

MEASURING PLANT STRUCTURE AND FUNCTION USING OPTICAL REMOTE SENSING

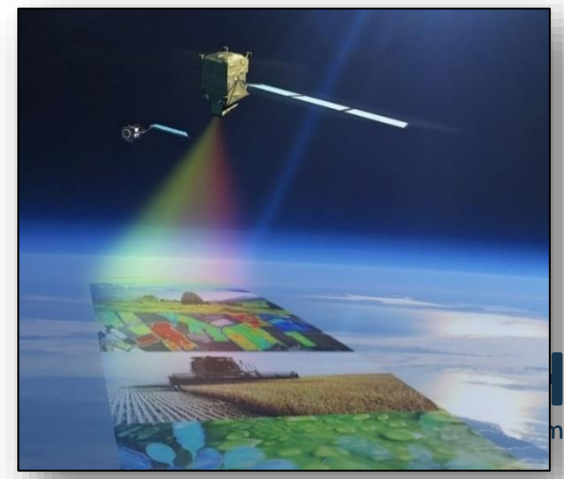
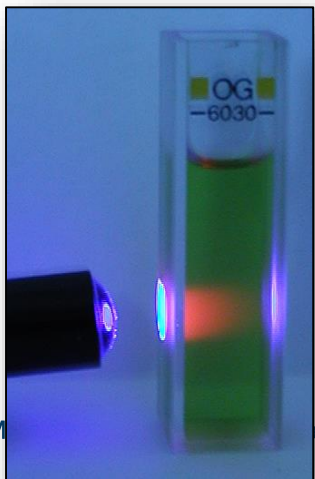
current status and recent developments of airborne and satellite remote sensing and their potential for disease detection

Uwe Rascher – Research Centre Jülich (Forschungszentrum Jülich), IBG-2: Plant Sciences, Germany

19. AUGUST 2023

Outline

- ❑ Optical remote sensing – measuring with light (photons) and the challenge of scale
- ❑ Laboratory studies – transferability to the field – increasing the scale with drones and aircrafts – possibilities and limitations for satellite based remote sensing
- ❑ Disease detection with optical approaches – spectrally resolved measurements, fluorescence and thermal approaches
- ❑ Possibilities to detect diseases on the large scale – the balance between spectral, spatial and temporal resolution
- ❑ Direct detection of disease symptoms and confounding factors

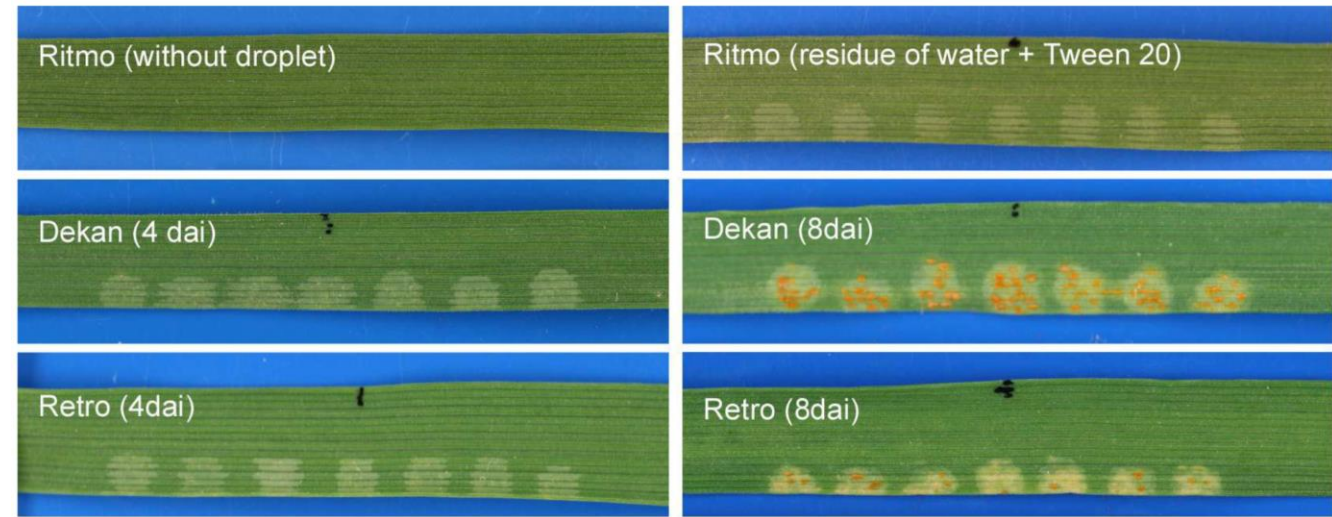


Disease detection using spectrally resolved and fluorescence imaging - going back 10-15 years and into the laboratory

- Well-controlled laboratory studies to develop approaches for disease detection with optical sensors

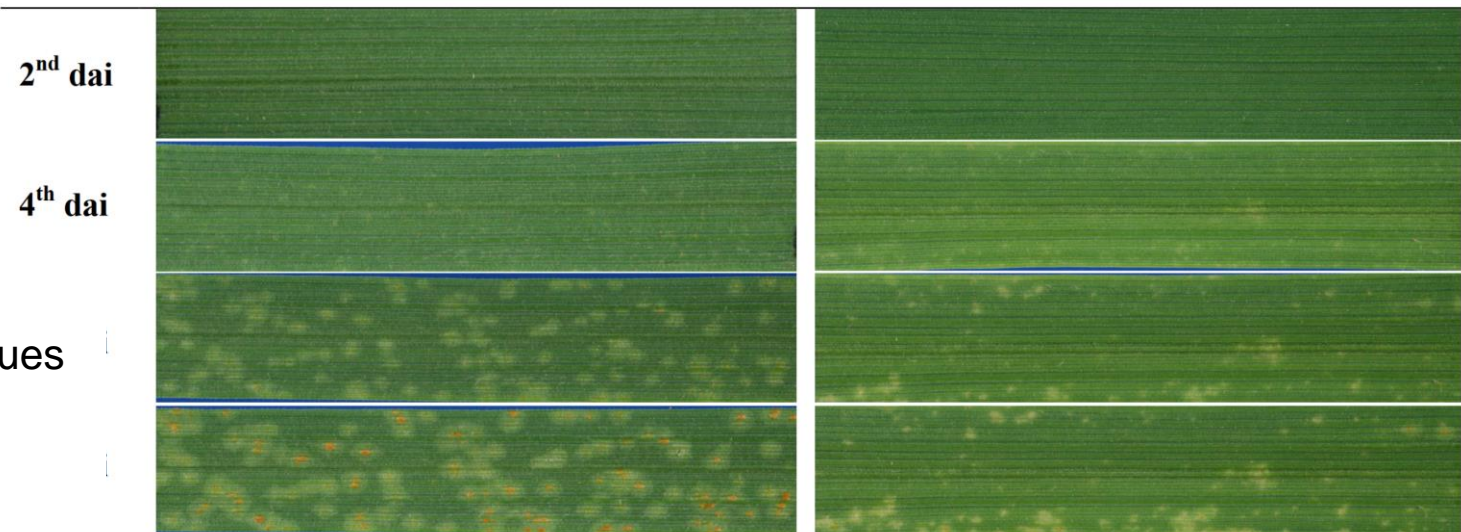
wheat leaves 4 and 8 days after inoculation with *P. triticina*

Leaf Rust on wheat leaves
susceptible cultivar, Dekan (left)
resistant cultivar, Retro (right)



