



# https://helda.helsinki.fi

# Chlorophyll a fluorescence illuminates a path connecting plant molecular biology to Earth-system science

# Porcar-Castell, Albert

2021-08-09

Porcar-Castell , A , Malenovsky , Z , Magney , T , Van Wittenberghe , S , Fernandez-Marin , B , Maignan , F , Zhang , Y , Maseyk , K , Atherton , J , Albert , L P , Robson , T M , Zhao , F , Garcia-Plazaola , J-I , Ensminger , I , Rajewicz , P A , Grebe , S , Tikkanen , M , Kellner , J R , Ihalainen , J A , Rascher , U & Logan , B 2021 , ' Chlorophyll a fluorescence illuminates a path connecting plant molecular biology to Earth-system science ' , Nature plants , vol. 7 , þÿ n o . 8 , pp. 998 1009 . https://doi.org/10.1038/s41477-021-00980-4

http://hdl.handle.net/10138/339926 https://doi.org/10.1038/s41477-021-00980-4

unspecified acceptedVersion

Downloaded from Helda, University of Helsinki institutional repository. This is an electronic reprint of the original article. This reprint may differ from the original in pagination and typographic detail. Please cite the original version. Chlorophyll-a fluorescence illuminates a path connecting plant molecular biology to Earth system science

3

4	Albert Porcar-Castell <sup>1*</sup> , Zbyněk Malenovský <sup>2</sup> , Troy Magney <sup>3</sup> , Shari Van Wittenberghe <sup>1,4</sup> , Beatriz
5	Fernández-Marín <sup>5</sup> , Fabienne Maignan <sup>6</sup> , Yongguang Zhang <sup>7</sup> , Kadmiel Maseyk <sup>8</sup> , Jon Atherton <sup>1</sup> ,
6	Loren P. Albert <sup>9, 10</sup> , Thomas Matthew Robson <sup>11</sup> , Feng Zhao <sup>12</sup> , Jose-Ignacio Garcia-Plazaola <sup>13</sup> , Ingo
7	Ensminger <sup>14</sup> , Paulina A. Rajewicz <sup>1</sup> , Steffen Grebe <sup>15</sup> , Mikko Tikkanen <sup>15</sup> , James R. Kellner <sup>9, 16</sup> , Janne
8	A. Ihalainen <sup>17</sup> , Uwe Rascher <sup>18</sup> , Barry Logan <sup>19</sup>
9	
10	<sup>1</sup> Optics of Photosynthesis Laboratory, Institute for Atmospheric and Earth System Research/Forest
11	Sciences, Viikki Plant Science Center, University of Helsinki, Helsinki, Finland.
12	<sup>2</sup> School of Geography, Planning, and Spatial Sciences, College of Sciences Engineering and
13	Technology, University of Tasmania, Private Bag 76, Hobart, TAS 7001, Australia.
14	<sup>3</sup> Department of Plant Sciences, University of California, Davis. Davis, CA, 95616 United States of
15	America.
16	<sup>4</sup> Laboratory of Earth Observation, University of Valencia, C/Catedrático José Beltrán, 2, 46980
17	Paterna, Spain.
18	<sup>5</sup> Department of Botany, Ecology and Plant Physiology, University of La Laguna (ULL), Tenerife
19	38200, Spain.
20	<sup>6</sup> Laboratoire des Sciences du Climat et de l'Environnement, LSCE/IPSL, CEA-CNRS-UVSQ,
21	Université Paris-Saclay, Gif-sur-Yvette, France.
22	<sup>7</sup> International Institute for Earth System Sciences, Nanjing University, Nanjing, Jiangsu 210023,
23	China.
24	<sup>8</sup> School of Environment, Earth and Ecosystem Sciences, The Open University, Milton Keynes MK7
25	6AA, United Kingdom.

27	States of America.
28	<sup>10</sup> Biology Department, West Virginia University, Morgantown, WV 26506-6300, United States of
29	America.
30	<sup>11</sup> Organismal and Evolutionary Biology, Viikki Plant Science Centre (ViPS), Faculty of Biological
31	and Environmental Science, 00014, University of Helsinki, Finland.
32	<sup>12</sup> School of Instrumentation Science and Opto-Electronics Engineering, Beihang University,
33	Beijing, 100083, China.
34	<sup>13</sup> Department of Plant Biology and Ecology, University of the Basque Country (UPV/EHU),
35	Bilbao, Spain.
36	<sup>14</sup> Department of Biology, Graduate Programs in Cell & Systems Biology and Ecology &
37	Evolutionary Biology, University of Toronto, 3359 Mississauga Road, Mississauga, ON L5L 1C6,
38	Canada.
39	<sup>15</sup> Molecular Plant Biology, University of Turku, FI-20520 Turku, Finland.
40	<sup>16</sup> Department of Ecology and Evolutionary Biology, Brown University, Providence RI 02912,
41	United States of America.
42	<sup>17</sup> Nanoscience Center, Department of Biological and Environmental Science, University of
43	Jyväskylä, Jyväskylä 40014, Finland.
44	<sup>18</sup> Institute of Bio- and Geosciences, Plant Sciences (IBG-2), Forschungszentrum Jülich GmbH,
45	Jülich, Germany.
46	<sup>19</sup> Biology Department, Bowdoin College, Brunswick, Maine, United States of America.
47	
48	*Corresponding author: joan.porcar@helsinki.fi
49	
50	

<sup>9</sup>Institute at Brown for Environment and Society, Brown University, Providence, RI 02912, United

26

For decades, the dynamic nature of chlorophyll-a fluorescence (ChlaF) has provided insight 51 52 into the biophysics and ecophysiology of the light reactions of photosynthesis from the subcellular to leaf scales. Recent advances in remote sensing methods now enable detection of 53 ChlaF induced by sunlight across a range of larger scales, using instruments mounted on 54 towers above plant canopies to Earth-orbiting satellites. This signal is referred to as solar-55 induced fluorescence (SIF) and its application promises to overcome spatial constraints on 56 studies of photosynthesis, opening new research directions and opportunities in ecology, 57 ecophysiology, biogeochemistry, agriculture and forestry. However, to unleash the full 58 potential of SIF, intensive cross-disciplinary work is required to harmonize these new 59 60 advances with the rich history of biophysical and ecophysiological studies of ChlaF, fostering the development of next-generation plant physiological and Earth system models. Here, we 61 introduce the scale-dependent link between SIF and photosynthesis, with an emphasis on 62 63 seven remaining scientific challenges, and present a roadmap to facilitate future collaborative research towards new SIF applications. 64

When illuminated, chlorophyll-a molecules weakly emit light in the 650-850 nm range; that is, they fluoresce. Steady state<sup>1,2</sup> and time-resolved fluorescence spectroscopy<sup>3,4</sup>, as well as pulse-amplitude modulated (PAM) fluorescence<sup>5,6</sup> have long been used by biophysicists, molecular biologists and ecophysiologists to elucidate the structure and function of the photosynthetic apparatus<sup>7-9</sup>. These techniques are regarded as active because the measured ChlaF originates from a controlled light source, and accordingly have largely<sup>10,11</sup> been restricted to measurements at the subcellular and leaf levels.

Interest in passive remote sensing methods capable of retrieving solar-induced ChlaF across a continuum of spatial scales emerged more than two decades ago<sup>12</sup>. These seminal activities led to the first demonstrations of tower-based<sup>13,14</sup> and satellite<sup>15</sup> SIF measurements over terrestrial ecosystems. The opportunity to remotely detect an energy flux (Box 1) that arises directly from

76 within the photosynthetic process spurred the rapid development of measurement techniques, 77 retrieval protocols, and models for estimating and interpreting SIF across scales. As reviewed in Mohammed et al.<sup>12</sup> and Aasen et al.<sup>16</sup>, SIF can now be measured from an expanding number of 78 sensors mounted on towers<sup>17,18</sup>, drones<sup>19,20</sup>, aircraft<sup>21,22</sup> and satellites with ever-improving spatial 79 and temporal resolution<sup>23,24</sup>. So far, all satellite SIF retrievals have been serendipitous, relying on 80 instruments originally designed to measure atmospheric gases. The first satellite mission designed 81 82 specifically for the measurement of SIF is the ESA FLuorescence EXplorer (FLEX) mission, which is set to launch in  $2024^{25}$ . 83

84 SIF methods are rapidly breaking through the scale bottleneck of traditional ChlaF measurements, opening up a range of new opportunities to study photosynthesis across the continuum of spatial 85 scales from the leaf, through plant canopies, and up to the globe. With SIF we now have the 86 87 potential to illuminate the path connecting plant molecular biology to Earth-system science. However, before the full potential of multiscale SIF observations can be realized, a number of 88 challenges must be overcome. Extracting the information embedded in the SIF signal requires a 89 90 fundamental understanding and a quantitative description of the processes that connect measured ChlaF with photosynthesis (Fig.1), as well as their variation across space and time (Fig. 2). In this 91 Perspective, we present these challenges and propose a roadmap of activities to facilitate future 92 93 research. Finally, we discuss key emerging SIF applications that can benefit from cross-disciplinary expertise. 94

*Challenge 1: APAR<sub>g</sub>*. The common denominator between ChlaF and the photosynthetic uptake of
CO<sub>2</sub> is the flux of photosynthetically active radiation absorbed by photosynthetic pigments, or
APAR<sub>g</sub> (where the g stands for green), which provides the foundation for the mechanistic
connection between SIF and photosynthesis. APAR<sub>g</sub> is the product of the incoming
photosynthetically active radiation (PAR) and the fraction of this PAR absorbed by photosynthetic
pigments (fAPAR<sub>g</sub>) (Fig.1). Importantly, although the absorption of radiation by leaves and plant

101 canopies can be quantified using radiometric sensors either coupled to an integrating sphere<sup>26</sup> (e.g. 102 leaf absorptance profile in Fig.1) or mounted above and below a plant canopy<sup>27</sup>, these 103 measurements also include a significant and dynamic contribution from non-photosynthetic 104 pigments and other canopy elements. While inaccuracies in the estimation of APAR<sub>g</sub> do not disrupt 105 the relationship between SIF and photosynthesis, accurate quantification of the energy flux entering 106 the photosynthetic process is essential for a mechanistic interpretation of SIF and remains a 107 challenge.

