFIED – For ESA Official Use Only



Background



Signal acquired by spectrometers:

$$R_{APP}(\lambda) = \frac{L^{\uparrow}(\lambda)}{L^{\downarrow}(\lambda)} = R(\lambda) + \frac{F(\lambda)}{L^{\downarrow}(\lambda)}$$

 $L^{\uparrow}(\lambda) = R(\lambda)L^{\downarrow}(\lambda) + F(\lambda)$

Top of canopy total radiance Canopy reflectance at-surface downwelling radiance Fluorescence signal







Our targets:

1- Retrieving of F spectrum from R_{app}
 2- Estimation of other fluorescence-related variables



THE EUROPEAN SPACE AGENCY → THE EUROPEAN SPACE AGENCY

Background



Signal acquired by spectrometers:

$$R_{APP}(\lambda) = \frac{L^{\uparrow}(\lambda)}{L^{\downarrow}(\lambda)} = R(\lambda) + \frac{F(\lambda)}{L^{\downarrow}(\lambda)}$$

 $L^{\uparrow}(\lambda) = R(\lambda)L^{\downarrow}(\lambda) + F(\lambda)$

Top of canopy total radiance Canopy reflectance at-surface downwelling radiance Fluorescence signal

R_{app} spectrum



Labs



Leaf Area Index (LAI) Chlorofyll Content (Cab) F Quantum Yield (Fqe) Absorbed Photosynthetic Active Radiation (APAR)

F and F at photosystem level (F_{RC})

Consiglio Nazionale delle Ricerche

COSA SPACE

European Space Agency

SOLUTION:

Machine Learning + Phasor Approach =

i-φ-MaLe



Standard Phasor Approach





- In hyperspectral images, each pixel is associated to a single spectrum



💻 📰 📲 🚍 💳 🕂 📲 🔚 📲 🔚 📲 🔚 📲 🔚 🚛 🚳 🍉 📲 🚼 💶 📾 🏜 🛊 🔸 🔹 the European space agency

Standard Phasor Approach





- (g,s) position is dependent on the spectrum functional shape.



💻 📰 📲 🚍 💳 🛶 📲 🔚 📲 🔚 📲 🔚 🚛 🚳 🛌 📲 🛨 📰 📾 🏜 🝁 🔹 🗰 ன 👘 🔹 The European space agency

Standard Phasor Approach





- Each spectrum is transformed in a single point (g,s) on the phasor plot.

- (g,s) position is dependent on the spectrum functional shape.

Transforming phasor approach in a **predictive model**:

#1 Training set of R_{app} spectra, simulated through a RT model
#2 R_{app} divided in spectral windows (windowed phasor approach)
#3 Statistical-based retrieval phase



💳 📰 📲 🚍 💳 🛶 📲 🔚 📰 📰 📲 🔚 📲 🔚 🔤 🐜 🚳 🌬 📲 🛨 📰 📾 🏜 🝁 🔹 этне European space Agency

Novelty #1: R_{app} simulations for Training set





Novelty #2: Windowed phasor approach





Novelty #2: Windowed phasor approach





💳 💶 📲 🚍 💳 🛶 📲 🔚 📲 🔚 📲 🚍 🛻 🚳 🛌 📲 🗮 🖿 🖬 📲 🗮 🖿 🖬











→ THE EUROPEAN SPACE AGENCY





Training Spectrum	Осс	D	
Training #1	0+0	0+0	
Training #2	0+1	0+2.3	
Training #3	0+0	0+0	

Daulina

- Occ. and D are computed for N spectral windows.

$$Occ. = \sum_{i=1}^{N} \alpha_i$$

$$D = \sum_{i=1}^{N} \left(\frac{r}{d_i}\right) \alpha_i \qquad \qquad \alpha_i = \begin{cases} 1 & if \ d_i \le r \\ 0 & if \ d_i > r \end{cases}$$

N = # of selected spectral windows









Ranking			
Training Spectrum	Осс	D	
Training #1	0+1	0+1.8	
Training #2	1+1	2.3+3.4	
Training #3	0+0	0+0	

- Occ. and D are computed for N spectral windows.

$$Occ. = \sum_{i=1}^{N} \alpha_i$$

$$D = \sum_{i=1}^{N} \left(\frac{r}{d_i} \right) \alpha_i \qquad \qquad \alpha_i = \begin{cases} 1 & if \ d_i \le r \\ 0 & if \ d_i > r \end{cases}$$

N = # of selected spectra windows



💻 🔜 📕 🚍 💳 🕂 📲 🔚 🔚 🔚 🔚 🔚 🔚 🔚 🔤 🛻 🚳 🍉 📲 🚼 🖬 🛨 📰 📾 🚔 🔶 → The European space agency





Ranking			
Training Spectrum	Occ	D	
Training #1	1	1.8	
Training #2	3	7.2	
Training #3	0	0	