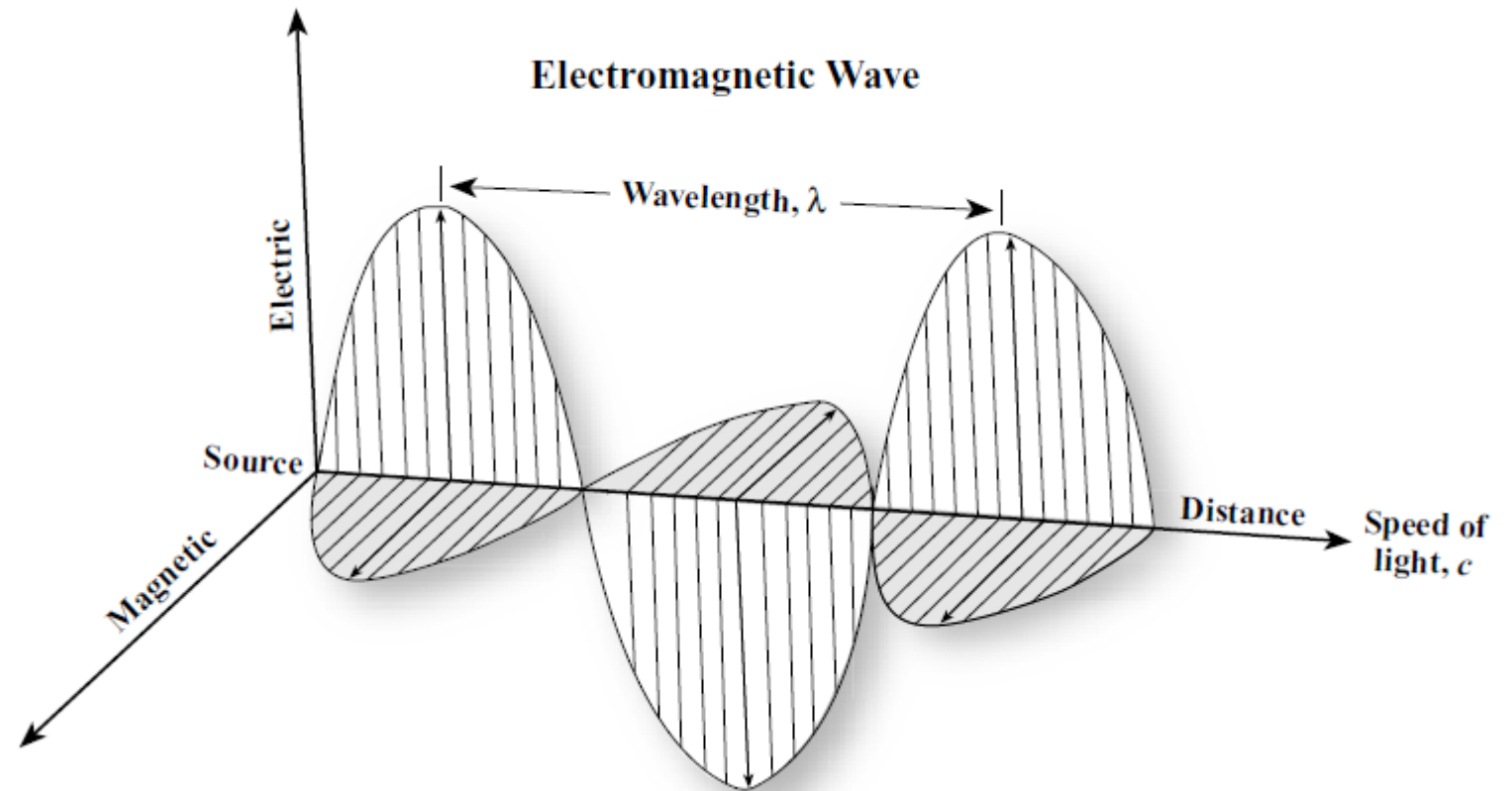
Dekan

Retro



What is light – a concept for the non-physicist

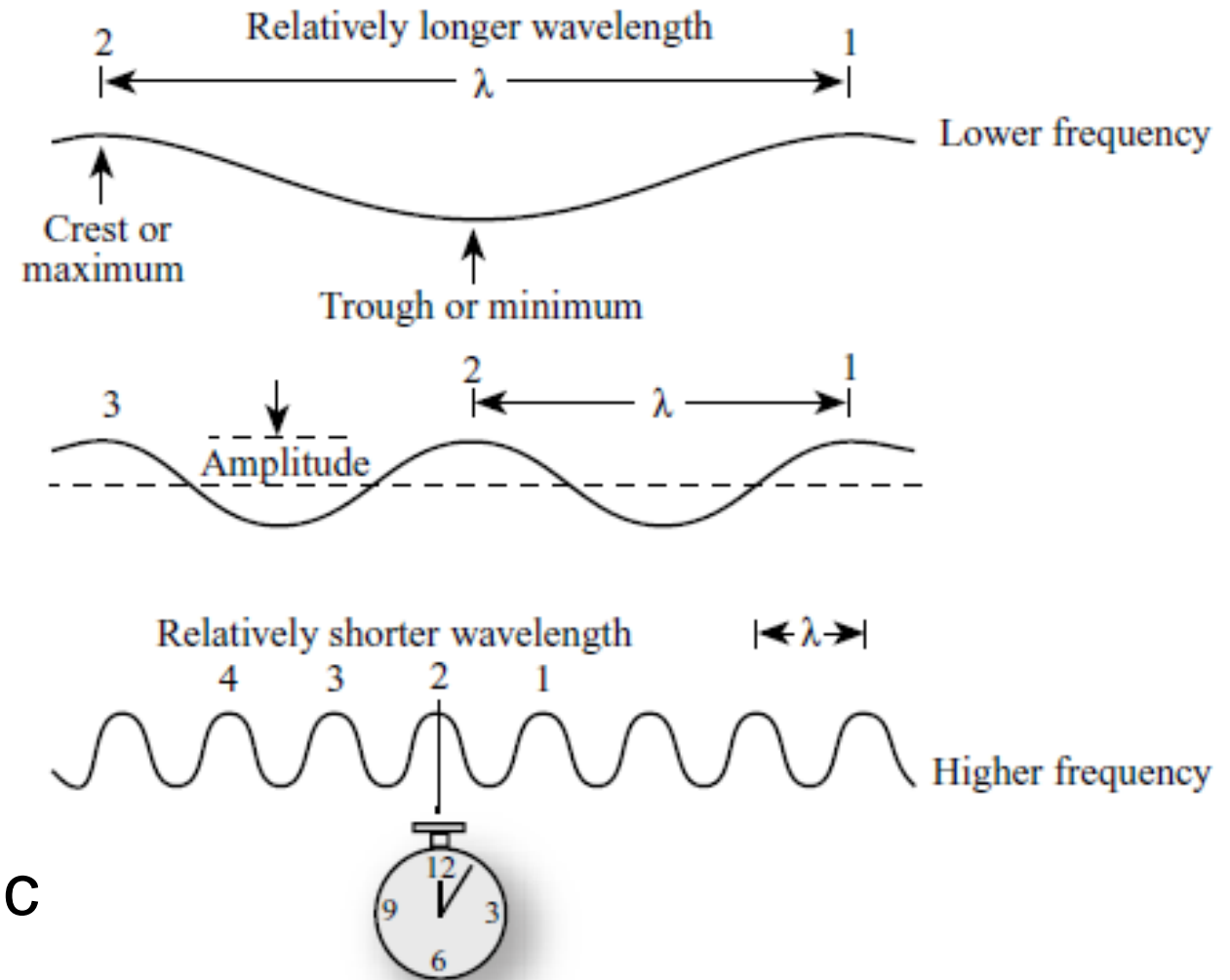
- „Light is a flux of photons, which travel with light speed and which can best be described with a wave function
- Speed of light (c) is constant and is 300 000 km/s



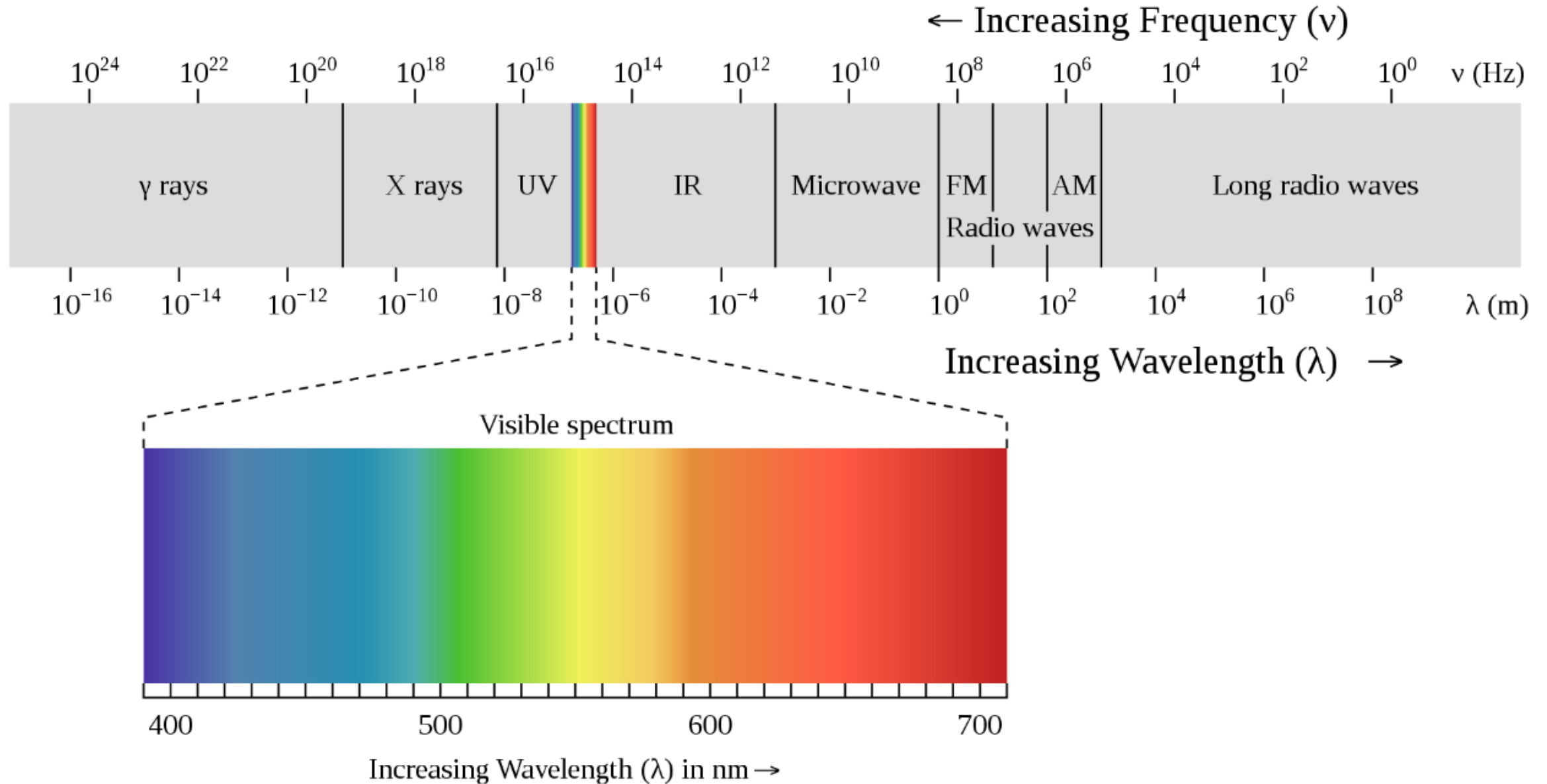
What is light – a concept for the non-physicist

- The amplitude of the wave is \pm constant
- The wavelength (λ) describes the distance between two maxima [measured in m]
- The frequency (ν) is the number of oscillations per seconds [measured in Hz]
- $\lambda = c / \nu$
- The higher the frequency the more energy an electromagnetic radiation carries

Inverse Relationship Between Wavelength and Frequency



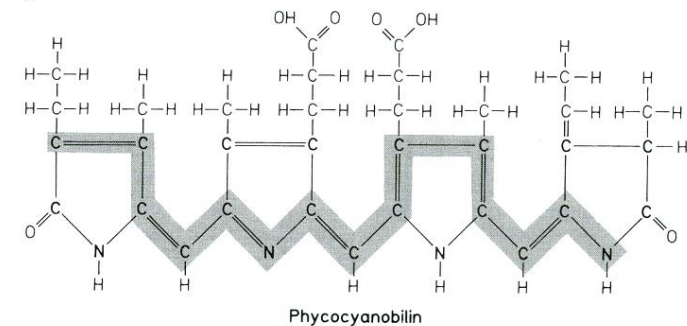
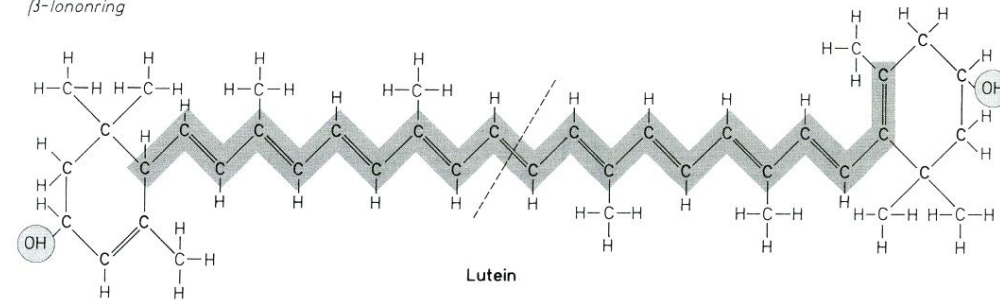
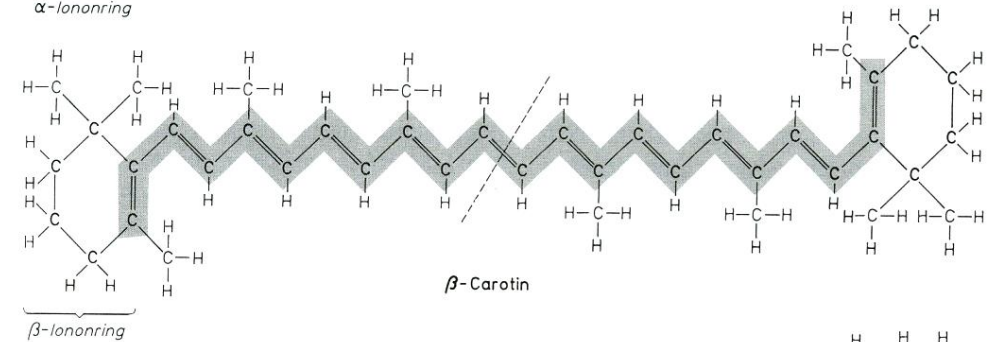
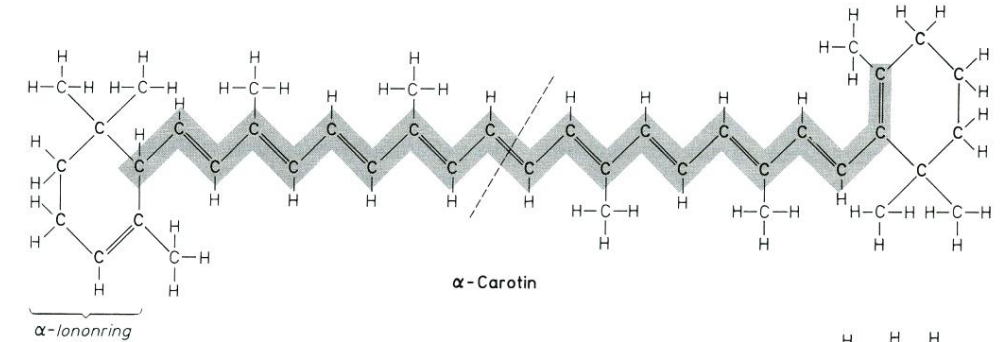
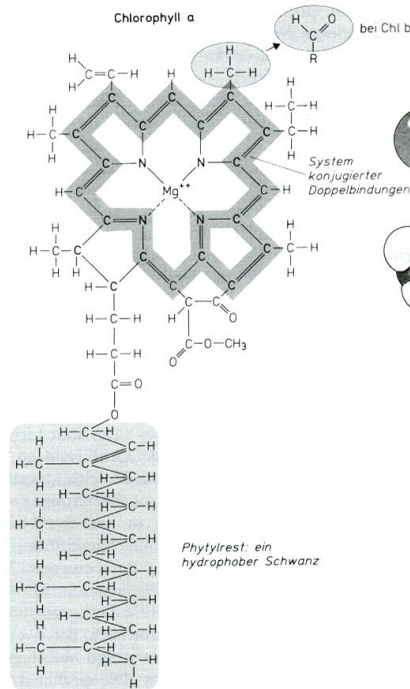
What is light – a concept for the non-physicist