# 108 Challenge 2: Distribution of excitation energy between PSII and PSI and their ChlaF emissions.

109

APAR<sub>g</sub> is absorbed mostly by chlorophyll-a and chlorophyll-b associated with either photosystem

II (PSII) or photosystem I (PSI) reaction centres. Interestingly, while both types of chlorophyll have 110 111 the capacity to fluoresce, essentially all chlorophyll fluorescence in vivo originates from 112 chlorophyll-a due to the efficient transfer of excitation energy from chlorophyll-b to chlorophyll-a within light harvesting antennae<sup>28</sup>. Likewise, although both photosystems emit ChlaF, ChlaF from 113 PSII typically dominates the signal, especially in the red region of the emission spectrum<sup>2</sup> (Fig.1), 114 115 and exhibits greater variation in quantum yield in response to photochemical and nonphotochemical processes<sup>7,29</sup>. The dynamic nature of PSII ChlaF explains the widespread application 116 of PAM fluorescence to probe the energy partitioning between photochemical and non-117 photochemical processes or to estimate the rate of linear electron transport (LET) in PSII<sup>30</sup>. 118 However, the estimation of LET requires knowledge of the distribution of absorption between the 119 photosystems (i.e. the use of an energy partitioning factor), which is rarely measured and often 120 assumed to be  $0.5^6$ . Although biochemical and biophysical methods to assess the stoichiometry and 121 antenna sizes of PSI and PSII do exist<sup>31-33</sup>, these methods only provide a relative assessment of the 122 energy distribution; absolute quantification requires the combination of simultaneous ChlaF and 123 820 nm absorption measurements to probe the energy partitioning in PSII and PSI, respectively, 124 along with photosynthetic gas exchange measurements<sup>34</sup>. Overall, the evidence gathered to date 125

suggests that neither the distribution of excitation energy between PSII and PSI nor the contribution
of ChlaF from PSI to SIF remain constant over time, between species or within canopy light
gradients<sup>35,36</sup>. Questions remain: how large is this variability? What controls it? And what is its
significance for the interpretation of SIF? Answers to these questions await the development of
versatile field methods and protocols (e.g. based on rapid optical measurements<sup>37</sup>) to enable
the characterization of these factors across a wide range of conditions.

Challenge 3: Energy partitioning in PSII. Energy absorbed in PSII is partitioned between three 132 main processes: a) photochemical quenching (PQ) of excitation energy, promoting linear electron 133 134 transport, b) non-photochemical quenching (NPQ), which includes both regulated and sustained forms of thermal dissipation, and c) emission of ChlaF. The quantum yield of a process, e.g. ChlaF 135 emission ( $\Phi$ F), can be expressed as the ratio of the rate constant associated to that process relative 136 137 to the sum of all rate constants. Importantly, the rate constants associated to PQ and NPQ are highly dynamic, which allows plants to regulate the flow of energy through PSII and to protect against 138 light-induced damage<sup>38, 39</sup>. During the growing season, the rate constants of PQ and NPQ vary over 139 140 time-scales of seconds to minutes in response to the redox dynamics of the quinone acceptor pool and induction and relaxation of regulated thermal dissipation, respectively. Outside of the growing 141 season, or during periods of profound environmental stress, rate constants can be affected by 142 photoinhibition of PQ and the induction of sustained NPQ. Accordingly, changes in the quantum 143 yield of ChlaF ( $\Phi$ F) reflect the combined effect of PQ and NPQ dynamics and a quantitative 144 145 connection between  $\Phi F$  and  $\Phi P$  (the quantum yield of photochemistry) cannot be established without knowledge of either PQ or NPQ<sup>8,40</sup>. PAM fluorescence uses saturating light pulses to solve 146 the energy partitioning and estimate  $\Phi P$ ; an approach that is not feasible during SIF measurements, 147 precluding partitioning from SIF alone. 148

Under certain conditions, either NPQ or PQ can dominate the relationships between ΦF and ΦP,
resulting in the emergence of a positive or negative relationship respectively. For example, under

low light intensities – when regulatory NPQ remains inactive - the relationship between  $\Phi P$  and  $\Phi F$ 151 152 is negative under the action of PQ, which exerts opposite effects on (i.e. decouples)  $\Phi P$  and  $\Phi F$ . Under high light - when PQ tends to saturate and NPQ is highly active - the relationship between 153  $\Phi P$  and  $\Phi F$  turns to positive under the action of NPQ, which competes for excitation with both (i.e. 154 couples)  $\Phi P$  and  $\Phi F^{12,52,53}$ . The latter case can explain the seasonal correlation between  $\Phi P$  and  $\Phi F$ 155 observed at the leaf<sup>41,42</sup> (Fig. 2) and canopy scales<sup>18</sup>, in response to the modulation of sustained 156 NPQ that protects the foliage from the harmful combination of excessive light and low 157 temperatures<sup>43,44</sup>. Despite the positive relationship between  $\Phi P$  and  $\Phi F$  that emerges in response to 158 certain stress conditions, the quantitative treatment of the energy partitioning in PSII requires the 159 use of mechanistic models and remains one of the core challenges to the interpretation of SIF<sup>40,45,46</sup>. 160

161 Challenge 4: Alternative energy sinks. Photosynthetic linear electron transport provides reducing power for a range of metabolic processes beyond CO<sub>2</sub> assimilation via the Calvin cycle, including 162 chlororespiration<sup>47</sup>, photorespiration<sup>48</sup>, nitrogen, sulphur and oxygen reduction (the latter known as 163 the Mehler reaction in the water-water cycle<sup>49</sup>), and the synthesis of volatile organic compounds<sup>50</sup>. 164 Importantly, the dynamics of these 'non-assimilatory' electron sinks can affect ChIF in a manner 165 not directly correlated with CO<sub>2</sub> assimilation. In particular, because alternative energy sinks can 166 have a protective function by sustaining LET under conditions when CO<sub>2</sub> assimilation is impaired<sup>51</sup>, 167 they could influence the capacity of SIF to detect certain plant stress responses. Therefore, it is 168 critical to address the extent that these dynamics decouple SIF from GPP, in particular during plant 169 stress. As with Challenge 2, answering this question will benefit from the development of versatile 170 field methods and protocols to promote the widespread characterization of these factors across a 171 wide range of conditions. 172

# 173 Challenge 5: Leaf and canopy ChlaF scattering, reabsorption and measurement geometry.

Although the lighter and darker green stripes seen on an athletic field may give the impression ofdifferent chlorophyll contents, they are an optical reflection effect created when the grass is bent in

a particular direction during mowing. SIF measurements over plant canopies are similarly affected 176 by the distribution of leaves, canopy architecture and measurement geometry<sup>27,52</sup>. The amount and 177 distribution of chlorophyll within a leaf (influenced by photosystem and thylakoid structure, 178 179 chloroplast distribution, and internal leaf morphology), as well as the amount and geometrical arrangement of leaves and other non-photosynthetic material within a plant canopy (influenced by 180 branch/stem architecture) drive APAR<sub>g</sub>, connecting SIF and photosynthesis at the leaf and canopy 181 182 scales, respectively. Once emitted, ChlaF photons travel through the same leaf and canopy structures, where some of the ChlaF photons are reabsorbed (Fig. 1 and Fig. 2 "spectral dynamics"). 183 As a result, spatial and temporal variations in leaf biochemistry, leaf morphology, and canopy 184 185 architecture, as well as foliage illumination and viewing geometry, influence the probability of 186 ChlaF photons contributing to a SIF measurement (known as the escape probability). These factors decouple the total emitted ChlaF from the measured SIF, and by extension from photosynthesis. 187 188 Physically-based radiative transfer models, which simulate the movement of photons through leaves and plant canopies (Box 2), can be used to provide a quantitative framework to investigate and 189 account for the impact of these factors on APAR<sub>g</sub> and SIF observations<sup>27,53</sup>. Although spatially 190 explicit RTM approaches are already available (see Supplementary Video 1 and 2), advances in the 191 parametrization of within-leaf and canopy drivers of SIF - e.g. canopy gradients in foliar 192 193 morphology, pigment contents (Challenge 1) or ChlaF contribution from PSI (Challenge 2) - remain areas of active development. 194

195 Challenge 6: Atmospheric absorption and scattering. Atmospheric gases, aerosols and other 196 particles absorb and scatter ChlaF photons traveling from a plant canopy to a remote detector. The 197 extent of atmospheric absorption and scattering of SIF depends on the retrieval wavelength, the 198 distance between target and sensor, and the properties of the atmosphere (Box 1). In particular, SIF 199 retrieval methods based on the in-filling of atmospheric gas absorption bands, such as the O2-A or 200 O2-B bands (Fig. 1), face the challenge that the gas absorption feature used for the SIF retrieval

simultaneously attenuates the ChlaF signal as it travels towards the detector. This effect requires a
 correction even for short-distance measurements from canopy towers and drones<sup>54</sup>. Although an
 atmospheric RTM can be used to characterize and correct for these effects, its application requires
 site-specific measurements of atmospheric profile parameters for model input, which remains an
 operational challenge<sup>55</sup>.

