

Overview on existing protocols for vegetation monitoring



Dr. Laura Mihai

Photonic Investigations Lab. - Center for Advanced Laser Technologies, National Institute for Laser, Plasma and Radiation Physics, Romania



SHORT OVERVIEW

WG3 Objective 3.1 To provide homogenized protocols and methods for synergistic

use of multi-sensor approaches.

(i) sensor calibration and characterization

(ii) sensor employment (iii) data fusion SENSEC SENSEC V WORKING GROUPS V NETWORKING V JOIN US V NEWS

Announcement Kick-off meeting WG3 SENSECO - Sensor synergies

Date: Tuesday February 05, 2019

Time: 09:00 -13:00

Location: Brno, Czech Republic, hotel BWP Premier (https://www.hotelinternational.cz/en/)

Meeting description

Multi-sensor or synergistic sensor use will provide a deeper insight into the relations between spectral features and associated plant conditions. Working Group 3 aims to help realize synergies between passive optical EO domains.

- Introduction into the topic
- Overview on the expertise and sensors used from the WG3 participants
- On which concrete topics do we like to work on? Plant stresses with different sensors? Which kind of stresses?
- Will we use or even acquire (as part of a training for example) a common data set from different sensors?
- What will be the next event?

First aims and deliverables for WG3 are best practice protocols and guidelines for synergistic use of multi-sensor approaches.

From the SENSECO MC, a budget has been foreseen for 8 persons for which travelling can be reimbursed. The selection of these persons will be based on order of registration by sending an email to miriam.machwitz@list.lu stating name and organization.

Minutes of the meeting can be downloaded here.



MEASUREMENTS TRACEABILITY



SENSOR CALIBRATION AND CHARACTERIZATION

<u>Spatial</u> performances



Temporal performances

Radiometric

Performances -Signal to noise ratio -Absolute radiometric accuracy -Polarization sensitivity

-Stray light sensitivity

Spectral performances

- Spectral sampling interval (SSI)
- Spectral resolution (SR)
- Spectral coregistration (smile)

Establishing metrological traceability for radiometric calibration of earth observation sensor in Malaysia. Available from:

https://www.researchgate.net/publication/309593558_Establishing_metrological_traceability_for_radiometric_calibration_of_earth_observation_sensor_in_Malaysia

<u>Geometric</u> Performances - Spatial sampling

SENSOR CALIBRATION AND CHARACTERIZATION

Search 📴 IS	SO/TS 19159-3:2018(en) ×									
ISO/TS 19159-3:2018(en) Geographic information — Calibration and validation of remote sensing imagery sensors and data — Part 3: SAR/InSAR										
Table of contents	٩.									
Foreword A										
Introduction	1 Scope									
1 Scope										
2 Normative references	I his document defines the calibration of SAR/InSAR sensors and validation of SAR/InSAR calibration information.									
3 Terms and definitions	This document addresses earth based remote sensing. The specified sensors include airborne and spaceborne SAR/InSAR sensitive	sors.								
4 Symbols, abbreviated terms and	This document also addresses the metadata related to calibration and validation									
4.1 Symbols										
4.2 Abbreviated terms										

Optical Radiation Metrology and Uncertainty

By Manal A. Haridy and Affia Aslam

DOI: 10.5772/intechopen.75205

017 Reviewed: February 12th 2018 Published: March 14th 2018





📉 remote sensing

Article

NISTHB 157

Guidelines for

Calibration of

Electro-Optical

Instruments for

Remote Sensing

Joe Tansock, Daniel Bancroft, Jim

Mlynczak, Tom Murdock, James

Peterson, David Pollock, Ray Russell,

Deron Scott, John Seamons, Tom Stone,

Butler, Changyong Cao, Raju Datla, Scott Hansen, Dennis Helder, Raghu Kacker, Harri Latvakoski, Martin

Radiometric

Optimized Spectrometers Characterization Procedure for Near Ground Support of ESA FLEX Observations: Part 1 Spectral Calibration and Characterisation

Laura Mihai^{1,*}¹, Alasdair Mac Arthur², Andreas Hueni³, Iain Robinson⁴ and Dan Sporea¹

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Received: 31 December 2017; Accepted: 10 February

Abstract: The paper presents two procedures for absorption spectral bands (O2-A, λ_c = 687 nm spectrometers used for reflectance and Sun-induc and Ar pen-type spectral lamps were employed monochromator setup. The double monochromator errors associated with different operating config