Light absorption: Lets have a closer look – what happens if photons interact with a leaf

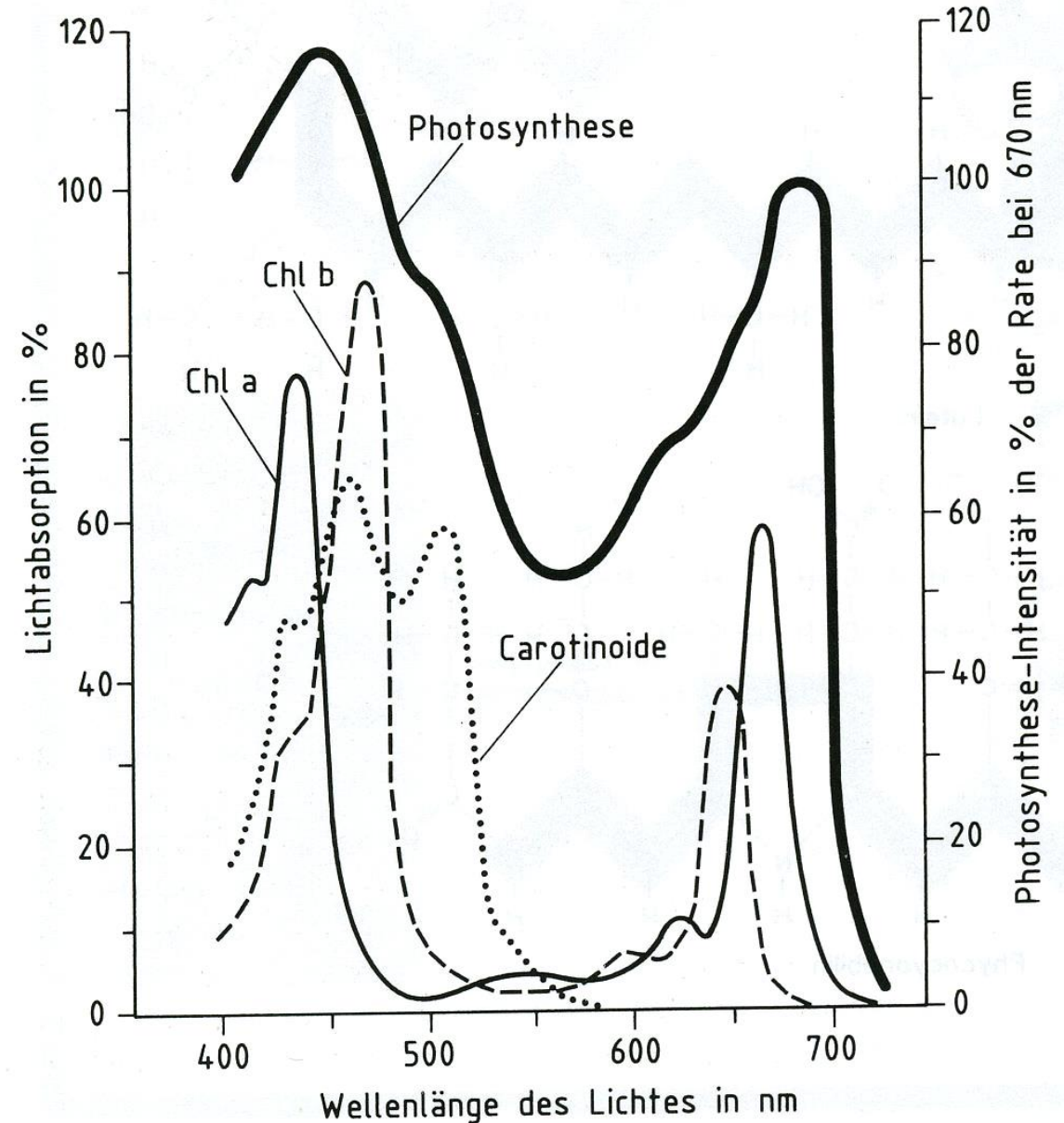
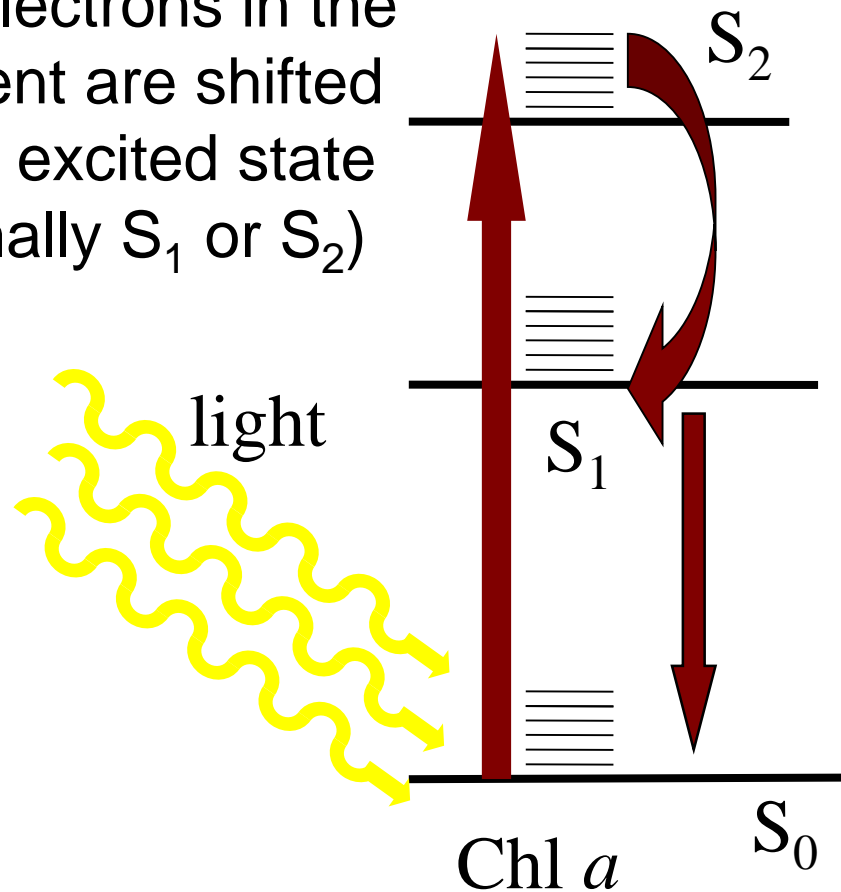
➤ Light interacts with pigments – photons in the visible spectral window are absorbed by plant pigments

- Chlorophylls
 - chlorophyll a
 - chlorophyll b
- Carotenoids
 - α carotin
 - β carotin
 - lutein
 - violaxanthin
 - antheaxanthin
 - ...
- Anthocyanins
- Betalaine



Light absorption: Lets have a closer look – what happens if photons interact with a leaf

Light absorption means that the energy in the photon is transferred to the pigment, the photon disappears and electrons in the pigment are shifted to the excited state (normally S_1 or S_2)

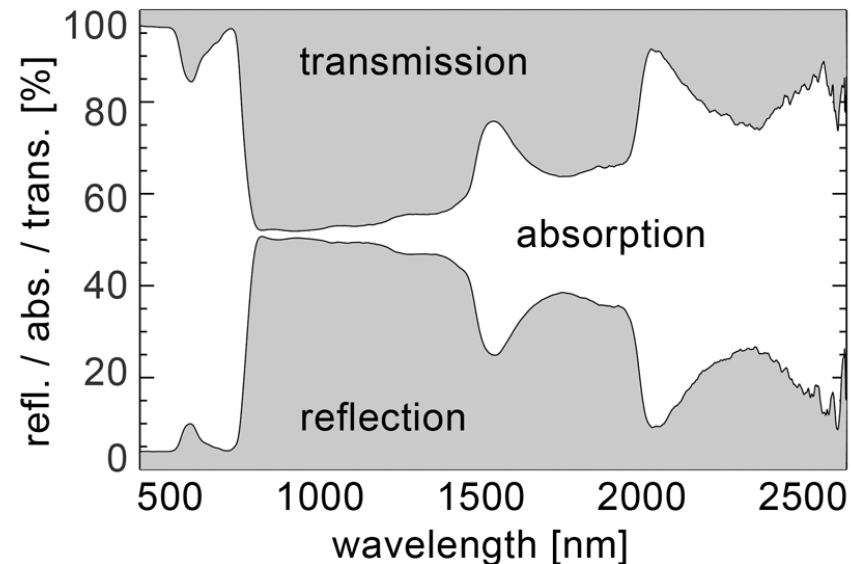
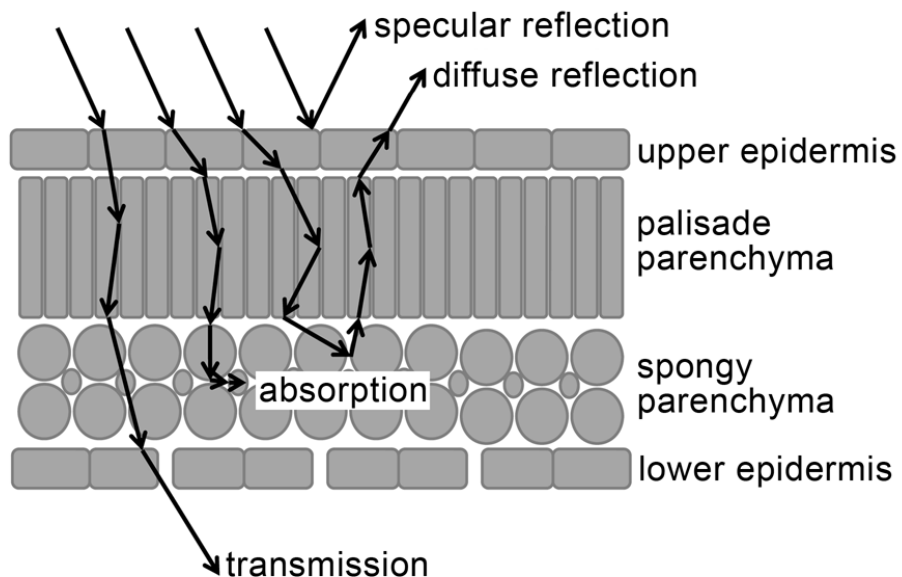


Light absorption: Lets have a closer look – what happens if photons interact with a leaf

- Photons describe a complex path within a leaf, but at the end they are either **absorbed**, **reflected** or **transmitted**

$$\text{Reflection [\%]} = \frac{\text{Reflected irradiance [energy unit]}}{\text{Incoming irradiance [energy unit]}}$$

$$\text{Transmittance [\%]} = \frac{\text{Transmitted irradiance [energy unit]}}{\text{Incoming irradiance [energy unit]}}$$