Challenge 7: Integrating SIF controls across space and time. A final challenge, and perhaps the 206 most relevant, is the contextualization of the interpretation of SIF (including the previous six 207 208 challenges) within the spatial and temporal domain of the measurements (Fig. 2). Temporally, ChlaF dynamics have been used to investigate the energy transfer within photosystems (femto-209 picosecond scale)<sup>7,56</sup>, the redox status of the donor and acceptor sides of the photosystem 210 (microsecond-millisecond scale)<sup>3,4</sup>, and the variations in PQ and NPQ (seconds-to-seasonal scale)<sup>38,</sup> 211 <sup>39,44</sup>. Spatially, the intensity and spectral properties of SIF are also controlled by factors that regulate 212 both APAR<sub>g</sub> and ChlaF scattering and reabsorption within a leaf or plant canopy<sup>57,58</sup> (Fig. 2, 213 "spectral dynamics"). When ChlaF is measured as SIF across coarser spatial and longer temporal 214 215 scales, the signal carries information that aggregates an expanding assortment of physical and biological factors<sup>59-61</sup>. New controls may appear while the effects of others may be subordinated, 216 strengthening (via 'couplers'; Fig. 2) or disrupting (via 'decouplers'; Fig. 2) the relationship 217 between SIF and GPP. 218

For example, tower-based SIF studies reveal a strong seasonal linear relationship between canopy SIF and ecosystem GPP across a wide range of ecosystems<sup>17,18,61</sup>, consistent with the coupling action of APAR<sub>g</sub> and NPQ described above. Yet, the sensitivity, strength and linearity of the seasonal SIF-GPP relationship is not universal and has been found to depend on additional physical and physiological decoupling factors, such as sun-vegetation-sensor geometries<sup>62,63</sup>, vegetation canopy structure<sup>52,64</sup>, or photosynthetic pathway (C3 vs. C4)<sup>27,65</sup>, with contrasting responses to different environmental stressors<sup>66,67</sup>. Clearly, integrating and disentangling the relationship between SIF and GPP across species, space, time and in response to environmental stress, remainsstill a challenge that calls for comprehensive field studies.

#### 228 Roadmap towards a consistent interpretation of SIF

The time for multiscale SIF measurements is already here (Fig. 3). Yet, converting these data into meaningful information and new applications still requires effort dedicated to scaling and standardizing methods for SIF interpretation, with particular attention to the seven challenges described above. This process requires accounting for the influence of 1) instrumental, 2) atmospheric, 3) structural and 4) physiological factors to unlock the quantitative association between measured SIF and photosynthesis (Fig. 4). Addressing these challenges requires new data, protocols and models to interpret SIF and bridge the gap between molecular processes, i.e.

236 photosynthesis, and satellite imagery.

At the leaf level, new instruments and techniques employing optical bandpass filters have been 237 developed to record fluorescence spectral dynamics under both natural or controlled illumination, 238 temperature, and CO<sub>2</sub> concentration<sup>16,68-71</sup>. Such spectral approaches, combined with foliar pigment 239 analysis, photosynthetic gas exchange, and PAM ChlaF measurements, provide new insights into 240 the connection between SIF and photosynthesis dynamics of leaves<sup>42,69,72</sup>. Going forward, 241 mechanistically modeling the link between SIF and GPP (Challenges 1-4) will require the 242 combination of field campaigns covering full growing seasons, multiple species and stress 243 responses with detailed experimentation under highly controlled conditions, for example using 244 Arabidopsis mutants with altered photochemical properties<sup>9,73</sup>. In particular, the development of 245 versatile field instrumentation and protocols for the estimation of APARg (Challenge 1), energy 246 247 distribution between PSII and PSI - including the ChlaF contribution from PSI - (Challenge 2), or the quantification of alternative energy sinks (Challenge 4), is key to resolving the spatial and 248 temporal influences of these factors on SIF. 249

250 The synergistic use of complementary data streams can also help to constrain the modelling of 251 photosynthesis and support SIF interpretation. For example, leaf and canopy reflectance data can inform us on the chlorophyll content in the leaf or the amount of leaves in the canopy<sup>74</sup>, relating to 252 APAR<sub>g</sub> (Challenge 1). In addition, reflectance data have been used to explore the regulatory 253 dynamics of NPQ<sup>75</sup>, which could contribute to resolving energy partitioning in PSII (Challenge 3). 254 This approach is feasible due to the spectral change that accompanies the operation of the 255 xanthophyll cycle<sup>76</sup> - by which violaxanthin is converted to antheraxanthin and zeaxanthin in a 256 process that modulates NPQ<sup>38,77</sup> - as well as the seasonal dynamics of leaf carotenoid and 257 chlorophyll contents<sup>78</sup>. These spectral changes, which have been captured by the photochemical 258 reflectance index (PRI)<sup>77,79</sup> or the Chlorophyll/Carotenoid Index (CCI)<sup>80</sup>, are now being revisited 259 and investigated in depth across the whole VIS-NIR region alongside with SIF dynamics<sup>71,76</sup>. 260 Clearly, as in the case of SIF, careful use of canopy and atmospheric RTMs will be needed to 261 262 disentangle these subtle physiologically-induced reflectance changes from those of a dynamic background<sup>55</sup>. In addition to synergies with spectral reflectance, use of thermal imaging<sup>81</sup>, radar<sup>82</sup>, 263 or multispectral laser scanning methods<sup>83</sup> offer interesting possibilities to constrain the carbon 264 reactions of photosynthesis by providing independent information on plant water status (Challenge 265 4). Likewise, leaf and ecosystem-level measurements of carbonyl sulfide (COS) uptake by 266 267 vegetation can provide an independent source of information on stomatal conductance in vascular plants<sup>84</sup>, which could be highly relevant for the development and validation of ecosystem-level SIF-268 GPP models. 269

Process-based and radiative transfer models are required to integrate physical and physiological
mechanisms operating at different scales (Challenge 7), providing excellent frameworks for
multidisciplinary collaborations to connect molecular-level with Earth-system processes. Clearly, as
our mechanistic understanding of the connection between SIF and GPP increases (Challenges 1-4),
so will the accuracy of process-based models. For example, the integration of the Farquhar-

Caemmerer-Berry<sup>85</sup> biochemical model of photosynthesis into dynamic land-surface models (e.g., 275 ORCHIDEE<sup>86</sup> or BETHY<sup>87</sup>) provides a gateway for assimilating satellite SIF data and improving 276 the accuracy of GPP estimations<sup>88,89</sup>. In addition, SIF resides at the core of a new generation of 277 photosynthesis models that emphasize the light reactions<sup>45,90</sup>. In the case of RTMs with established 278 SIF capabilities (Box 2), further improvements can be achieved by coupling with new techniques 279 measuring detailed 3D structures. Leaf RTMs would benefit from including variations in leaf 280 281 morphology, thylakoid structure, or the spectral signatures of PSI and PSII. The 3D parameterization of canopy RTMs via lidar-based reconstruction methods<sup>91,92</sup>, coupled to non-282 imaging<sup>17,19</sup> and imaging proximal/airborne SIF measurements (Fig. 3)<sup>93,94</sup>, offers excellent 283 284 opportunities to integrate and resolve the diversity of factors that control SIF across space and time (Challenge 7). Drone-based measurements could serve to investigate and model the impact of 285 atmospheric properties on SIF retrieval approaches, by hovering at different distances above the 286 target<sup>54</sup> (**Challenge 6**). Finally, less accurate but simpler alternative methods for separating the 287 physiological and structural influences on the SIF signal have been recently proposed based on the 288 theory of vegetation canopy near-infrared spectral invariants<sup>95,96</sup>. Whether this or other correction 289 methods are applicable to canopy SIF acquisitions across scales, especially observations at very 290 high spatial resolutions (Fig. 3) should be further investigated. 291

Equally critical for the consistent interpretation of SIF is the establishment of a global network and database of leaf and ecosystem-level SIF measurements covering different biomes, and supporting model development as well as airborne and satellite calibration/validation activities. While regional SIF networks are starting to emerge in North America, Europe, and Asia, their global connectivity should be a priority to promote the adoption of standards for instrument calibrations and long-term monitoring operations (Fig. 4).

Our roadmap for resolving the seven SIF challenges will only succeed through multidisciplinary
 collaboration involving specialists from across molecular biology, plant physiology, optical physics

and remote sensing. Together, the characterization and modeling of the interplay between structural,
 optical and functional dynamics of leaves and plant canopies, can turn our crops and forests into
 observable field laboratories.

## 303 Emerging and potential SIF applications

304 Satellite SIF data are already providing new insight into photosynthetic dynamics at the global scale<sup>97,98</sup>. Likewise, with the advent of multiscale SIF measurements (Fig. 3), and as the remaining 305 306 challenges are overcome (Fig. 4), a new range of SIF applications unfolds across fields of biochemistry, biophysics, ecology, ecophysiology, biogeochemistry, agriculture and forestry (Fig. 307 308 5). Equally important, the continuum of scales at which SIF can be measured provides a focal point to promote and strengthen the interaction between research communities, from plant molecular 309 biology to Earth-system science. Here, we outline four examples of potential and emerging SIF 310 applications. 311

Spatial and 3D photosynthesis. Photosynthetic CO<sub>2</sub> assimilation can be measured using infrared 312 gas analyzers, either coupled to chambers or enclosures at the leaf, shoot, and whole-plant level<sup>99</sup>, 313 or with a sonic anemometer at the ecosystem level using the eddy covariance approach<sup>100</sup>. These 314 methods, however, lack detailed spatial information. Spatial measurements of photosynthesis, in 315 terms of photochemical rates of the light reactions, require the use of imaging systems that, to date, 316 have remained restricted to the scale of leaves or small-sized plants, e.g. PAM imaging methods<sup>101</sup>. 317 SIF measurements have potential to fill this scale gap. For example, SIF imaging (Fig. 3) could be 318 319 benchmarked with eddy-covariance methods to reveal the spatial variability of photosynthesis within the footprint of ecosystem eddy covariance measurements, allowing us to investigate the 320 321 influence of microenvironment, understory and vertical canopy structure, or the interplay between biological and functional diversity within the ecosystem. Likewise, SIF imaging could be applied to 322

resolve photosynthesis dynamics in 3D, helping to advance our understanding of the interaction
 between plant structure and function<sup>102,103</sup>.