Calibration of a Field Spectroradiometer

MDPI

Calibration and Characterization of a Non–Imaging Field Spectroradiometer Supporting Imaging Spectrometer Validation and Hyperspectral Sensor Modelling

Improving the Calibration of Airborne Hyperspectral Sensors for Earth Observation

> Dissertation zur Erlangung der naturwissenschaftlichen Doktorwürde (Dr. sc. nat.) vorgelegt der Mathematisch-naturwissenschaftlichen Fakultät der Universität Zürich von Karim Lenhard aus

> > Deutschland

Dissertation

zur **g der naturwissenschaftlichen Doktorwürde** (Dr. sc. nat.)

vorgelegt der ematisch–naturwissenschaftlichen Fakultät der Universität Zürich

> von **Michael Ellert Schaepman** von Zürich / ZH

SENSOR CALIBRATION AND CHARACTERIZATION

Raw Data Level 0B Level 1B w AW¹ AW1 m-2 TrackAir FMS Data Applanix Consistency **INS Data** Check Position and Attitude Archived Transformation Instrument Data, Raw Cryogenic radiometer Trap detector Quicklooks Filter radiometer Calibrated Data Instrument Data Download Calibration Parameters Data Segregation Quality Reports Synchronization Assimilation Sensor Attribute Bad Pixel IFC Calibration Processor Detection Data Parameter Files Data Calibration: Laboratory - Electronic smear. C alib ration Calibration Radiometric Processor Laboratory Parameter Files response, Wm² sr¹ nm¹ Calibration Data Bad pixels, Optical effects K Vicarious Calibration Independent Vicarious Vicarious Vicarious Vicarious Validation Quality Report Calib ration Parameter Files Calibration Data Calibration Methodology for the Airborne Dispersive Pushbroom Imaging Spectrometer (APEX) Eutectic blackbody Model helicopter Spectrometer REFERENCE SPECTROMETRY FOR CALIBRATION OF OPTICAL EARTH Jens Nieke, Johannes W. Kaiser, Daniel Schläpfer, Jason Brazile, Klaus I. Itten **OBSERVATION SYSTEMS** (RSL), Horst Schwarzer, Peter Strobl (DLR), Michael E. Schaepman (WUR), S. G. R. Salim^{a,*}, N. P. Fox^a, E. R. Woolliams^a. R. Winkler^a, H. M. Pegrum^a, T. Sun^b, K. T. V. Grattan^b Gerd Ulbrich (ESA) ^a National Physical Laboratory, Hampton road, Teddington, Middlesex, TW11 0LW, UK - (saber.salim, nigel.fox, emma.woolliams, rainer.winkler, heather.pegrum)@npl.co.uk and the APEX team ^b City University, North Hampton Square, London, EC1V 0HB, UK - (T.Sun, K.T.V.Grattan)@city.ac.uk

Radiometric Spectral Geometric

UNCERTAINTIES ANALYSIS GUIDELINES



Intermediate Uncertainty Analysis for Earth Observation Instrument Calibration Module



Training Course Textbook Emma Woolliams Andreas Hueni Javier Gorroño

EMRP

The steps to an uncertainty budget:

- Understanding the problem
- Step 1: Describing the traceability chain
- Step 2: Writing down the calculation equations
- Step 3: Considering the sources of uncertainty
- Determining the formal relationships
- Step 4: Creating the measurement equation
- Step 5: Determining the sensitivity coefficients

Step 6: Assigning uncertainties

- **Propagating the uncertainties** Step 7: Combining and propagating uncertainties
- Step 8: Expanding uncertainties





Optimise papers \rightarrow protocols for SIF measurements

- Paper 1: Sun-induced chlorophyll fluorescence I: Instrumental considerations for proximal spectroradiometers
- Paper 2: Sun-induced chlorophyll fluorescence II: Review of passive measurement setups, protocols and their application at leaf to canopy level
- Paper 3: Sun-induced chlorophyll fluorescence III: benchmarking retrieval methods and sensor characteristics for proximal sensing

Aims to provide a better understanding on some possible instrumental sources of error on the retrieval of SIF, more specifically for the latest generation sensors available and on a single SIF retrieval method adapted to their features.