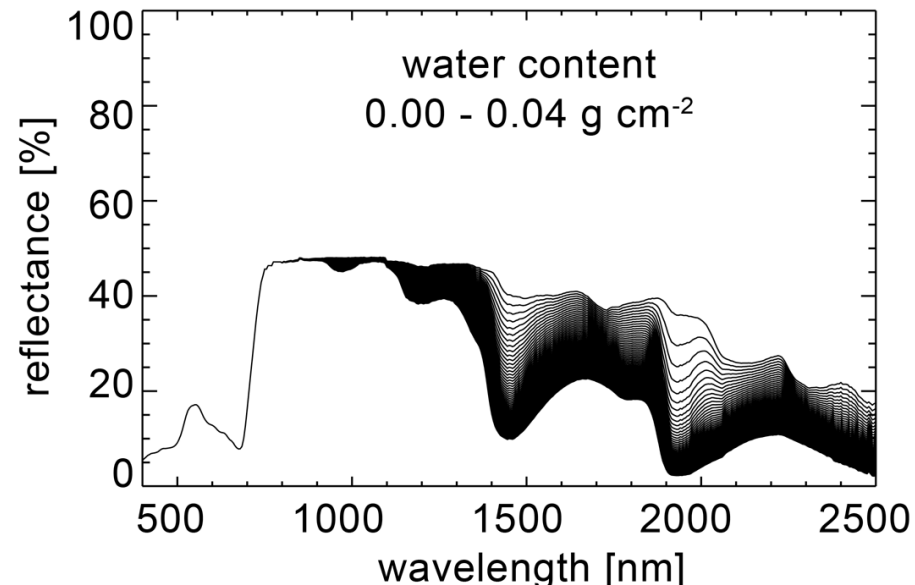
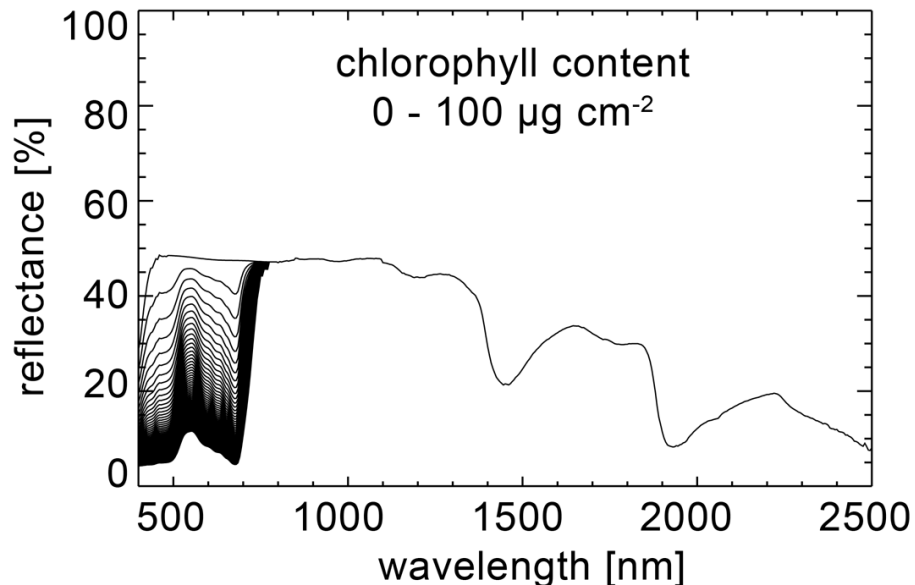
Rascher et al. (2010) Sensing of photosynthetic activity of crops. In Precision Crop Protection - the Challenge and Use of Heterogeneity. Springer Science+Business Media B.V., doi 10.1007/978-90-481-9277-9_6.

Light absorption: Lets have a closer look – what happens if photons interact with a leaf

- Photons describe a complex path within a leaf, but at the end they are either **absorbed**, **reflected** or **transmitted**

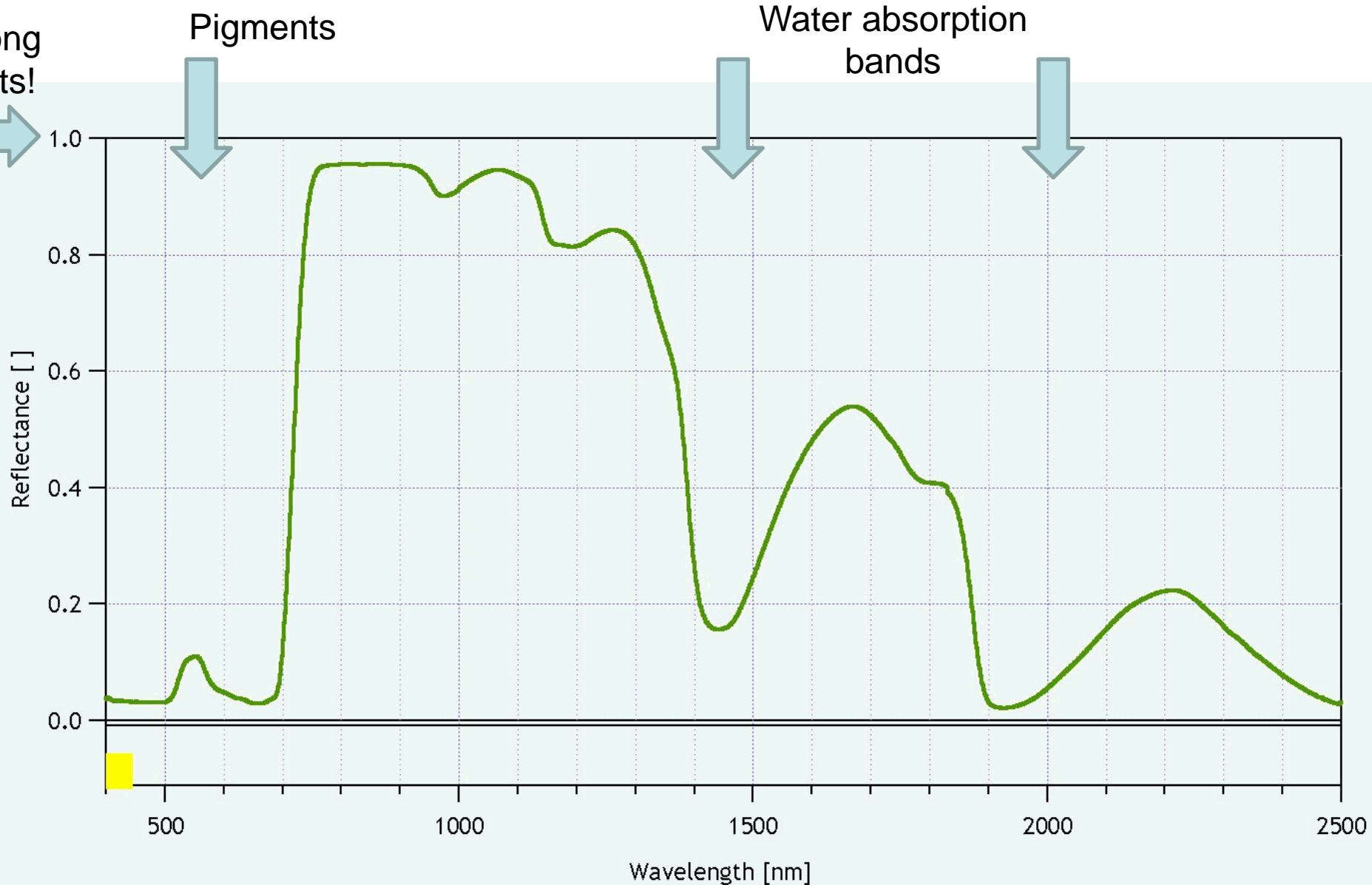
$$\text{Reflection [\%]} = \frac{\text{Reflected irradiance [energy unit]}}{\text{Incoming irradiance [energy unit]}}$$

$$\text{Transmittance [\%]} = \frac{\text{Transmitted irradiance [energy unit]}}{\text{Incoming irradiance [energy unit]}}$$



Light absorption: Lets have a closer look – what happens if photons interact with a leaf