Physiological phenotyping and pre-visual stress detection. Spatial and temporal variations in plant 325 326 morphological traits (e.g., canopy height, leaf area, and plant growth) have been widely used as markers for field phenotypic variability and to investigate long-term plant stress responses. 327 However, these traits are insufficiently responsive to rapid plant physiological changes. This makes 328 them ill-suited for physiological phenotyping (i.e. breeding plant phenotypes displaying specific 329 physiological responses to the environment), or pre-visual stress detection and subsequent 330 331 optimization of water, pesticide and fertilizer use. The current phenotyping focus has, therefore, shifted towards measurements in the visible and infrared spectral ranges, where reflectance changes 332 can be associated with specific physiological and biochemical traits<sup>104</sup> or used for early-stress 333 detection<sup>105</sup>. In this context, emerging SIF imaging systems have already provided promising results 334 for applications in precision agriculture and detection of pest infestations<sup>93,106</sup>. In the near future, 335 these methods could also support precision forestry applications related to seedling production or 336 337 tree-scale forest management.

Functional plant diversity and spatial ecology. Functional diversity is a fundamental component of 338 the biodiversity concept<sup>107</sup>. As a global network for monitoring biodiversity through remotely 339 sensed plant functional traits is being developed<sup>108</sup>, SIF could become one of the new essential 340 variables for mapping functional diversity across ecosystem and landscape scales, given the wide 341 range of biochemical and physiological factors that SIF is sensitive to (Fig. 2) in relation to plant 342 productivity. For example, SIF has been shown to convey spatial information on leaf mass and 343 chlorophyll content<sup>109</sup>, and other functional plant traits<sup>110</sup> in various forest ecosystems. Additionally 344 and importantly, the combination of high-resolution structural, spectral and SIF data is potentially 345 the only viable option to investigate ecosystem functions that have remained hidden from our 346 observational abilities, such as photosynthetic phenology in evergreen forests<sup>18</sup>, cryptogamic 347

biocrusts<sup>111</sup> and spatially fragmented Antarctic mosses<sup>74</sup>. Together with spatial photosynthesis, SIF
could also offer unique opportunities for studies in spatial ecology<sup>111,112</sup>, where plant environmental
responses and biotic interactions could leave their imprint on SIF.

Carbon and water cycle studies. The carbon and water cycles of terrestrial ecosystems are 351 intricately connected via stomatal regulation and total leaf area. Because both canopy 352 evapotranspiration and canopy SIF dynamics are strongly controlled by leaf area, and since ChlaF 353 can also decrease with stomatal closure - via increased NPQ in response to water stress<sup>113, 114</sup>; tower 354 and satellite SIF have been preliminarily used to investigate canopy conductance and plant 355 transpiration<sup>115,116</sup>. No doubt, better constraints on transpiration and photosynthetic dynamics in 356 land-surface models will be achieved as the mechanistic basis of SIF is elucidated across scales 357 (Challenges 1-7), and the integration of SIF with other remote sensing datasets increases, such as 358 land-surface temperature<sup>115</sup>, surface soil moisture<sup>89</sup>, radar-measured vegetation optical depth 359 characterizing canopy structure and water content<sup>117</sup>, or column-averaged atmospheric CO<sub>2</sub> <sup>92</sup>. New 360 knowledge of photosynthesis at the ecosystem and regional scales will bring further insight into the 361 large-scale interactions between environmental drivers and plant productivity, and feedbacks 362 between the biosphere and atmosphere. 363

# 364 Concluding remarks

The SIF signal gathers a wealth of physiological, biochemical, and structural information as it travels from the photosystems to the top of canopy and beyond (Fig. 2). This can leave the impression that SIF is, to use the classic analogy, the 'Swiss Army Knife' of photosynthesis measurements. Critically, the variation in SIF caused by physical and biotic factors is entangled in the spatiotemporal domain, and our capacity to disentangle it into useful informative components requires further attention. Historically, photosynthesis research has been a multidisciplinary endeavor, with breakthroughs in the 20<sup>th</sup> century emerging from collaboration between chemists,

biologists and physicists. We are now entering a new era of multiscale observations of
photosynthesis which requires the interdisciplinary research environment to flourish further, this
time to resolve the mechanistic connection between SIF and GPP and to scale it across space and
time. The technology to measure SIF is developing at a faster pace than our capacity to interpret the
acquired data. With the challenges, roadmap and unfolding opportunities introduced here we hope
to encourage more scientists to join the multidisciplinary quest to reveal the true potential of SIF
observation.

379

#### 380 **References**

381	1.	Genty, B., Wonders, J. & Baker, N. R. Non-photochemical quenching of Fo in leaves is
382		emission wavelength dependent: consequences for quenching analysis and its interpretation.
383		Photosynth. Res. 26, 133-139 (1990).

- Franck, F., Juneau, P. & Popovic, R. Resolution of the photosystem I and photosystem II
   contributions to chlorophyll fluorescence of intact leaves at room temperature. *Biochim. Biophys. Acta-Bioenergetics* 1556, 239-246 (2002).
- Neubauer, C. & Schreiber, U. The polyphasic rise of chlorophyll fluorescence upon onset of
   strong continuous illumination: I. Saturation characteristics and partial control by the
   photosystem II acceptor side. *Zeitschrift für Naturforschung C* 42, 1246-1254 (1987).
- Strasser, R. J., Tsimilli-Michael, M. & Srivastava, A. Analysis of the chlorophyll a
   fluorescence transient. In: Papageorgiou G.C., Govindjee (eds) Chlorophyll a Fluorescence.
   Advances in Photosynthesis and Respiration, vol 19. Springer, Dordrecht (2004).
- 5. Schreiber, U., Schliwa, U. & Bilger, W. Continuous recording of photochemical and non-
- 394 photochemical chlorophyll fluorescence quenching with a new type of modulation
- 395 fluorometer. *Photosynth. Res.* **10**, 51-62 (1986).

- Maxwell, K. & Johnson, G. N. Chlorophyll fluorescence—a practical guide. *J. Exp. Bot.* 51,
  659-668 (2000).
- 398 7. Govindjee, E. 63 Years since Kautsky-chlorophyll-a fluorescence. *Aust. J. Plant Physiol.* 22, 131-160 (1995).
- 8. Porcar-Castell, A. *et al.* Linking chlorophyll a fluorescence to photosynthesis for remote
  sensing applications: mechanisms and challenges. *J. Exp. Bot.* 65, 4065-4095 (2014).
- 402 9. Tikkanen, M., Rantala, S., Grieco, M. & Aro, E. Comparative analysis of mutant plants
  403 impaired in the main regulatory mechanisms of photosynthetic light reactions-From
  404 biophysical measurements to molecular mechanisms. *Plant Physiol. Biochem.* 112, 290-301
  405 (2017).
- Kolber, Z. *et al.* Measuring photosynthetic parameters at a distance: laser induced
  fluorescence transient (LIFT) method for remote measurements of photosynthesis in
  terrestrial vegetation. *Photosynth. Res.* 84, 121-129 (2005).
- Keller, B. *et al.* Genotype specific photosynthesis x environment interactions captured by
  automated fluorescence canopy scans over two fluctuating growing seasons. *Front. Plant Sci.* 10, 1482 (2019).
- 412 12. Mohammed, G. H. *et al.* Remote sensing of solar-induced chlorophyll fluorescence (SIF) in
  413 vegetation: 50 years of progress. *Remote Sens. Environ.* 231, 111177 (2019).
- 414 13. Evain, S., Camenen, L. & Moya, I. Three-channel detector for remote sensing of chlorophyll
  415 fluorescence and reflectance from vegetation. In: M. Leroy (ed.), 8<sup>a</sup> International
- 416 symposium: physical measurements and signatures in remote sensing, pp. 395-400. Aussois,
  417 CNES, France (2001).
- 418 14. Louis, J. *et al.* Remote sensing of sunlight-induced chlorophyll fluorescence and reflectance
  419 of Scots pine in the boreal forest during spring recovery. *Remote Sens. Environ.* 96, 37-48
  420 (2005).