Steps:

- → Simulation of a typical field spectroradiometer used for retrieval of SIF in O2-A and O2-B absorption bands.
- → Sensor characterization → signal changes to different environmental variables (e.g., temperature, θ s...) or the sensor configuration.
- → Were determined the calibration coefficients → propagating the uncertainties in the process.



Paper 1

→instrument characterization

Paper 1 →instrument characterization

- Spectral Calibration Uncertainties
- Temperature-Induced Spectral Changes
- Radiometric Calibration Uncertainties
- Temperature-Induced Changes in Photoresponse
- Non-linearity
- Directional response characterization



Chamber temperature effects on central position of the spectral band. Blue dots represent observations, whereas Biharmonic interpolated values are represented with surfaces.

Review the current approaches to measure F from leaf to canopy scale from ground based and airborne platforms: instrumentation, measurement setups, protocols, quality checks and data processing strategies.

Paper structure:

 \rightarrow Review on measuring F on:

 \rightarrow the leaf level: setups and protocols

 \rightarrow Canopy level: setups and protocols used from the proximal to the airborne scale

→ Current approaches to open challenges of F estimations from proximal to airborne scale: atmospheric influences (at different altitudes), data quality check (before analysis), metadata documentation, influence of the spatial measurement scale, development of computer models for radiative transfer of fluorescence Paper 2 → measurement setups and protocols

Paper 3

Paper 3 → retrieval methods

Guideline on how to pre-evaluate possible SIF retrieval accuracies for *F*R (F emitted at the red region of the spectrum) and *F*FR (F emitted at the far-red region of the spectrum) considering the characteristics of novel instrumentation (i.e. *SR*, spectral sampling interval (*SSI*), signal-to noise ratio (*SNR*)) in combination with frequently applied retrieval schemes (i.e. sFLD, 3FLD, iFLD, and SFM).

Paper is divided in two section:

- most commonly used retrieval methods and outline their advantages and disadvantages → evaluation of uncertainties in SIF retrievals (i.e. FLD approaches and Spectral Fitting Method, SFM) for a combination of state-of-the-art spectrometers (i.e. ASD, MAYA, HR4000, QE Pro).
- sensitivity analysis including used simulated dataset and methodology to benchmark the performance of the retrieval methods under different sensor characteristicsevaluation of effects when the retrieval methods are implemented in a wrong way and the increase of uncertainty when retrieving SIF



Down-welling irradiance ($E\downarrow$, black line) at ground level and Solar-induced chlorophyll fluorescence spectrum (red line). Maximum value of fluorescence at 685 nm (*maxF685*) and 740 nm (*maxF740*). Sun-induced fluorescence at the O2B (*F687*) and O2A (*F760*) absorption bands

SENSORS EMPLOYMENT

Beskydy, 2017, 10 (1, 2): 75–86 © Mendelova univerzita v Brně ISSN: 1805-9538 (Online) http://dx.doi.org/10.11118/beskyd201710010075



remote sensing

In situ data supporting remote sensing estimation of spruce forest parameters at the ecosystem station Bílý Kříž

*Lucie Homolová¹, Růžena Janoutová¹, Petr Lukeš¹, Jan Hanuš¹, Jan Novotný¹, Olga Brovkina¹, Rolling Richard Loayza Fernandez¹

¹⁾ Global Change Research Institute CAS, Bélidla 986/4a, 603 00 Brno, (e-mail: homolova.l, lukes.p, janoutova.r, hanus.j, novotny.j, brovkina.o, lo

Abstract: Homolová L., Janoutová R., Lukeš P., Hanuš J., Novoti R. R. 2017: In situ data collection supporting remote parameters at the ecosystem station Bilý Kříž – Besky, Article

> Remote sensing offers an effective way of mapping in situ mapping approaches that are typically accur location and few repetitions. In case of forest ecosyst quantitative parameters or indicators related to for index, leaf pigment content, chlorophyll fluoresce and validation of remote sensing-based methods, how data. The aim of this contribution is to introduce th the framework for the retrieval of forest quantitative spectroscopy data. All measurements were acquire campaign that took place at the Norway spruce domin Silesian Beskydy Mts., Czech Republic) during Au remote sensing data acquisition, the in situ activities for tree 3D modelling, measurements of needle bioc area index measurements and spectral measureme surfaces. Leaf pigments varied between 25.2 and 49.1 4.9 - 10.6 µg cm⁻² for carotenoid content depending