wrong units!
→

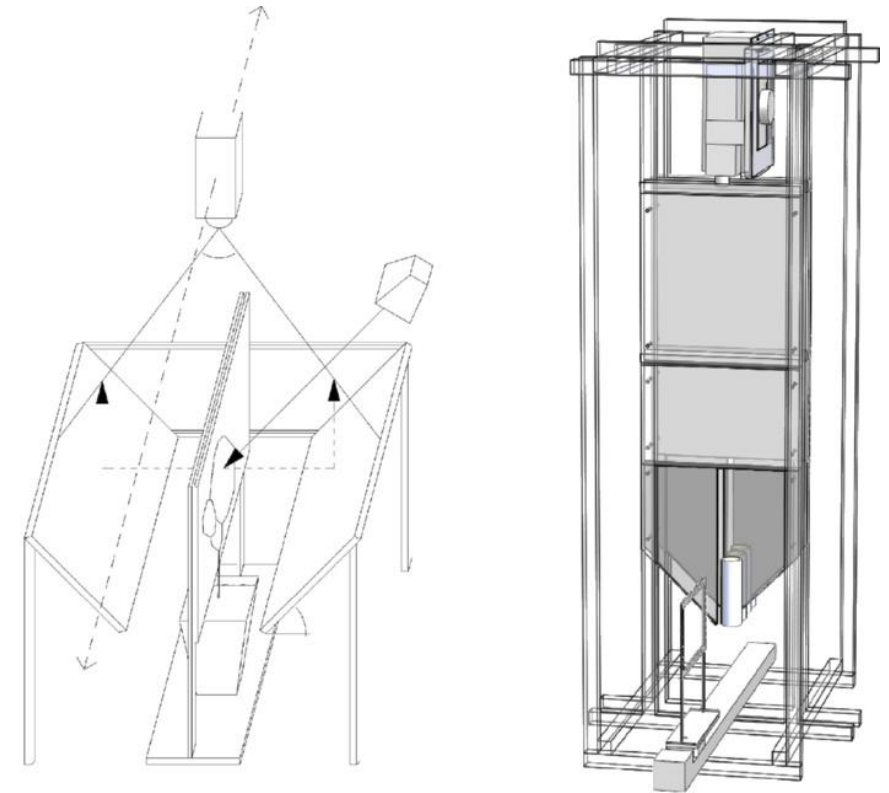


8 hours during which the healthy leaf
(i) dries out and
(ii) loses its pigments



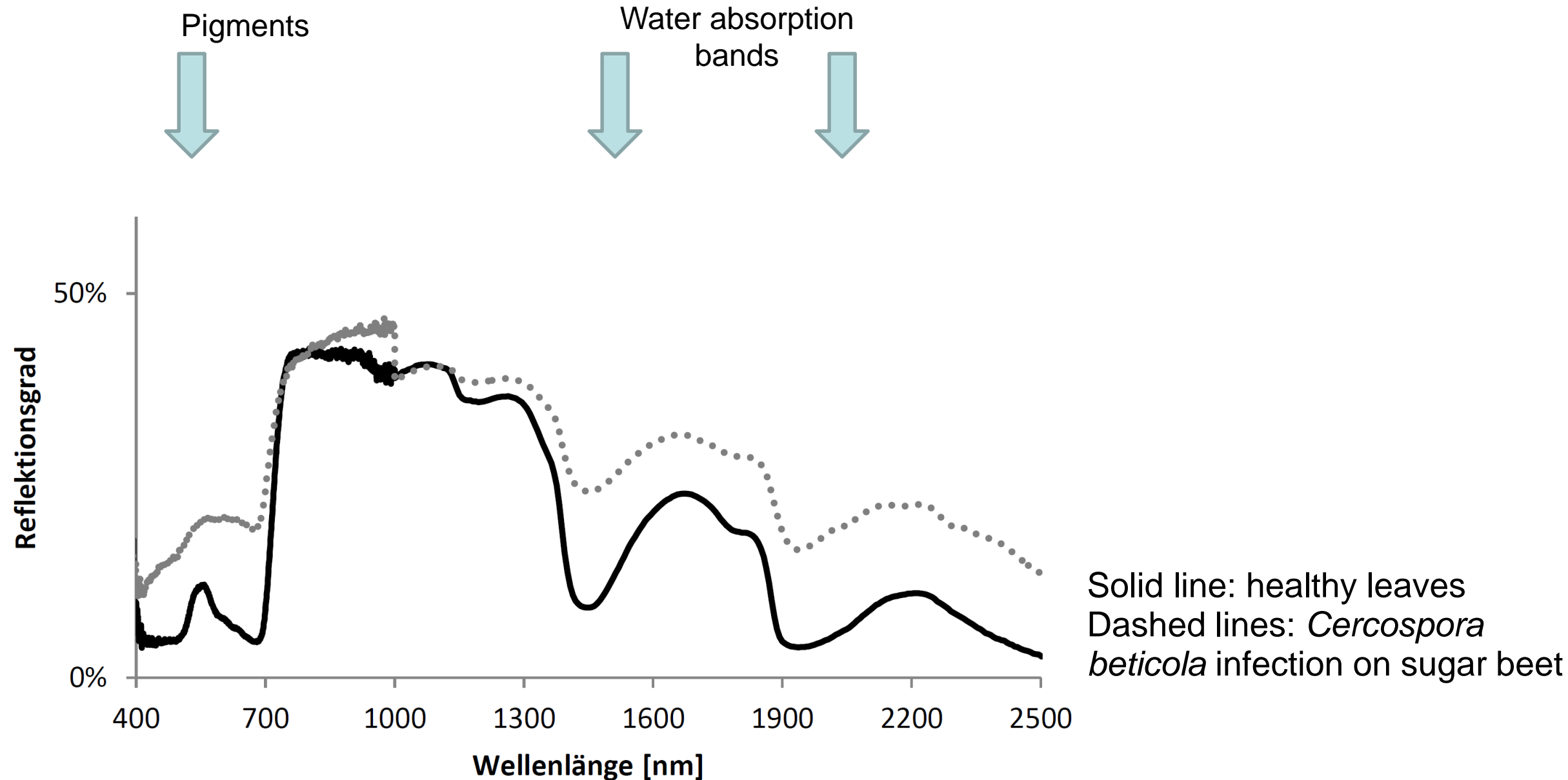
Light absorption / transmission / reflection changes after disease infection – a case study

- ❑ Disease infection leaves a characteristic signature in reflectance / transmission / absorbance

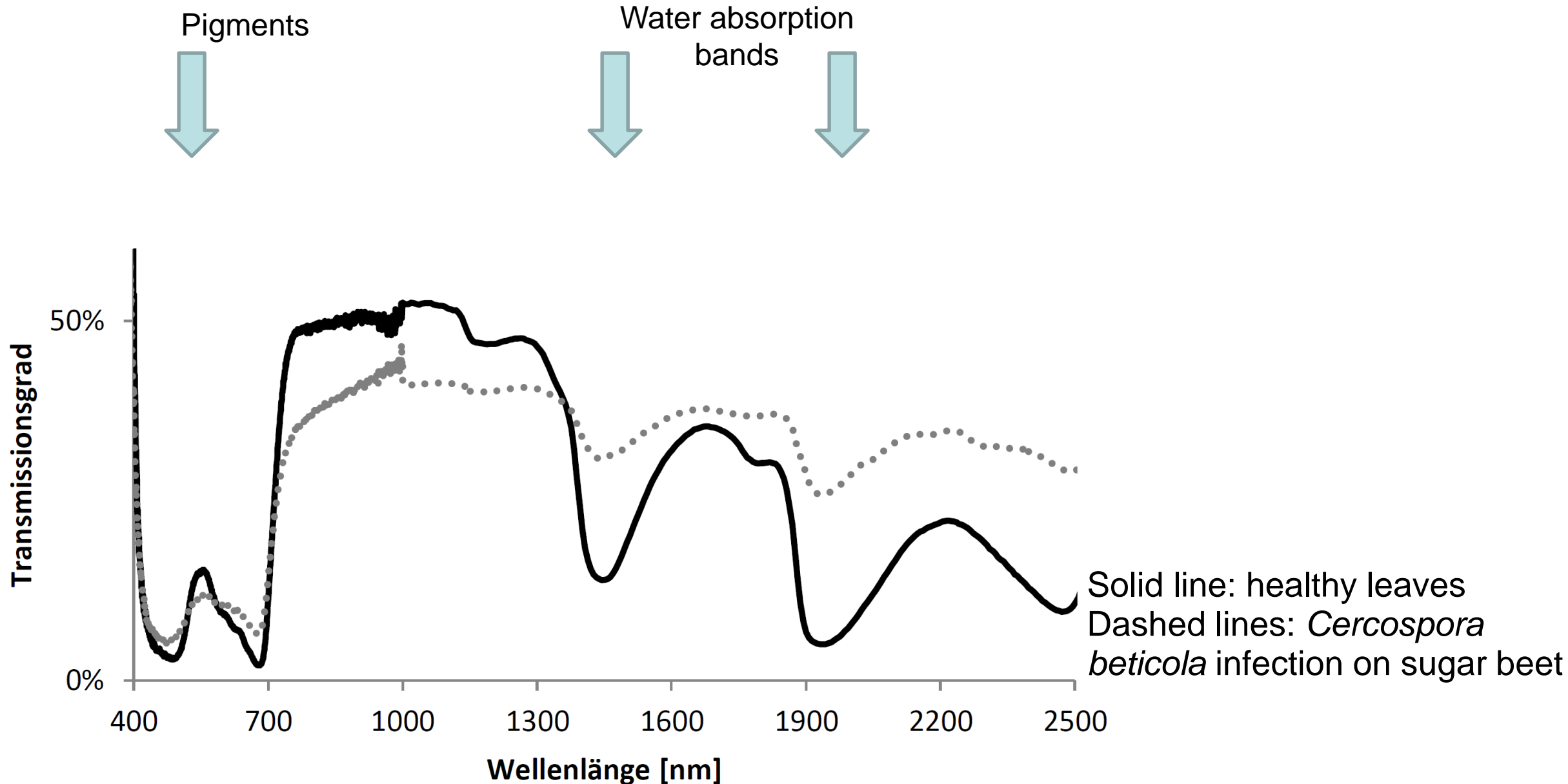


$$1 - \tau = \rho + \alpha$$

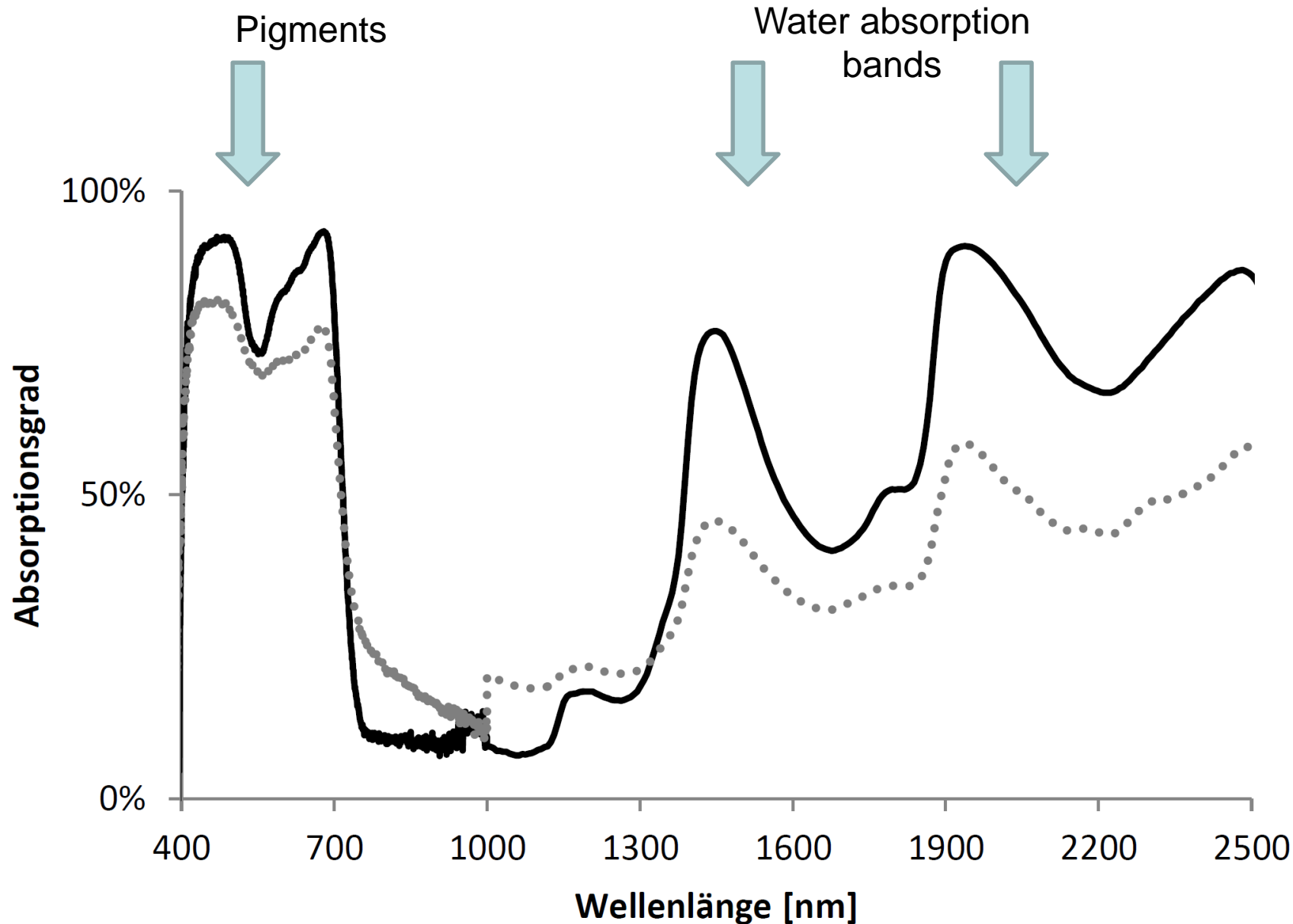
Light absorption / transmission / reflection changes after disease infection – a case study



Light absorption / transmission / reflection changes after disease infection – a case study