- 421 15. Guanter, L. *et al.* Estimation of solar-induced vegetation fluorescence from space
  422 measurements. *Geophys. Res. Lett.* 34 (2007).
- 423 16. Aasen, H. *et al.* Sun-induced chlorophyll fluorescence II: review of passive measurement
  424 setups, protocols, and their application at the leaf to canopy level. *Remote Sensing* 11, 927
  425 (2019).
- Yang, X. *et al.* Solar-induced chlorophyll fluorescence that correlates with canopy
  photosynthesis on diurnal and seasonal scales in a temperate deciduous forest. *Geophys. Res. Lett.* 42, 2977-2987 (2015).
- Magney, T. S. *et al.* Mechanistic evidence for tracking the seasonality of photosynthesis
  with solar-induced fluorescence. *PNAS* 116, 11640-11645 (2019).
- 431 19. Bendig, J., Malenovský, Z., Gautam, D. & Lucieer, A. Solar-Induced Chlorophyll
  432 Fluorescence Measured From an Unmanned Aircraft System: Sensor Etaloning and Platform
  433 Motion Correction. *IEEE Trans. Geosci. Remote Sens.* (2019).
- Vargas, J. Q. *et al.* Unmanned aerial systems (UAS)-based methods for solar induced
  chlorophyll fluorescence (SIF) retrieval with non-imaging spectrometers: state of the art. *Remote Sens.* 12, 1624 (2020).
- 437 21. Rascher, U. *et al.* Sun-induced fluorescence–a new probe of photosynthesis: First maps from
  438 the imaging spectrometer HyPlant. *Global Change Biol.* 21, 4673-4684 (2015).
- 439 22. Frankenberg, C. *et al.* The Chlorophyll Fluorescence Imaging Spectrometer (CFIS),
  440 mapping far red fluorescence from aircraft. *Remote Sens. Environ.* 217, 523-536 (2018).
- 441 23. Frankenberg, C. *et al.* New global observations of the terrestrial carbon cycle from GOSAT:
  442 Patterns of plant fluorescence with gross primary productivity. *Geophys. Res. Lett.* 38
- 443 (2011).
- 444 24. Köhler, P. *et al.* Global Retrievals of Solar-Induced Chlorophyll Fluorescence at Red
  445 Wavelengths With TROPOMI. *Geophys. Res. Lett.* 47, e2020GL087541 (2020).

- 25. Drusch, M. *et al.* The fluorescence explorer mission concept ESA's earth explorer 8. *IEEE Trans. Geosci. Remote Sens.* 55, 1273-1284 (2016).
- 448 26. Olascoaga, B., Mac Arthur, A., Atherton, J. & Porcar-Castell, A. A comparison of methods
  449 to estimate photosynthetic light absorption in leaves with contrasting morphology. *Tree*450 *Physiol.* 36, 368-379 (2016).
- 27. Zhang, Z. *et al.* Assessing bi-directional effects on the diurnal cycle of measured solarinduced chlorophyll fluorescence in crop canopies. *Agric. For. Meteorol.* 295, 108147
  (2020).
- 454 28. Bittner, T., Irrgang, K., Renger, G. & Wasielewski, M. R. Ultrafast excitation energy
  455 transfer and exciton-exciton annihilation processes in isolated light harvesting complexes of
- 456 photosystem II (LHC II) from spinach. J. Phys. Chem. 98, 11821-11826 (1994).
- 457 29. Kalaji, H. M. *et al.* Frequently asked questions about chlorophyll fluorescence, the sequel.
  458 *Photosynth. Res.* 132, 13-66 (2017).
- Genty, B., Briantais, J. & Baker, N. R. The relationship between the quantum yield of
  photosynthetic electron transport and quenching of chlorophyll fluorescence. *Biochim. Biophys. Acta-General Subjects* 990, 87-92 (1989).
- 462 31. Anderson, J. M., Chow, W. S. & Goodchild, D. J. Thylakoid membrane organisation in
  463 sun/shade acclimation. *Funct. Plant Biol.* 15, 11-26 (1988).
- Ballottari, M., Dall'Osto, L., Morosinotto, T. & Bassi, R. Contrasting behavior of higher
  plant photosystem I and II antenna systems during acclimation. *J. Biol. Chem.* 282, 89478958 (2007).
- 33. Schreiber, U., Klughammer, C. & Kolbowski, J. Assessment of wavelength-dependent
  parameters of photosynthetic electron transport with a new type of multi-color PAM
  chlorophyll fluorometer. *Photosynth. Res.* 113, 127-144 (2012).

470	34.	Laisk, A. et al. A computer-operated routine of gas exchange and optical measurements to
471		diagnose photosynthetic apparatus in leaves. Plant, Cell Environ. 25, 923-943 (2002).
472	35.	Pfündel, E. Estimating the contribution of photosystem I to total leaf chlorophyll
473		fluorescence. Photosynthesis Res. 56, 185-195 (1998).
474	36.	Peterson, R. B. et al. Fluorescence Fo of photosystems II and I in developing C3 and C4
475		leaves, and implications on regulation of excitation balance. Photosynth. Res. 122, 41-56
476	37.	Pfündel, E. E. Simultaneously measuring pulse-amplitude-modulated (PAM) chlorophyll
477		fluorescence of leaves at wavelengths shorter and longer than 700 nm. Photosynth. Res., 1-
478		14 (2021).
479	38.	Demmig-Adams, B. & Adams III, W. W. Photoprotection in an ecological context: the
480		remarkable complexity of thermal energy dissipation. New Phytol. 172, 11-21 (2006).
481	39.	Porcar-Castell, A. A high-resolution portrait of the annual dynamics of photochemical and
482		non-photochemical quenching in needles of Pinus sylvestris. Physiol. Plant. 143, 139-153
483		(2011).
484	40.	Van der Tol, C., Berry, J. A., Campbell, P. & Rascher, U. Models of fluorescence and
485		photosynthesis for interpreting measurements of solar-induced chlorophyll fluorescence. J.
486		Geophys. Res.: Biogeosciences 119, 2312-2327 (2014).
487	41.	Springer, K. R., Wang, R. & Gamon, J. A. Parallel seasonal patterns of photosynthesis,
488		fluorescence, and reflectance indices in boreal trees. Remote Sens. 9, 691 (2017).
489	42.	Zhang, C. et al. Do all chlorophyll fluorescence emission wavelengths capture the spring
490		recovery of photosynthesis in boreal evergreen foliage? <i>Plant, Cell Environ.</i> 42, 3264-3279
491		(2019).
492	43.	Ensminger, I. et al. Intermittent low temperatures constrain spring recovery of
493		photosynthesis in boreal Scots pine forests. Glob. Change Biol. 10, 995-1008 (2004).

- 494 44. Verhoeven, A. Sustained energy dissipation in winter evergreens. *New Phytol.* 201, 57-65
  495 (2014).
- 496 45. Gu, L., Han, J., Wood, J. D., Chang, C. Y. & Sun, Y. Sun-induced Chl fluorescence and its
  497 importance for biophysical modeling of photosynthesis based on light reactions. *New*498 *Phytol.* 223, 1179-1191 (2019).
- 499 46. Raczka, B. *et al.* Sustained nonphotochemical quenching shapes the seasonal pattern of
  500 solar-induced fluorescence at a high-elevation evergreen forest. *J. Geophys. Res.:*501 *Biogeosciences* 124, 2005-2020 (2019).
- 502 47. Nixon, P. J. Chlororespiration. *Philos. Trans. R. Soc. Lond., B, Biol. Sci.* 355, 1541-1547
  503 (2000).
- 48. Ogren, W. L. Photorespiration: pathways, regulation, and modification. *Annu. Rev. Plant Physiol.* 35, 415-442 (1984).
- 49. Asada, K. The water-water cycle in chloroplasts: scavenging of active oxygens and
  dissipation of excess photons. *Annu. Rev. Plant Biol.* 50, 601-639 (1999).
- 508 50. Morfopoulos, C. *et al.* A model of plant isoprene emission based on available reducing
  509 power captures responses to atmospheric CO2. *New Phytol.* 203, 125-139 (2014).
- 510 51. Maseyk, K., Lin, T., Cochavi, A., Schwartz, A. & Yakir, D. Quantification of leaf-scale
  511 light energy allocation and photoprotection processes in a Mediterranean pine forest under
  512 extensive seasonal drought. *Tree Physiol.* **39**, 1767-1782 (2019).
- 513 52. Migliavacca, M. *et al.* Plant functional traits and canopy structure control the relationship
  514 between photosynthetic CO<sub>2</sub> uptake and far-red sun-induced fluorescence in a Mediterranean
  515 grassland under different nutrient availability. *New Phytol.* 214, 1078-1091 (2017).
- 516 53. Kallel, A. FluLCVRT: Reflectance and fluorescence of leaf and canopy modeling based on
- 517 Monte Carlo vector radiative transfer simulation. J. Quant. Spectrosc. Radiat. Transf. 253,
- 518 107183 (2020).

519	54.	Sabater, N. <i>et al.</i> Compensation of oxygen transmittance effects for proximal sensing
520		retrieval of canopy-leaving sun-induced chlorophyll fluorescence. Remote Sens. 10, 1551
521		(2018).

- 52. Sabater, N., Kolmonen, P., Van Wittenberghe, S., Arola, A. & Moreno, J. Challenges in the
  atmospheric characterization for the retrieval of spectrally resolved fluorescence and PRI
  region dynamics from space. *Remote Sens. Environ.* 254, 112226 (2021).
- 525 56. Iermak, I., Vink, J., Bader, A. N., Wientjes, E. & van Amerongen, H. Visualizing
  526 heterogeneity of photosynthetic properties of plant leaves with two-photon fluorescence
  527 lifetime imaging microscopy. *Biochim. Biophys. Acta-Bioenergetics* 1857, 1473-1478
  528 (2016).
- 529 57. Romero, J. M., Cordon, G. B. & Lagorio, M. G. Modeling re-absorption of fluorescence
  530 from the leaf to the canopy level. *Remote Sens. Environ.* 204, 138-146 (2018).
- 531 58. Magney, T. S. *et al.* Disentangling changes in the spectral shape of chlorophyll
  532 fluorescence: Implications for remote sensing of photosynthesis. *J. Geophys. Res.:*533 *Biogeosciences* 124, 1491-1507 (2019).
- 534 59. Murchie, E. H. *et al.* Measuring the dynamic photosynthome. *Ann. Bot.* 122, 207-220
  535 (2018).
- Magney, T. S., Barnes, M. L. & Yang, X. On the covariation of chlorophyll fluorescence
  and photosynthesis across scales. Geophys. Res. Lett. 47, e2020GL091098 (2020).
- 538 61. Yang, P., van der Tol, C., Campbell, P. K. & Middleton, E. M. Unraveling the physical and
  539 physiological basis for the solar-induced chlorophyll fluorescence and photosynthesis
  540 relationship using continuous leaf and canopy measurements of a corn crop. *Biogeosciences*
- **18**, 441-465 (2021).
- 542 62. Liu, X. *et al.* Downscaling of solar-induced chlorophyll fluorescence from canopy level to
  543 photosystem level using a random forest model. *Remote Sens. Environ.* 231, 110772 (2019).