ARTICLE Doi: 10.1038/s41467-017-01530-3 OPEN

COMMUNICATIONS

nature

Mapping functional diversity from remotely sensed morphological and physiological forest traits

Fabian D. Schneider (), Felix Morsdorf (), Bernhard Schmid (), Owen L. Petchey (), Andreas Hueni (), David S. Schimel (), & Michael E. Schaepman ()

Assessing functional diversity from space can help predict productivity and stability of forest ecosystems at global scale using biodiversity-ecosystem functioning relationships. We present a new spatially continuous method to map regional patterns of tree functional diversity

Remote sensing offers an effective way of mapping continuous manner, at larger spatial scales and repeat for Detection of Water Stress Symptoms

Max Gerhards ^{1,2,*}⁽²⁾, Martin Schlerf ¹, Uwe Rascher ³⁽²⁾, Thomas Udelhoven ², Radoslaw Juszczak ⁴, Giorgio Alberti ⁵, Franco Miglietta ⁶⁽²⁾ and Yoshio Inoue ⁷

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- ⁷ Research Center for Advanced Science and Technology, University of Tokyo, Tokyo 153-8904, Japan; yinoue@affrc.go.jp
- * Correspondence: gerhardsm@uni-trier.de; Tel.: +49-651-201-4596

Received: 13 June 2018; Accepted: 16 July 2018; Published: 19 July 2018

check for updates

Abstract: High-resolution airborne thermal infrared (TIR) together with sun-induced fluorescence (SIF) and hyperspectral optical images (visible, near- and shortwave infrared; VNIR/SWIR) were jointly acquired over an experimental site. The objective of this study was to evaluate the potential of these state-of-the-art remote sensing techniques for detecting symptoms similar to those occurring during water stress (hereinafter referred to as 'water stress symptoms') at airborne level. Flights with two camera systems (Telops *Hyper-Cam LW*, Specim *HyPlant*) took place during 11th and 12th June



Supplementary Figure 14: Flowchart visualising the work-flow from remote sensing data to physiological (left) and morphological (right) diversity measures. The functional traits are combined to a three-dimensional trait space. By iterating through the pixels using a moving window approach and changing the extent of the neighbourhood, functional diversity measures can be calculated for many scales.

Fig. 4

Fig. 5

Fig. 6

Fig. 6

Fig. 4

DATA FUSION PROTOCOLS



Benefits of hyperspectral imaging for plant disease detection and plant protection: a technical perspective

 $\begin{array}{l} {\rm Stefan\ Thomas}^{1,2}\cdot {\rm Matheus\ Thomas\ Kuska}^1\cdot {\rm David\ Bohnenkamp}^1\cdot \\ {\rm Anna\ Brugger}^1\cdot {\rm Elias\ Alisaac}^1\cdot {\rm Mirwaes\ Wahabzada}^3\cdot {\rm Jan\ Behmann}^1\cdot \\ {\rm Anne-Katrin\ Mahlein}^{1,4} \end{array}$

overview on hyperspectral sensors for plant disease detection from the laboratory to the field



DATA FUSION PROTOCOLS



ISPRS Int. J. Geo-Inf. 2015, 4, 2472-2495; doi:10.3390/ijgi4042472

ISPRS International Journal of Geo-Information ISSN 2220-9964

www.mdpi.com/journal/jigi/

Article

Towards a Standard Plant Species Spectral Library Protocol for Vegetation Mapping: A Case Study in the Shrubland of Doñana National Park

Marcos Jim énez ^{1,*} and Ricardo D íaz-Delgado ²

Remote Sensing Area, National Institute of Aerospace Technologies (INTA), Ctra. Ajalvir s/n, Torrej ún de Ardoz, 28850 Madrid, Spain

² Remote Sensing and GIS Lab., Do ñana Biological Station, CSIC, Avda. Americo Vespucio, 41092 Sevilla, Spain; E-Mail: rdiaz@ebd.csic.es

* Author to whom correspondence should be addressed; E-Mail: jimenezmm@inta.es;

- → establish standard procedures for the acquisition of plant spectra in the field, based on a number of solid bases for any plant type, but also including flexible procedures depending on the type of plant considered (i.e., forest, shrub, pasture).
- → One of these solid bases for the protocol is to estimate interspecies spectral similarity and intra-species variability by accounting for environmental gradients and phenological changes.
- \rightarrow To demonstrate the usability of the protocol \rightarrow a practical example