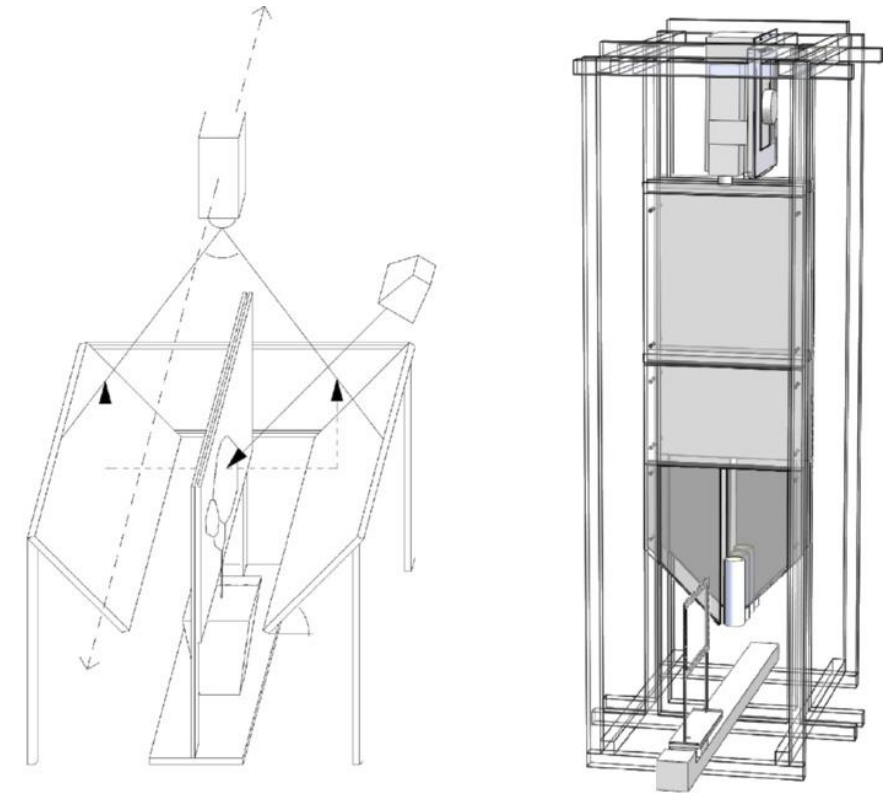
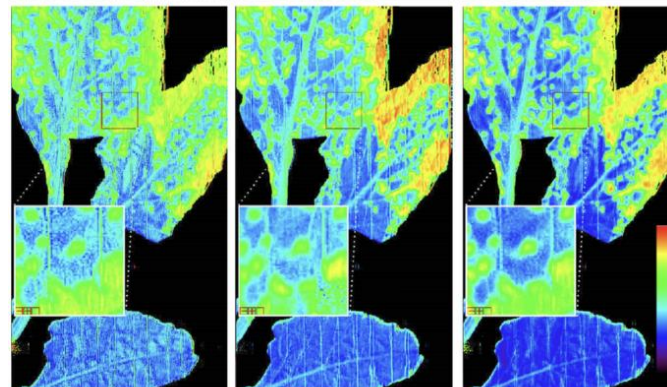
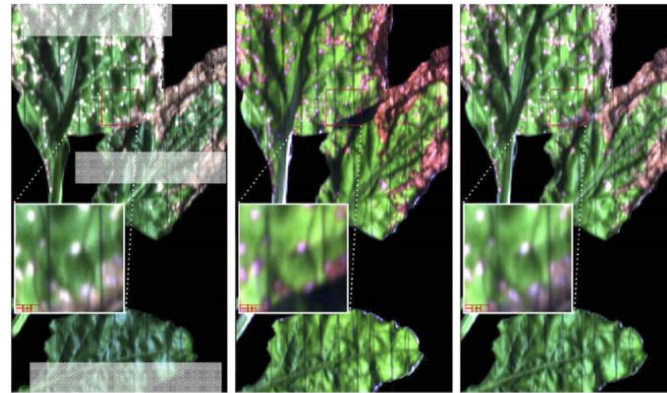
Light absorption / transmission / reflection changes after disease infection – a case study



Solid line: healthy leaves
Dashed lines: *Cercospora beticola* infection on sugar beet

Light absorption / transmission / reflection changes after disease infection – a case study

- ❑ Disease infection leaves a characteristic signature in reflectance / transmission / absorbance
- ❑ High performance spectroscopy has been shown to detect plant disease symptoms



$$1 - \tau - \rho = \alpha$$

Bergsträsser et al. (2015) HyperART: non-invasive quantification of leaf traits using hyperspectral absorption-reflectance-transmittance imaging. *Plant Methods*, 11:1, doi:10.1186/s13007-015-0043-0

Vegetation Indices: An easy way to empirically translate spectrally resolved reflectance into plant traits

1. Classical greenness indices

- i. measure chlorophyll and LAI
- ii. widely used and can be derived from broad band sensors
- iii. require red and infra-red bands only

$$\text{Simple Ratio (SR)} = \frac{R_{\text{infra-red}}}{R_{\text{red}}}$$

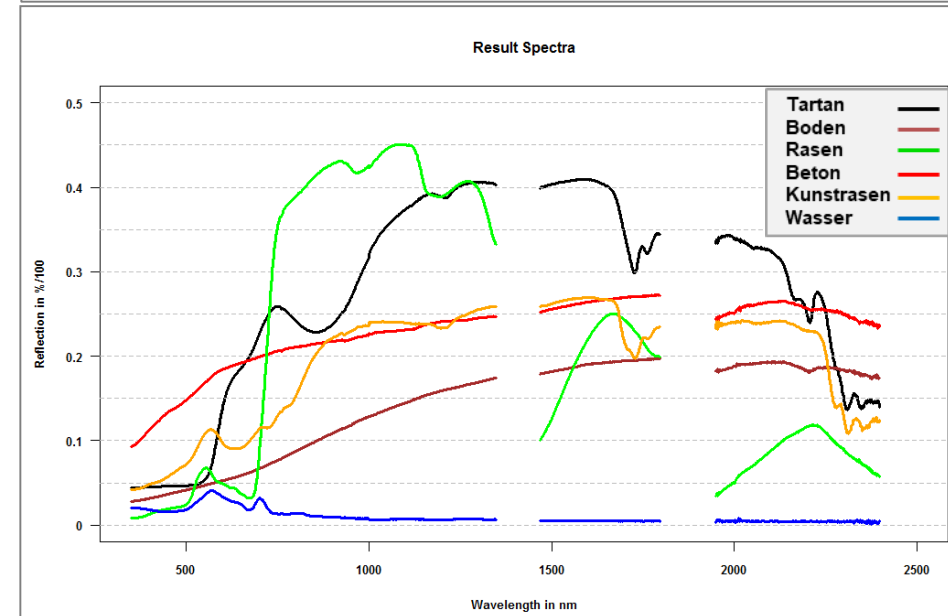
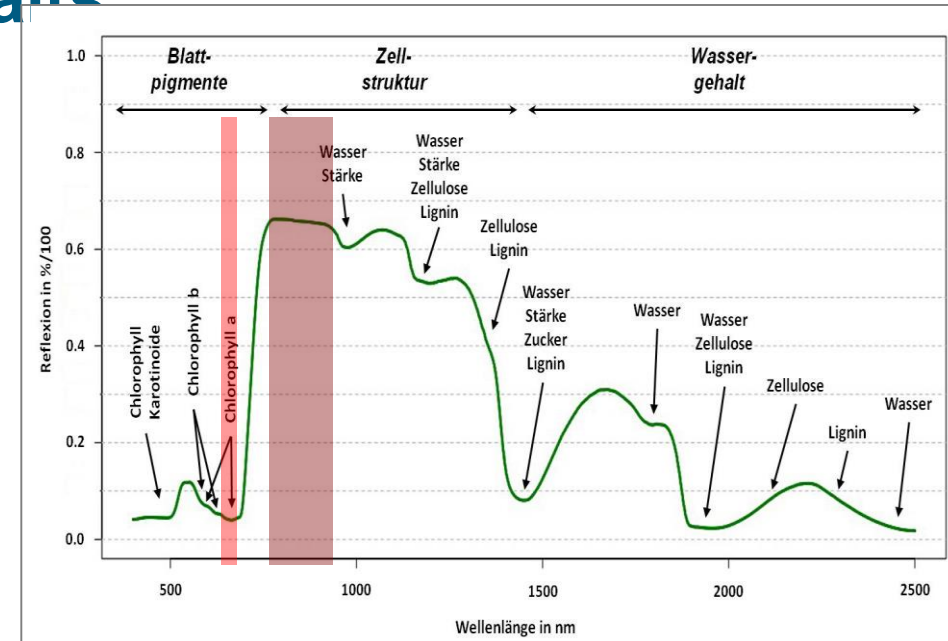
value range

0 - 7

$$\text{NDVI} = \frac{R_{\text{infra-red}} - R_{\text{red}}}{R_{\text{infra-red}} + R_{\text{red}}}$$

0 - 1

Normalized Difference Vegetation Index (NDVI) is independent to additive and multiplicative shifts in the data



Vegetation Indices: An easy way to empirically translate spectrally resolved reflectance into plant traits

2. More advanced greenness indices

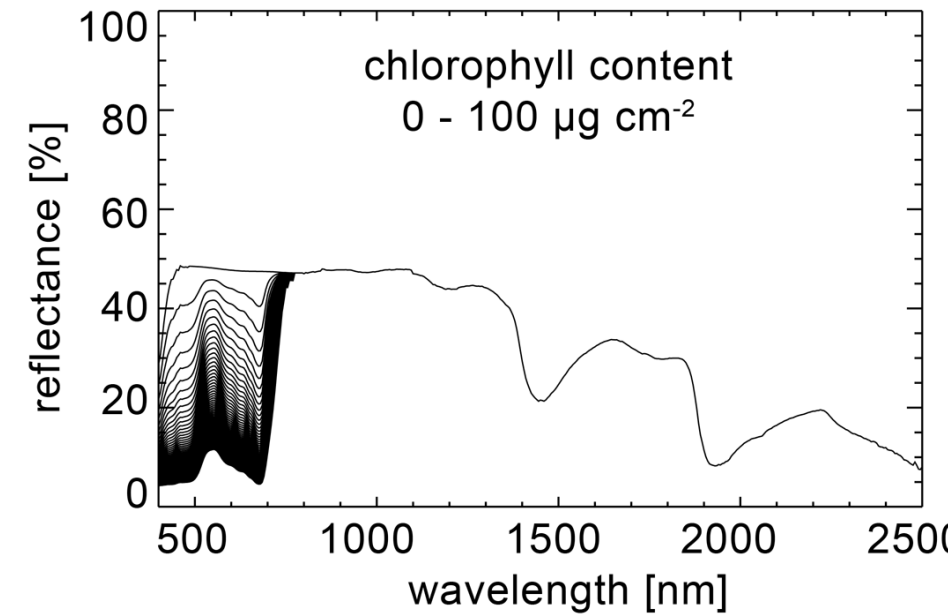
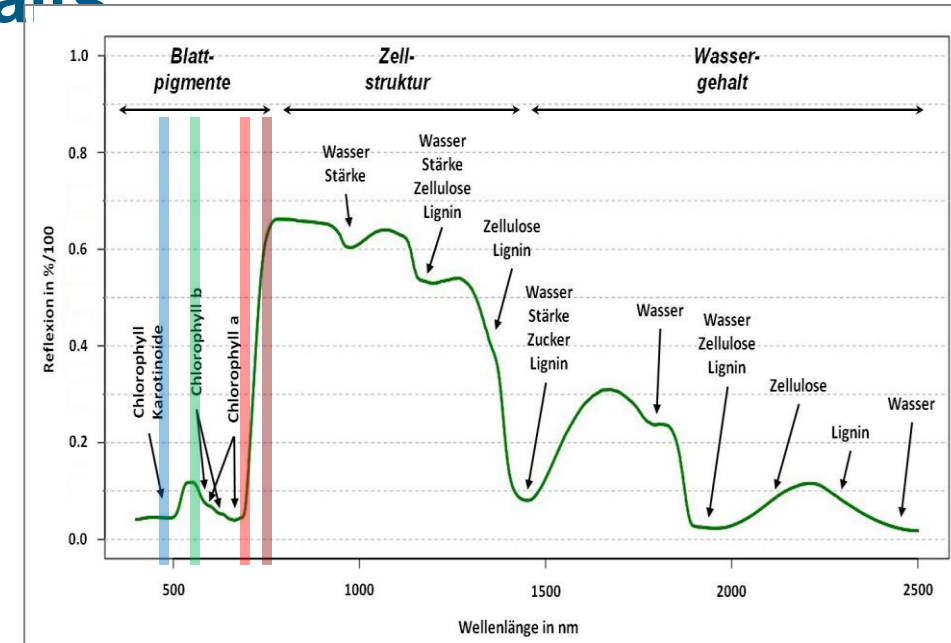
- i. measure chlorophyll and LAI
- ii. can also be used at dense vegetation
- iii. use smaller spectral windows or additional spectral bands

$$NDVI_{re} = \frac{R_{735-750} - R_{695-710}}{R_{735-750} + R_{695-710}}$$

$$EVI = 2.5 \left[\frac{R_{795-810} - R_{665-680}}{R_{795-810} + 6R_{665-680} - 7.5R_{475-490} + 1} \right]$$

$$TCARI = 3 \left[(R_{700 \pm 4} - R_{670 \pm 4}) - 0.2(R_{700 \pm 4} - R_{550 \pm 4}) \left(\frac{R_{700 \pm 4}}{R_{670 \pm 4}} \right) \right]$$