544	63.	Joiner, J. et al. Systematic Orbital Geometry-Dependent Variations in Satellite Solar-
545		Induced Fluorescence (SIF) Retrievals. Remote Sens. 12, 2346 (2020).
546	64.	Dechant, B. et al. Canopy structure explains the relationship between photosynthesis and
547		sun-induced chlorophyll fluorescence in crops. Remote Sens. Environ. 241, 111733 (2020).
548	65.	He, L. et al. From the Ground to Space: Using Solar-Induced Chlorophyll Fluorescence to
549		Estimate Crop Productivity. Geophys. Res. Lett. 47, e2020GL087474 (2020).
550	66.	Ač, A. et al. Meta-analysis assessing potential of steady-state chlorophyll fluorescence for
551		remote sensing detection of plant water, temperature and nitrogen stress. Remote Sens.
552		Environ. 168, 420-436 (2015).
553	67.	Wohlfahrt, G. et al. Sun-induced fluorescence and gross primary productivity during a heat
554		wave. Sci. Rep. 8, 1-9 (2018).
555	68.	Van Wittenberghe, S., Alonso, L., Verrelst, J., Moreno, J. & Samson, R. Bidirectional sun-
556		induced chlorophyll fluorescence emission is influenced by leaf structure and light
557		scattering properties: A bottom-up approach. Remote Sens. Environ. 158, 169-179 (2015).
558	69.	Magney, T. S. et al. Connecting active to passive fluorescence with photosynthesis: A
559		method for evaluating remote sensing measurements of Chl fluorescence. New Phytol. 215,
560		1594-1608 (2017).
561	70.	Rajewicz, P. A., Atherton, J., Alonso, L. & Porcar-Castell, A. Leaf-level spectral
562		fluorescence measurements: comparing methodologies for broadleaves and needles. Remote
563		Sens. 11, 532 (2019).
564	71.	Van Wittenberghe, S., Alonso, L., Malenovský, Z. & Moreno, J. In vivo photoprotection
565		mechanisms observed from leaf spectral absorbance changes showing VIS-NIR slow-
566		induced conformational pigment bed changes. Photosynth. Res. 142, 283-305 (2019).

567	72.	Meeker, E. W., Magney, T. S., Bambach, N., Momayyezi, M. & McElrone, A. J.
568		Modification of a gas exchange system to measure active and passive chlorophyll
569		fluorescence simultaneously under field conditions. AoB Plants 13, plaa066 (2021).
570	73.	Acebron, K. et al. Diurnal dynamics of nonphotochemical quenching in Arabidopsis npq
571		mutants assessed by solar-induced fluorescence and reflectance measurements in the field.
572		New Phytol. (2020).
573	74.	Malenovský, Z., Lucieer, A., King, D. H., Turnbull, J. D. & Robinson, S. A. Unmanned
574		aircraft system advances health mapping of fragile polar vegetation. Methods Ecol. Evol. 8,
575		1842-1857 (2017).
576	75.	Atherton, J., Nichol, C. J. & Porcar-Castell, A. Using spectral chlorophyll fluorescence and
577		the photochemical reflectance index to predict physiological dynamics. Remote Sens.
578		Environ. 176, 17-30 (2016).
579	76.	Van Wittenberghe, S. et al. Combined dynamics of the 500-600 nm leaf absorption and
580		chlorophyll fluorescence changes in vivo: Evidence for the multifunctional energy
581		quenching role of xanthophylls. Biochim. Biophys. Acta-Bioenergetics 1862, 148351 (2021).
582	77.	Gamon, J. A. et al. Remote sensing of the xanthophyll cycle and chlorophyll fluorescence in
583		sunflower leaves and canopies. Oecologia 85, 1-7 (1990).
584	78.	Filella, I. et al. PRI assessment of long-term changes in carotenoids/chlorophyll ratio and
585		short-term changes in de-epoxidation state of the xanthophyll cycle. Int. J. Remote Sens. 30,
586		4443-4455 (2009).
587	79.	Peñuelas, J., Filella, I. & Gamon, J. A. Assessment of photosynthetic radiation-use
588		efficiency with spectral reflectance. New Phytol. 131, 291-296 (1995).
589	80.	Gamon, J. A. et al. A remotely sensed pigment index reveals photosynthetic phenology in
590		evergreen conifers. PNAS 113, 13087-13092 (2016).

- 591 81. Costa, J. M., Grant, O. M. & Chaves, M. M. Thermography to explore plant-environment
  592 interactions. *J. Exp. Bot.* 64, 3937-3949 (2013).
- 593 82. Konings, A. G., Rao, K. & Steele-Dunne, S. C. Macro to micro: microwave remote sensing
  594 of plant water content for physiology and ecology. *New Phytol.* 223, 1166-1172 (2019).
- 595 83. Junttila, S. *et al.* Terrestrial laser scanning intensity captures diurnal variation in leaf water
  596 potential. *Remote Sens. Environ.* 255, 112274 (2021).
- 597 84. Whelan, M. E. *et al.* Two Scientific Communities Striving for a Common Cause:
  598 innovations in carbon cycle science. *Bull. Am. Meteorol. Soc.* (2020).
- 59985.Farquhar, G. D., von Caemmerer, S. v. & Berry, J. A. A biochemical model of
- 600 photosynthetic CO 2 assimilation in leaves of C 3 species. *Planta* **149**, 78-90 (1980).
- 86. Bacour, C. *et al.* Improving estimates of gross primary productivity by assimilating solarinduced fluorescence satellite retrievals in a terrestrial biosphere model using a processbased SIF model. *J. Geophys. Res.: Biogeosciences* 124, 3281-3306 (2019).
- 87. Norton, A. J. *et al.* Estimating global gross primary productivity using chlorophyll
- fluorescence and a data assimilation system with the BETHY-SCOPE model.
- 606 *Biogeosciences* **16**, 3069-3093 (2019).
- 88. Thum, T. *et al.* Modelling sun-induced fluorescence and photosynthesis with a land surface
  model at local and regional scales in northern Europe. *Biogeosciences* 14, 1969-1987
  (2017).
- 610 89. Qiu, B., Chen, J. M., Ju, W., Zhang, Q. & Zhang, Y. Simulating emission and scattering of
  611 solar-induced chlorophyll fluorescence at far-red band in global vegetation with different
  612 canopy structures. *Remote Sens. Environ.* 233, 111373 (2019).
- 90. Johnson, J. E. & Berry, J. A. The role of Cytochrome b 6 f in the control of steady-state
  photosynthesis: a conceptual and quantitative model. *Photosynth. Res.*, 1-36 (2021).

- 91. Janoutová, R. *et al.* Influence of 3D spruce tree representation on accuracy of airborne and
  satellite forest reflectance simulated in DART. *Forests* 10, 292 (2019).
- 617 92. Liu, W. *et al.* Simulating solar-induced chlorophyll fluorescence in a boreal forest stand
  618 reconstructed from terrestrial laser scanning measurements. *Remote Sens. Environ.* 232,
  619 111274 (2019).
- 93. Pinto, F. *et al.* Sun-induced chlorophyll fluorescence from high-resolution imaging
  spectroscopy data to quantify spatio-temporal patterns of photosynthetic function in crop
  canopies. *Plant, Cell Environ.* **39**, 1500-1512 (2016).
- Siegmann, B. *et al.* The high-performance airborne imaging spectrometer HyPlant—From
  raw images to top-of-canopy reflectance and fluorescence products: Introduction of an
  automatized processing chain. *Remote Sens.* 11, 2760 (2019).
- 95. Yang, P., van der Tol, C., Campbell, P. K. & Middleton, E. M. Fluorescence Correction
  Vegetation Index (FCVI): A physically based reflectance index to separate physiological
  and non-physiological information in far-red sun-induced chlorophyll fluorescence. *Remote*
- 629 Sens. Environ. **240**, 111676 (2020).
- 630 96. Zeng, Y. *et al.* A radiative transfer model for solar induced fluorescence using spectral
  631 invariants theory. *Remote Sens. Environ.* 240, 111678 (2020).
- 632 97. Green, J. K. *et al.* Large influence of soil moisture on long-term terrestrial carbon uptake.
  633 *Nature* 565, 476 (2019).
- 634 98. Wang, S. *et al.* Urban– rural gradients reveal joint control of elevated CO 2 and temperature
  635 on extended photosynthetic seasons. *Nat. Ecol. Evo.* 3, 1076-1085 (2019).
- 636 99. Long, S. P., Farage, P. K. & Garcia, R. L. Measurement of leaf and canopy photosynthetic
  637 CO2 exchange in the field. *J. Exp. Bot.* 47, 1629-1642 (1996).