ISPRS Int. J. Geo-Inf. 2014, 3, 1003-1022; doi:10.3390/ijgi3031003

OPEN ACCESS

ISPRS International Journal of

Geo-Information

ISSN 2220-9964 www.mdpi.com/journal/ijgi/

Article

Field Spectroscopy Metadata System Based on ISO and OGC Standards

Marcos Jiménez^{1,*}, Magdalena González¹, Alberto Amaro² and Alix Fernández-Renau¹

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² Technical Cabinet, Spanish Agency for International Development Cooperation (AECID), Avda. Reyes Católicos, 4, 28040 Madrid, Spain; E-Mail: alberto.amaro@aecid.es

DATA FUSION PROTOCOLS



Metadata structure for field spectroscopy based on ISO 19156 (O&M), ISO 19115 (MD) and Sensor Model Language (SensorML).

PROTOCOLS FOR EXPERIMENTS

Experimental Protocol for Field trials assessing drought stress

Field Lay Out

The experimental fields should be split into three plots. It has to be possible to give different amounts of irrigation water to the three blocks. Water inflow from one block into the other has to be avoided.

In each plot, the experimental clones are planted in blocks of 10 plants in a random complete block design with 3 replicated blocks for each treatment. For trials that also serve to collect modeling parameters, 5 replicate blocks for control and terminal drought have to be planted. 2 of the 5 blocks serve for sequential harvest, 3 for drought susceptibility evaluation. Distance between rows should be 80 cm and between plants 30 cm. We might choose 25 cm plant- and 70 cm row distances in India in order to test the clones under the local conditions.

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No. of Concession, Name									S

Experimental Design to Determine Drought Stress Response and Early Leaf Senescence in Barley (*Hordeum vulgare* L.)



DOI: 10.21769/BioProtoc.1749 Published: Vol 6, Iss 5, March 05, 2016
Plant Science > Plant physiology > Abiotic stress

Plant Science > Plant physiology > Photosynthesi

Plant Science > Plant physiology > Plant growth

CKa Plants > Barley > Leaf > Other compound

Prometheus Wiki



Protocols in ecological & environmental plant physiology

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(Cached)

Field Experiments in Crop Physiology

Kathy Schneebeli * Contributors :Kathy Schneebeli *

12 Votes

Protocols that receive sufficient votes and a high star rating will be considered for Gold Leaf Status by the PrometheusWiki Editorial Board.

Protocols for experimental plot sampling, handling and processing of cereals in field experiments

Standardised methods for use in large agronomic, physiological and genetic field studies

G. Rebetzke^{at}, A. van Herwaarden^b, K. Chenu^c, C. Moeller^d, B. Biddulph^e, R. Richards^a, A. Rattey^a

INTEREST IN MULTI-SENSOR USE

Geophysical Research Abstracts Vol. 21, EGU2019-15718, 2019 EGU General Assembly 2019 © Author(s) 2019. CC Attribution 4.0 license.



Multisensor distribution regression for crop yield estimation

Anna Mateo Sanchis, Jose Adsuara, Maria Piles, Adrián Perez-Suay, Jordi Muñoz-Marí, and Gustau Camps-Valls Universitat de València, Image Processing Laboratory, Image and signal processing, Spain (anna.n Geophysical Research Abstrac

Earth observation (EO) remote sensing data provide a unique source of information temporally resolved and spatially explicit manner. This is of paramount relevance given of biofuels and food. Traditional remote sensing applications have exploited vegetation crop phenology cycles, and have vastly relied on summarizing the time series in a set c descriptors. It is customary to summarize EO time series with temporal metrics like th start/end of season, as well as to summarize all pixel-based observations within a region We posit here that summarizing is not a good idea, and propose two nonlinear regressio all time and space observations that allows blending multisensor (e.g. optical and microv We illustrate the performance of the methods in two scenarios. First, we combine synergi EVI) and microwave (SMAP-VOD) data using full time series stacked at county level into a standard linear and nonlinear (kernel-based) machine learning regression to obtain estimates over the U.S. corn belt. It is shown that the kernel regression outperforms the li the use of full time series from multisensor data improves the results obtained with sta sensors. The second experiment takes into account all goals simultaneously. In this case regression strategy that does not need to summarize the behavior of a county in an averag machine learning method exploits higher-order relations between all time series in a cou the native spatial resolution of each sensor, improves accuracy and bias over previous 1 the validity of the multisensor fusion and the advantage of using distribution regressive series for crop yield estimation.