NDVI_{re}: red-edge Normalized Difference Vegetation Index
 EVI: Enhanced Vegetation Index
 TCARI: Transformed Chlorophyll Absorption Reflectance Index



Vegetation Indices: An easy way to empirically translate spectrally resolved reflectance into plant traits

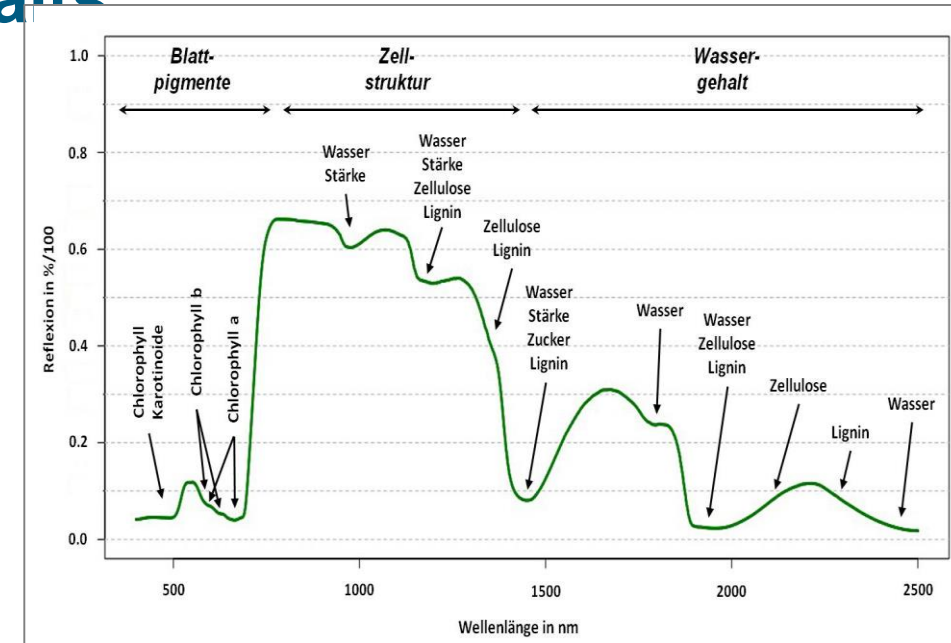
3. Other plant pigments / components

i. large number of 'dedicatd' indices

ii. Examples are

- NDNI: Normalized Difference Nitrogen Index (N absorption at 1510 nm)
- NDLI: Normalized Difference Lignin Index (lignin absorption at 1754 nm)
- CAI: Cellulose Absorption Index (absorption at 2000 – 2200 nm)
- PSRI: Plant Senescence Index (empirically correlated with senescens and fruit ripening)
-
- *CLSI*: *Cercospora* Leaf Spot Index
-

iii. use with extreme care as many of these indices are not specific

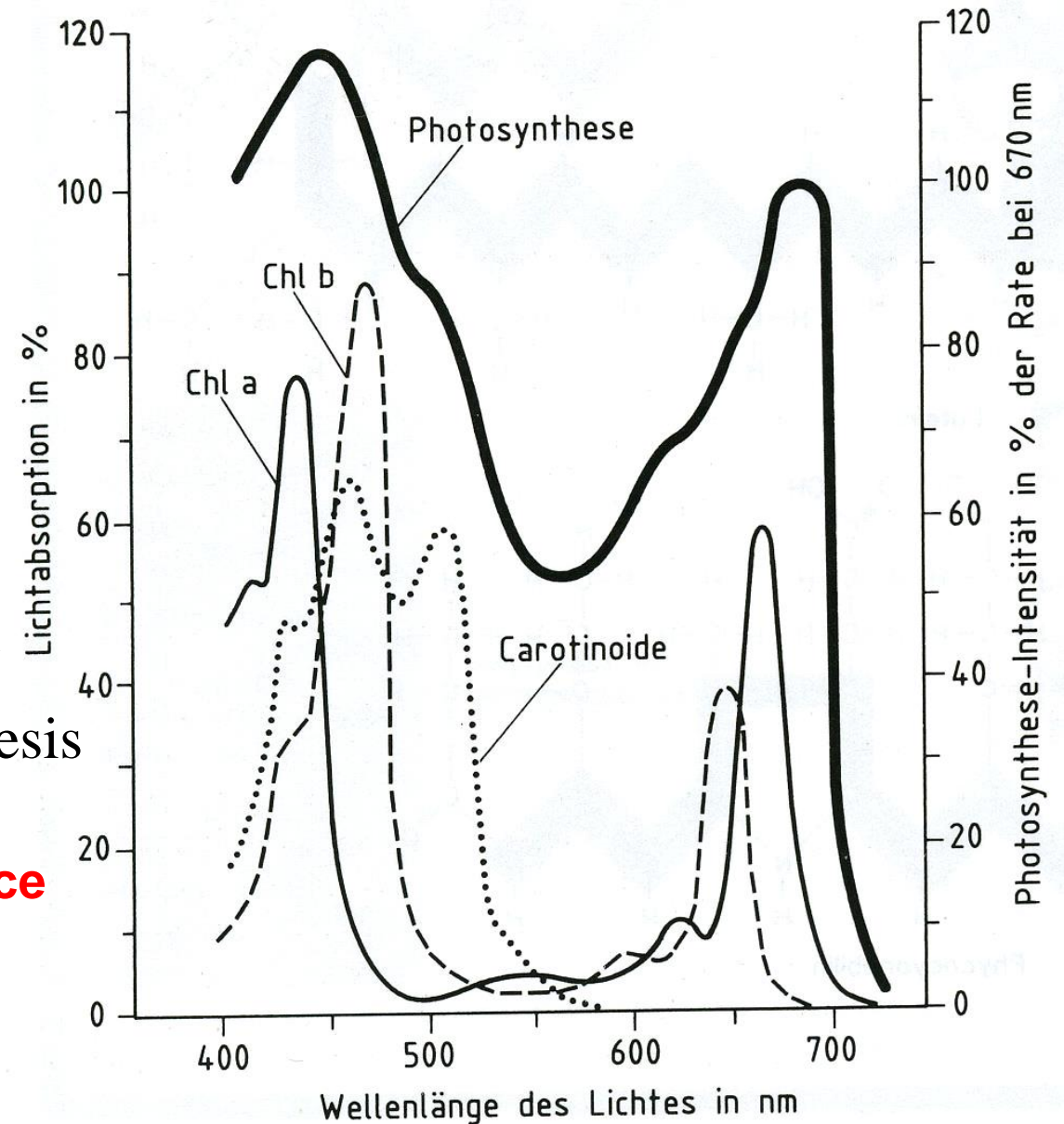
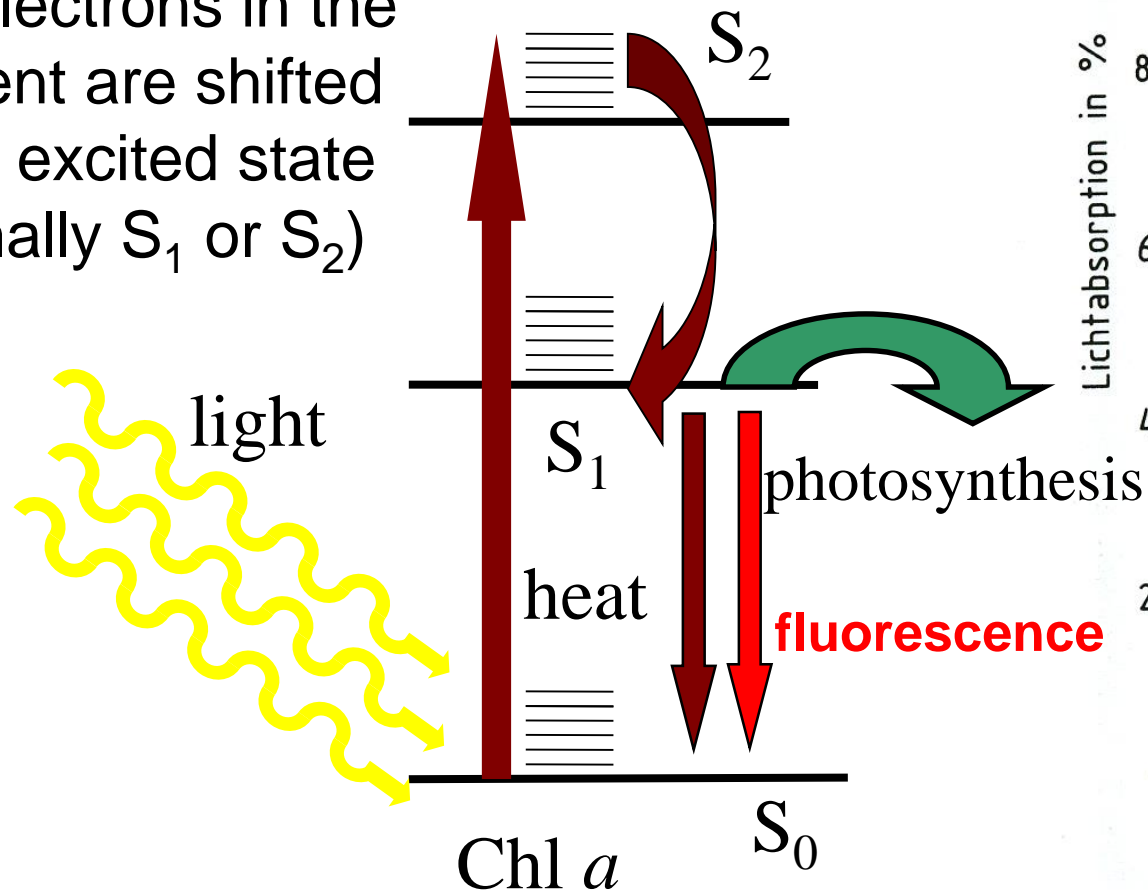