- Baldocchi, D. D. Assessing the eddy covariance technique for evaluating carbon dioxide
  exchange rates of ecosystems: past, present and future. *Glob. Change Biol.* 9, 479-492
  (2003).
- 641 101. Kaiser, Y. I., Menegat, A. & Gerhards, R. Chlorophyll fluorescence imaging: a new method
  642 for rapid detection of herbicide resistance in A lopecurus myosuroides. *Weed Res.* 53, 399643 406 (2013).
- 644 102. Sievänen, R., Godin, C., DeJong, T. M. & Nikinmaa, E. Functional–structural plant models:
  645 a growing paradigm for plant studies. *Ann. Bot.* 114, 599-603 (2014).
- 103. Damm, A., Paul-Limoges, E., Kükenbrink, D., Bachofen, C. & Morsdorf, F. Remote sensing
  of forest gas exchange: Considerations derived from a tomographic perspective. *Glob*.
- 648 *Change Biol.* **26**, 2717-2727 (2020).
- 649 104. Ensminger, I. Fast track diagnostics: Hyperspectral reflectance differentiates disease from
  650 drought stress in trees. *Tree Physiol.* 40, 1143-1146 (2020).
- 651 105. Mutka, A. M. & Bart, R. S. Image-based phenotyping of plant disease symptoms. *Frontiers*652 *in plant science* 5, 734 (2015).
- 106. Zarco-Tejada, P. J. *et al.* Previsual symptoms of Xylella fastidiosa infection revealed in
  spectral plant-trait alterations. *Nat. Plants* 4, 432-439 (2018).
- Díaz, S. & Cabido, M. Vive la différence: plant functional diversity matters to ecosystem
  processes. *Trends Ecol. Evol.* 16, 646-655 (2001).
- 657 108. Skidmore, A. K. *et al.* Environmental science: Agree on biodiversity metrics to track from
  658 space. *Nat. News* 523, 403 (2015).
- 109. Tagliabue, G. *et al.* Sun–induced fluorescence heterogeneity as a measure of functional
  diversity. *Remote Sens. Environ.* 247, 111934 (2020).

- 110. Pacheco-Labrador, J. *et al.* Multiple-constraint inversion of SCOPE. Evaluating the potential
  of GPP and SIF for the retrieval of plant functional traits. *Remote Sens. Environ.* 234,
  111362 (2019).
- 664 111. Smith, W. K. *et al.* Remote sensing of dryland ecosystem structure and function: Progress,
  665 challenges, and opportunities. *Remote Sens. Environ.* 233, 111401 (2019).
- Kellner, J. R., Albert, L. P., Burley, J. T. & Cushman, K. C. The case for remote sensing of
  individual plants. *Am. J. Bot.* 106, 1139-1142 (2019).
- Flexas, J. *et al.* Steady-state chlorophyll fluorescence (Fs) measurements as a tool to follow
  variations of net CO2 assimilation and stomatal conductance during water-stress in C3
  plants. *Physiol. Plant.* 114, 231-240 (2002).
- 114. Marrs, J. K. *et al.* Solar-induced fluorescence does not track photosynthetic carbon
  assimilation following induced stomatal closure. *Geophys. Res. Lett.* 47, e2020GL087956
  (2020).
- Maes, W. H. *et al.* Sun-induced fluorescence closely linked to ecosystem transpiration as
  evidenced by satellite data and radiative transfer models. *Remote Sens. Environ.* 249,
  112030 (2020).
- 677 116. Shan, N. *et al.* A model for estimating transpiration from remotely sensed solar-induced
  678 chlorophyll fluorescence. *Remote Sens. Environ.* 252, 112134 (2021).
- Liu, W. *et al.* Simulating solar-induced chlorophyll fluorescence in a boreal forest stand
  reconstructed from terrestrial laser scanning measurements. *Remote Sens. Environ.* 232,
  111274 (2019).
- Albert, L. P. *et al.* Stray light characterization in a high-resolution imaging spectrometer
  designed for solar-induced fluorescence. Proc. SPIE 10986, Algorithms, Technologies, and
  Applications for Multispectral and Hyperspectral Imagery XXV, 109860G (2019).

685	119.	Meroni, M. et al. Remote sensing of solar-induced chlorophyll fluorescence: Review of
686		methods and applications. Remote Sens. Environ. 113, 2037-2051 (2009).
687	120.	Cendrero-Mateo, M. P. et al. Sun-induced chlorophyll fluorescence III: Benchmarking
688		retrieval methods and sensor characteristics for proximal sensing. Remote Sens. 11, 962
689		(2019).
690	121.	Vilfan, N. et al. Extending Fluspect to simulate xanthophyll driven leaf reflectance
691		dynamics. Remote Sens. Environ. 211, 345-356 (2018).

- Yang, P., Prikaziuk, E., Verhoef, W. & van der Tol, C. SCOPE 2.0: A model to simulate
  vegetated land surface fluxes and satellite signals. *Geosci. Model Dev. Discuss.*, 1-26
  (2020).
- Gastellu-Etchegorry, J. *et al.* DART: recent advances in remote sensing data modeling with
  atmosphere, polarization, and chlorophyll fluorescence. IEEE J. Sel. Top. Appl. Earth Obs.
  Remote Sens. 10, 2640-2649 (2017).

# 699 **Corresponding author:** Albert Porcar-Castell (joan.porcar@helsinki.fi)

700

701 Acknowledgements: This perspective idea originated during the Fluorescence Across Space and 702 Time (FAST) Workshop, which took place in Hyytiälä Forestry Research Station (SMEARII, 703 Finland) during February 2019. We thank the following participants for active discussions during the workshop: Juliane Bendig, Kukka-Maaria Erkkilä, Noda Hibiki, Laura V. Junker-Frohn, 704 705 Valentyna Kuznetsova, Hannakaisa Lindqvist, Paul Näthe, Jaakko Oivukkamaki, Neus Sabater, Twinkle Solanki, Tea Thum, Shan Xu and Chao Zhang. We also thank Barry Osmond and Josep 706 707 Peñuelas for valuable comments to the manuscript, to Nuria Altimir for improving graphic design 708 of Figs 1 and 5, and to Bastian Siegmann for the preparation of the HyPlant image in Fig. 3. The

709	Academy of Finland (Project # 288039 and 319211) is acknowledged for the financial support. ZM
710	was supported by the Australian Research Council (FT160100477), TM was supported by the
711	National Aeronautics and Space Administration (80NSSC19M0129), and SVW was supported by
712	the Generalitat Valenciana and the European Social Fund (APOSTD/2018/162). Headwall SIF
713	images from LPA and JRK were supported by grants from the Institute at Brown for Environment
714	and Society at Brown University.
715	Author contributions: APC conceived the original idea and wrote the manuscript with ZM, TM,
716	BL, SVW, BFM, FM, YZ, KM with comments and contributions from all co-authors. In addition,
717	these authors had special contribution to the following parts: Fig.1 (APC, ZM and SVW), Fig.2
718	(APC, BFM, TM and SVW), Fig. 3 (LPA, UR and JRK), Fig. 4. (APC, ZM, UR, BFM), Fig. 5
719	(JIGP, JA, ZM, IE), Box 1 (TM, APC), Box 2 (ZM, APC), Supplementary information (ZM, FZ).
720	
721	
722	
723	
724	
725	
726	
727	
728	
120	
729	

### 730 Figure Legends

731 Figure 1. From incoming radiation to observed SIF and photosynthesis: mechanistic

challenges. Solar radiation reaching the top of the atmosphere (TOA) is partly absorbed and 732 733 scattered by atmospheric gases and particles, decreasing its intensity as it reaches the bottom of the atmosphere (BOA), generating specific absorption features. Part of the radiation is absorbed by 734 photosynthetic pigments in vegetation and leaves (fAPAR<sub>g</sub>) (Challenge 1), associated with either 735 photosystem I (PSI) or photosystem II (PSII), which contribute with differential dynamics and 736 spectral properties to overall SIF emission (Challenge 2). Within each photosystem, energy is 737 738 further partitioned into three dynamic processes (Challenge 3): i) photochemistry (leading mainly to linear (LET) or cyclic (CET) electron transport, the latter involving PSI only), ii) thermal energy 739 740 dissipation, and iii) ChlaF. Photosynthetic energy (expressed for simplicity in terms of NADPH 741 equivalents) is further partitioned between alternative energy sinks and gross photosynthesis (A<sub>G</sub>), and again between gross primary productivity (GPP) and photorespiration (P<sub>R</sub>), with dynamics that 742 are not necessarily seen by SIF (Challenge 4). Notably, because it is only possible to measure the 743 744 net flux of CO<sub>2</sub> from a leaf or ecosystem, i.e. net photosynthesis or net primary productivity (NPP), the rate of daytime respiration (R<sub>D</sub>) must be known or estimated. In turn, because emitted ChlaF 745 overlaps with the absorption spectra of leaves and plant canopies, some SIF photons - especially 746 those in the red wavelengths - are re-absorbed within the canopy (Challenge 5). Emitted ChlaF is 747 further scattered and absorbed by aerosols and gases in the atmosphere (Challenge 6). 748

749

Figure 2. The connection between SIF and GPP across space and time. The relationship
between SIF and GPP is affected by multiple factors as we move across spatial and temporal scales.
Some factors exert a similar effect on SIF and GPP, keeping them positively correlated - we call
these couplers. Other factors differentially affect SIF and GPP - we call these decouplers. Factors

driving the dynamics of NPQ and APAR will tend to keep SIF and GPP coupled both across space 754 and time, whereas factors adding variation to the energy partitioning between ChIF and GPP, or 755 influencing the reabsorption of ChIF, will tend to decouple SIF from GPP (see examples in the 756 757 figure). Note how the shape of the ChIF spectrum ("Spectral dynamics") changes across scales in response to reabsorption within the chloroplast, leaf and canopy, measurable as SIF only within 758 759 discrete wavelengths at the canopy and ecosystem levels (Box 1). Equally important to our 760 understanding of the spatial context of the factors that couple/decouple SIF to GPP is understanding their temporal range of action (lower panels). For example, the rapid (second/minute) decrease in 761 ChlaF upon saturating illumination of dark acclimated leaves reflects the dynamics of NPO<sup>76</sup>. 762 763 Similar dynamics can be seen under natural conditions at the **diurnal/seasonal** scale in Scots pine needles, as the quantum yield of fluorescence ( $\Phi$ F) responds to PQ and NPQ (redrawn from Porcar-764 Castell<sup>39</sup>). Here, SIF was estimated for illustrative purposes as SIF (r.u.) = PAR x 0.8 x 0.5 x  $\Phi$ F, 765 766 where 0.8 and 0.5 are estimates for fAPAR<sub>g</sub> and the fraction of radiation absorbed by PSII. Likewise, **interannual** dynamics at the regional-to-local scales<sup>24</sup> can reflect changes in canopy 767 768 structure, physiological stress responses or other functional traits. Ultimately, the challenge of integrating and disentangling the impact of these couplers/decouplers across space, time, species 769 and plant stress responses remains (Challenge 7). 770