Geophysical Research Abstracts Vol. 21, EGU2019-16696, 2019 EGU General Assembly 2019 © Author(s) 2019. CC Attribution 4.0 license.

A multi-sensor approach for monitoring vegetation biophysical variables

Gerardo López Saldaña (1), Jose Gomez-Dans (2), Feng Yin (2), Nicola Pounder (1), and Phillip Lewis (2) (1) Assimila Ltd, Reading, United Kingdom (gerardo.lopezsaldana@assimila.eu), (2) University College London, Department of Geography, United Kingdom

Monitoring vegetation biophysical parameters at global scale over a climate timescale (25+ years) is needed to understand long-term land surface processes such as desertification and degradation. In order to create a time series capable of capturing the variability of vegetation and ecosystem properties a multi-sensor approach is needed to generate consistent climate data records.

Using observations from different sensors onboard of different platforms requires a consistent treatment in order to combine data and keep track of the uncertainties along the whole processing chain, since not all sensor will have the same characteristics and radiometric accuracy. This communications uses the MODIS (onboard Terra and Aqua) and OLCI (onboard Sentinel-3) sensors to demonstrate the synergistic approach. The synergistic approach uses the Sensor Independent Atmospheric Correction (SIAC) approach applied to MODIS and OLCI data to derive surface reflectance on a set of common spectral bands. Then all the daily observations are used within a optimal estimation framework to derived BRDF descriptors with associated uncertainties. The final step comprises using Data Assimilation techniques to produced different vegetations parameters. The demonstrator product uses one year of MODIS and OLCI data over different geographic areas with heterogeneous vegetation coverage. The results show that using the two sensors the uncertainties in the vegetation parameters are reduced and that the land surface characterisation is better than using a single sensor, nevertheless the synergistic approach can be applied to different Earth Observation coarse resolution data products such as AVHRR, VIIRS and PROBA-V.

Geophysical Research Abstracts Vol. 21, EGU2019-12877, 2019 EGU General Assembly 2019 © Author(s) 2019. CC Attribution 4.0 license.



An optimized high spatial resolution fluorescence dataset to better understand terrestrial ecosystem dynamics of productivity

Gregory Duveiller (1), Federico Filipponi (1), Sophia Walther (2), Christian Frankenberg (3,4), Philipp Kölher (3), Luis Guanter (5), and Alessandro Cescatti (1)

(1) European Commission Joint Research Centre, Ispra, Italy (gregory.duveiller@ec.europa.eu), (2) Max Planck Institute for Biogeochemistry, Jena, Germany, (3) Division of Geological and Planetary Sciences, California Institute of Technology, Pasadena, CA, USA, (4) Jet Propulsion Lab, California Institute of Technology, Pasadena, CA, USA, (5) Helmholtz Centre Potsdam, German Research Centre for Geosciences, Potsdam, Germany

Mapping and monitoring the spatial and temporal patterns of gross primary productivity (GPP) through the use of satellite remote sensing is of paramount interest to enhance our understanding of terrestrial ecosystem dynamics. While the rate of terrestrial photosynthesis cannot be directly measured from space, research in the last years



Il fluorescence (SIF) retrieved from satellite spectrometers can be a al is very small and notoriously difficult to measure, requiring high hich generally comes at the expense of revisit frequency and spatial decently long time series, such as the widely used GOME-2 data, approximately 50 km), which is too coarse for many applications in will provide information at finer spatio-temporal scales, this doesn't est possible archive is necessary.

thodology, here we provide an optimized SIF dataset at a 0.05° ring the period 2007-2018 with a revisit frequency of 8 days. The GOME-2 SIF retrieval is obtained by locally constraining a light very time step based on different explanatory biophysical variables tatial resolution from other satellite instruments (namely MODIS ig approach, we ensure that this data-driven downscaling follows al understanding of the relationships between GPP, radiation, water roaches based on purely empirical machine learning techniques. An luct is done based on yet another satellite, OCO-2, which provides ry sparse and discontinuous spatial sampling scheme for the period

for improving our understanding of the dynamics in terrestrial it for distinct plant functional types across various gradients of on, precipitation). This enables the characterization of ecosystem unental drivers and stresses, which should further lead to potential

WE STILL NEED PROTOCOLS FOR MULTI-SENSORS RESPONSE SYNERGY!!!!