Analyzed R_{app} is associated to the biophysical/fluorescence variables of the training spectrum characterized by: 1) Highest Occ; 2) Highest D; 3)Average of training spectra with highest Occ and D

$$Occ. = \sum_{i=1}^{N} \alpha_i$$

$$D = \sum_{i=1}^{N} \left(\frac{r}{d_i} \right) \alpha_i \qquad \qquad \alpha_i = \begin{cases} 1 & if \ d_i \le r \\ 0 & if \ d_i > r \end{cases}$$

N = # of selected spectra windows



Algorithm Validation





→ THE EUROPEAN SPACE AGENCY

18

Algorithm Validation





→ THE EUROPEAN SPACE AGENCY

19

Field Measurements





Both <u>Canopy and Tower-level</u> experimental data have been analyzed



💳 🔜 📲 🚍 💳 🛶 📲 🔚 🔚 🔚 🔜 📲 💳 🛶 🛐 🖕 📲 🚼 🖬 🖬 ன 🖗 → THE EUROPEAN SPACE AGENCY

Canopy-level Field Measurements



→ THE EUROPEAN SPACE AGENCY



Trends are comparable to the experimentally monitored vegetation status

DOY 57 - 146 : Forage growth DOY 146 – 150: Alfalfa growth DOY 150: cut of Alfalfa DOY 150 – 192: Alfalfa growth

Data Processing

Each point is the mean \pm std. dev. of 40 R_{app} acquired at solar noon ± 15 min



Canopy-level Field Measurements

Seasonal Trends



AlfaAlfa: day 146 SFM (687nm) 6 RRMSE%: 7.7% F [mW/(m² sr nm)] SpecFit (687nm) [(mu Proposed (687nm) SFM (760nm) SpecFit (760nm) SL 4.5 Proposed (760nm) 4.5 2 3.0 1.5 ш 0↓ 640 0.0 200 680 100 720 760 800 50 150 DOY Wavelength [nm]

i-φ-MaLe F trends are compatible with SFM and SpecFit*** results

* Cogliati et al. (2015). Retrieval of sun-induced fluorescence using advanced spectral fitting methods. Remote Sensing of Environment.

** Cogliati et al. (2019). A Spectral Fitting Algorithm to Retrieve the Fluorescence Spectrum from Canopy Radiance. Remote Sensing.



Data Processing

Each point is the mean ± std. dev. of 40 R_{app} acquired at solar noon ± 15min



Canopy-level Field Measurements





Fluorescence Associated with Photosynthesis at Leaf and Canopy Scales. Remote Sensing.

DISAT





24

Main Achievements



i-φ-MaLe is the <i>first algorithm to successfully and simultaneously retrieve:

- Fluorescence Quantum Yield
- Fluorescence spectrum at photosystem level

Moreover, it also estimates:

- Fluorescence spectrum at canopy level
- Crucial biophysical parameters (LAI, Cab, APAR)

Experimental analysis are performed in <u>real time</u> (< 2ms per spectrum), from <u>Top-of-Canopy</u> level measurements (1-3m) to <u>Tower-level</u> acquisitions (~100m – not shown)



Future Perspectives



- Exploitation of <u>3D Radiative Transfer Models</u>
- Wide training set to simulate more experimental scenarios (e.g. varying LIDF)
- Application to hyperspetral images and satellite-level measurements





Acknowledgements





Biophysics Lab. Physics Dept., Unimib



Laura Sironi, Riccardo Scodellaro Giuseppe Chirico, Maddalena Collini, Laura D'Alfonso, Margaux Bouzin

Further information & Contacts



Remote Sensing Lab. DISAT, Unimib

Ilaria Cesana Roberto Colombo Sergio Cogliati



HE SPACE for ESA Marco Celesti



Sa ESA-ESTEC Dirk Schuettemeyer



IBIMET Franco Miglietta

Scodellaro et al. (2022), A novel hybrid machine learning phasor-based approach to retrieve a full set of solar-induced fluorescence metrics and biophysical parameters. Remote Sensing of Environment (under review).

laura.sironi@unimib.it

i-φ-MaLe method





Score Index

Amount of informative content characterizing the spectral window

Configuration 2

<u>*O*</u>₂ bands are excluded, in order to avoid atmospheric oxygen absorption effects on the Rapp spectra.



= ___ +- ||| := ___ ||| ||| || = += ___ ||| ||| || + COPEAN SPACE AGENCY

Tower-level Field Measurements



→ THE EUROPEAN SPACE AGENCY



Seasonal Trends

No negative F values (differently from inversion-based methods)

Algorithm Configuration 2 - O2 bands exclusion

 Higher uncertainties in F and F_{RC} retrievals (lower fluorescence contribute in Solar Fraunhofer Lines*)

Data Processing

mean \pm std. dev. of R_{app} acquired at solar noon ± 15 min

* Mohammed et al. (2019). Remote sensing of solar-induced chlorophyll fluorescence (SIF) in vegetation: 50 years of progress. Remote Sensing of Environment.