771

**Figure 3.** State-of-the-art SIF imaging methods allow for the observation of SIF across a continuum of scales: from the leaf-to-individual (top row) to the individual-to-landscape (bottom row). Panel A shows an RGB image of a senescing maple tree next to an oak tree with green leaves. Panel B shows the SIF image of the same trees retrieved in the O2A band at 760 nm (SIF760) using a commercial, off-the-shelf imaging spectrometer<sup>118</sup> mounted on a tripod some meters away and after applying a filter to exclude non-vegetation pixels (pixels with an normalized difference vegetation index (NDVI) < 0.65). As expected, the green and photosynthetically active oak emitted SIF at

higher magnitude (Panel C) than the senescing maple. Similarly, panels D-E present an airborne 779 780 RGB and SIF760 map obtained with data from the HyPlant sensor collected at an altitude of 680 m above ground<sup>94</sup>. The scene shows several plots within an experimental apple tree plantation at the 781 782 agricultural research site Campus Klein-Altendorf (University of Bonn, Germany), where apple tree varieties of different ages were growing in a typical row structure. Single tree crowns were 783 segmented by overlaying the SIF images with a 3D surface map and all pixels that were related to a 784 785 background signal (defined as ground level + 30 cm) were excluded. The image visualizes the signal of individual trees, where each pixel corresponds to an area of 1x1 meters and thus the small 786 clusters represent the signal of an individual tree. 787

788

Figure 4. A roadmap towards the standardized interpretation of SIF. The critical steps, data sources and methods that will be required to overcome the seven challenges are introduced to allow for a consistent interpretation of spectral observations in terms of leaf, canopy and ecosystem traits.

792

**Figure 5.** Potential and emerging SIF applications illustrated in the form of a "SIF-city" metro plan, where different colors denote five fields of plant science. Identified research applications (metro stops) are causally connected in individual communication lines, but the final trajectories and number of stops will depend on how the field of SIF research evolves over the next years. The redcolored stops denote the application topics elaborated in Section 3.

798

800

799

- ....
- 801

#### 803 **Boxes**

#### Box 1 | Principle of solar-induced fluorescence (SIF) retrieval

SIF measurements take place outdoors, under ambient sunlight. Accordingly, when pointing a spectroradiometer towards a leaf or plant canopy to make a SIF measurement, we face the challenge that vegetation is highly reflective in the near infrared (NIR) wavelengths, and the signal is dominated by reflected light. The retrieval of SIF from the background reflected radiation is made possible thanks to the spectral properties of incoming light.

The solar spectrum, as measured above a plant canopy, is not continuous; rather, radiation is strongly attenuated within so-called Fraunhofer absorption lines and telluric absorption bands originating from absorption by gases in the Sun's photosphere or the Earth's atmosphere, respectively (see Fig. 1 and an idealized spectral feature in **A**). These features are exploited by the Fraunhofer line depth (FLD) methods<sup>119</sup> where at least four spectral measurements, usually more<sup>120</sup>, are required: the irradiance of the incoming sunlight and the apparent reflected radiance (called apparent, as it includes also SIF), inside and outside of the spectral absorption feature ( $E_{in}/E_{out}$  and  $L_{in}/L_{out}$ , respectively). Since SIF contributes photons similarly both inside and outside the spectral feature (**B**), the relative contribution of SIF to reflected radiation is significantly greater inside the spectral feature, causing an increase in the apparent reflectance (**C**). This increase is proportional to the amount of SIF and can be used to construct a system of equations to retrieve SIF.



Although not mutually exclusive, SIF measurements are often conducted using either the Fraunhofer or Telluric absorption bands, which involve some tradeoffs:

- **Fraunhofer lines** (multiple lines across the SIF spectrum). The advantage of these retrievals lies in their lower sensitivity to atmospheric properties, which is practical for remote measurements as well as applications with variable target-to-sensor distances (e.g., multiangular tower measurements). The main disadvantage is that they require spectrometers with extremely high spectral resolutions and generally require longer periods of signal integration.
- **Telluric bands** (mainly oxygen absorption bands B and A, centered around 687-692 nm; and 759-770 nm, respectively). Since these bands are broader, measurements do not require as high spectral resolution and can be also conducted with shorter integration time, which can be especially suitable for some applications (e.g. drone-based observations). Their main disadvantage is that attention must be paid to corrections for atmospheric absorption (Challenge 6).

805 806

807

- 808
- 809

- 811
- ....
- 812

### Box 2 | Radiative transfer models (RTMs)

- Forward mode. When the required inputs are provided, RTMs are capable of simulating leaf and canopy SIF together with reflected and emitted optical and thermal radiance. Once successfully validated by independent measurements, RTMs can be used in the forward mode to investigate the sensitivity of outputs, (i.e., surface reflectance and SIF) to different structural, biochemical, and physiological inputs, extending our mechanistic understanding of reflected and emitted photons' propagation across scales.
- **Inverse mode.** RTMs can be also inverted (i.e., run backwards) to estimate from laboratory, field and remote sensing spectral data those leaf and canopy traits that match measured reflectance and SIF data.



- **1D models.** 1D leaf RTMs assume that leaf constituents are horizontally homogeneously distributed in vertically stacked plate structures, and hence require only basic morphological and biochemical inputs (e.g., pigment contents driving PAR absorption and within-leaf reabsorption, the intrinsic PSII and PSI fluorescence spectra, and the dynamics in the quantum yield of fluorescence as the mechanistic link to photosynthesis). This simplicity, however, ignores potentially important factors, such as within-leaf heterogeneity or chloroplast movements. As with the 1D leaf construct, 1D canopy RTMs assume that vegetation can be represented by horizontally homogeneous layers filled with leaves of a predefined size, density and geometry (angular distribution), which allows for minimal model inputs and a relatively straightforward application. The 1D architecture has its uses for spatially homogeneous canopies (e.g., crops).
- **3D models.** Structurally complex leaves and spatially heterogeneous plant communities (e.g., forests and savannas) require 3D representations. 3D leaf RTMs can model optical interactions within a genuine 3D digital representation of leaf interior reconstructed, for example, with imaging tomography or confocal microscopy. As demonstrated in the Supplementary Videos 1 and 2, 3D RTM solutions also exist for spatially diversified plant canopies, allowing for accurate physical simulations of APAR<sub>g</sub> and SIF fluxes in complex canopies.

813

814







<b>CRITICAL STEPS/FACTORS</b>		DATA SOURCES	METHODS	SPECTRAL OBSERVATIONS		
INSTRUMENTAL						
<ul> <li>Standardized</li> <li>measurement and calibration protocols</li> </ul>	database	Instrument • characterization and calibration data	<ul><li>Instrument calibrations</li><li>SIF retrieval methods</li></ul>	AT-SENSOR SIF/ REFLECTANCE		
ATMOSPHERIC	/: and				P	
6 7 Atm. corrections of reflectance and SIF observations (including towers and drones)	F community sor network	IF community nsor network	<ul> <li>Atm. databases (e.g. Aeronet)</li> <li>Atm. <i>in-situ</i> measurements Irradiance vertical profiles (drone-based)</li> </ul>	Atmospheric radiative transfer models (RTMs)	ATMOSPHERE CORRECTED SIF/REFLECTANCE	ROCESSING
STRUCTURAL	al S e se				୭	
<ol> <li>Improved parameterization of leaf and canopy RTMs with morphological and functional data</li> <li>Standardization of canopy SIF in terms of efficiency</li> </ol>	Glok Cal/Val multiscal	<ul> <li>3D canopy structural data (e.g. terrestrial and airborne LiDAR)</li> <li>3D leaf morphology</li> <li>Leaf and canopy empirical data for model calibration/validation (including PSI and PSII ChlaF spectra)</li> </ul>	<ul> <li>Canopy &amp; leaf RTMs</li> <li>Spectral invariant</li> <li>probability-based</li> <li>reflectance corrections</li> </ul>	STRUCTURE CORRECTED SIF/REFLECTANCE	INTERPRETATIO	
PHYSIOLOGICAL					ž	
<ul> <li>Synergistic use of complementary/ancillary data</li> <li>Full characterization of dynamics photosynthetic energy partitioni between SIF and GPP</li> <li>Comprehensive campaigns: long</li> <li>term, multiple species, stressors and biomes</li> </ul>	s in ng -	<ul> <li>Leaf and plant empirical data for model calibration/validation:</li> <li><i>pigment content and dynamics</i></li> <li><i>PSI and PSII ChlaF contributions</i></li> <li><i>energy partitioning between SIF</i> <i>and GPP</i></li> <li>Data from a wide range of organisms (functional diversity)</li> </ul>	Process-based model development Empirical models Versatile protocols to resolve PSI contribution and full energy partitioning Multisource data assimilation frameworks	ECOSYSTEM, CANOPY AND LEAF PHOTOSYNTHETIC TRAITS		