$$CLSI = \frac{R_{698} - R_{570}}{R_{698} + R_{570}} + R_{734}$$

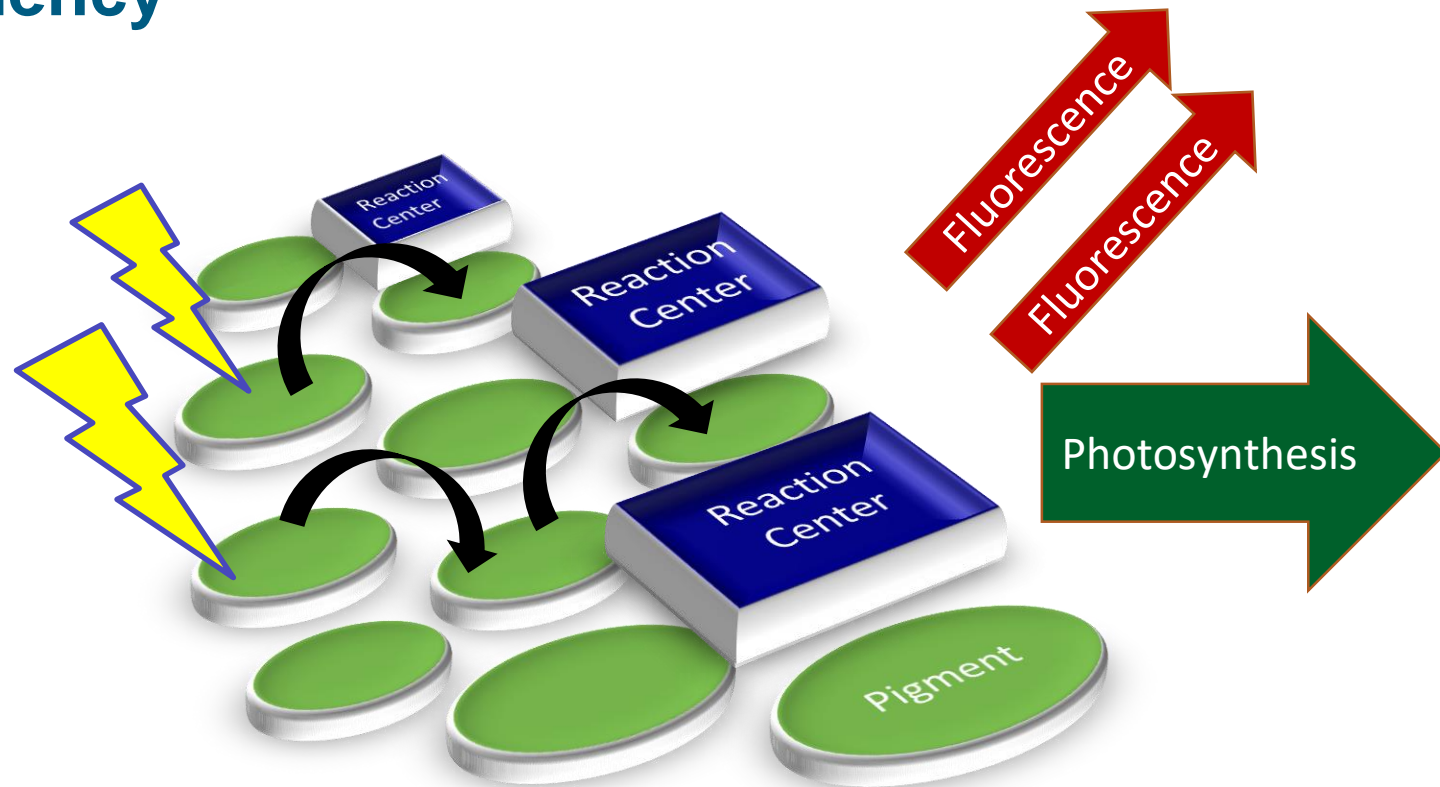
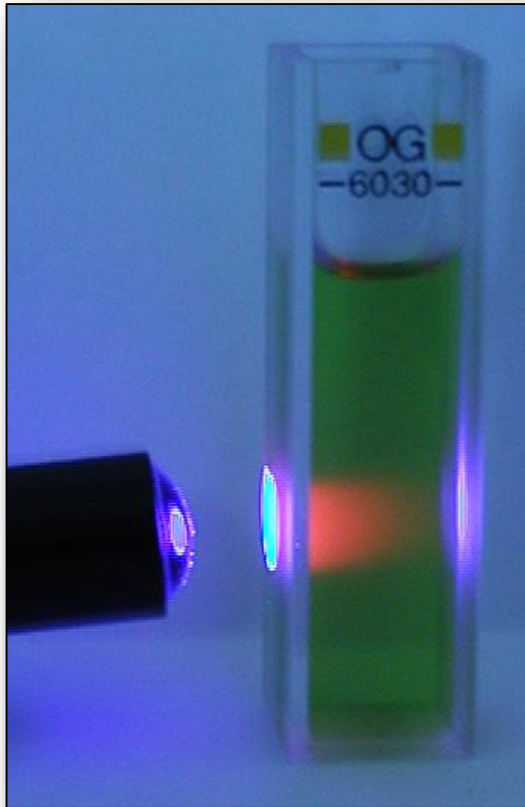
Mahlein, A. K., Rumpf, T., Welke, P., Dehne, H. W., Plümer, L., Steiner, U. and Oerke, E. C. (2013). Development of spectral indices for detecting and identifying plant diseases. *Remote Sensing of Environment* **128**, 21-30.

Light absorption: Lets have a closer look – what happens if photons interact with a leaf

Light absorption means that the energy in the photon is transferred to the pigment, the photon disappears and electrons in the pigment are shifted to the excited state (normally S_1 or S_2)



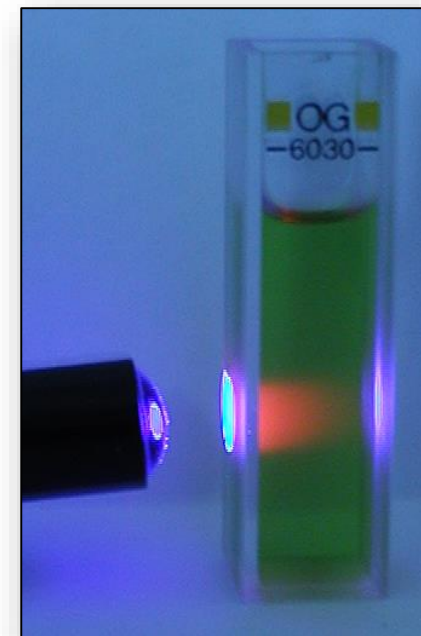
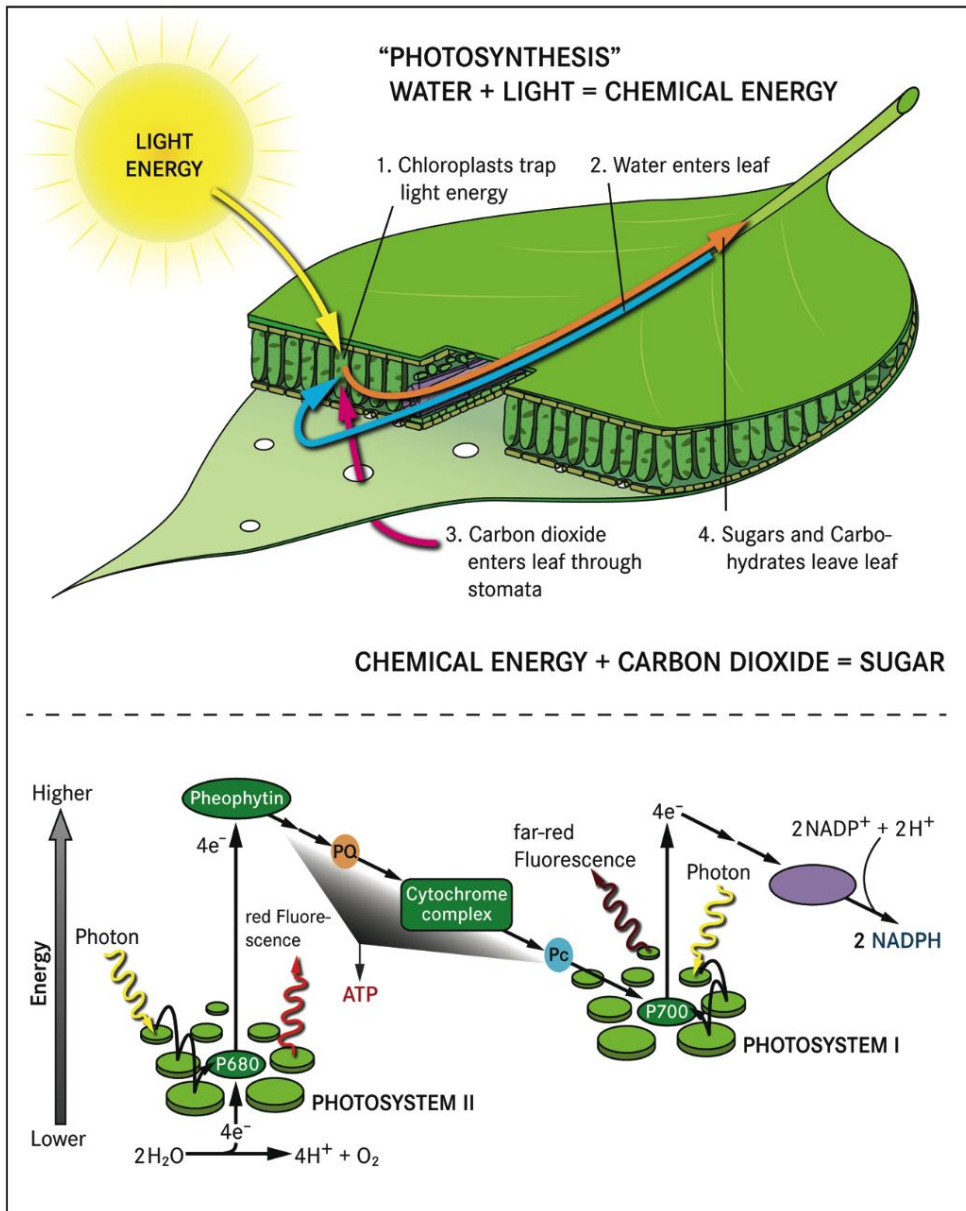
The origin of fluorescence – an indicator for photosynthetic efficiency



1. Chlorophyll molecules emit fluorescence. The intensity of the fluorescence signal is a function of light intensity and the concentration of chlorophyll
2. Additionally, the functional status of photosynthesis modulates the intensity of the fluorescence signal

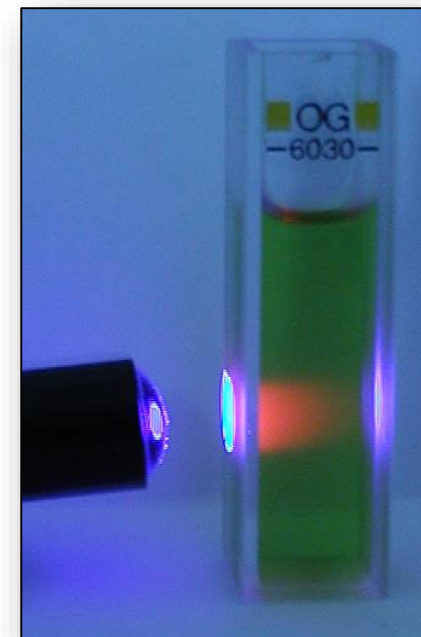
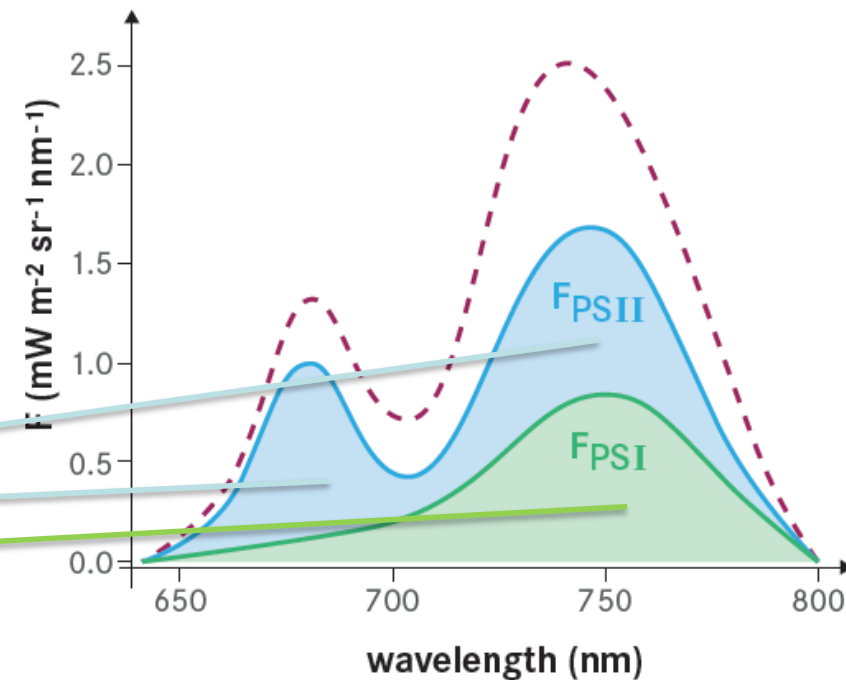
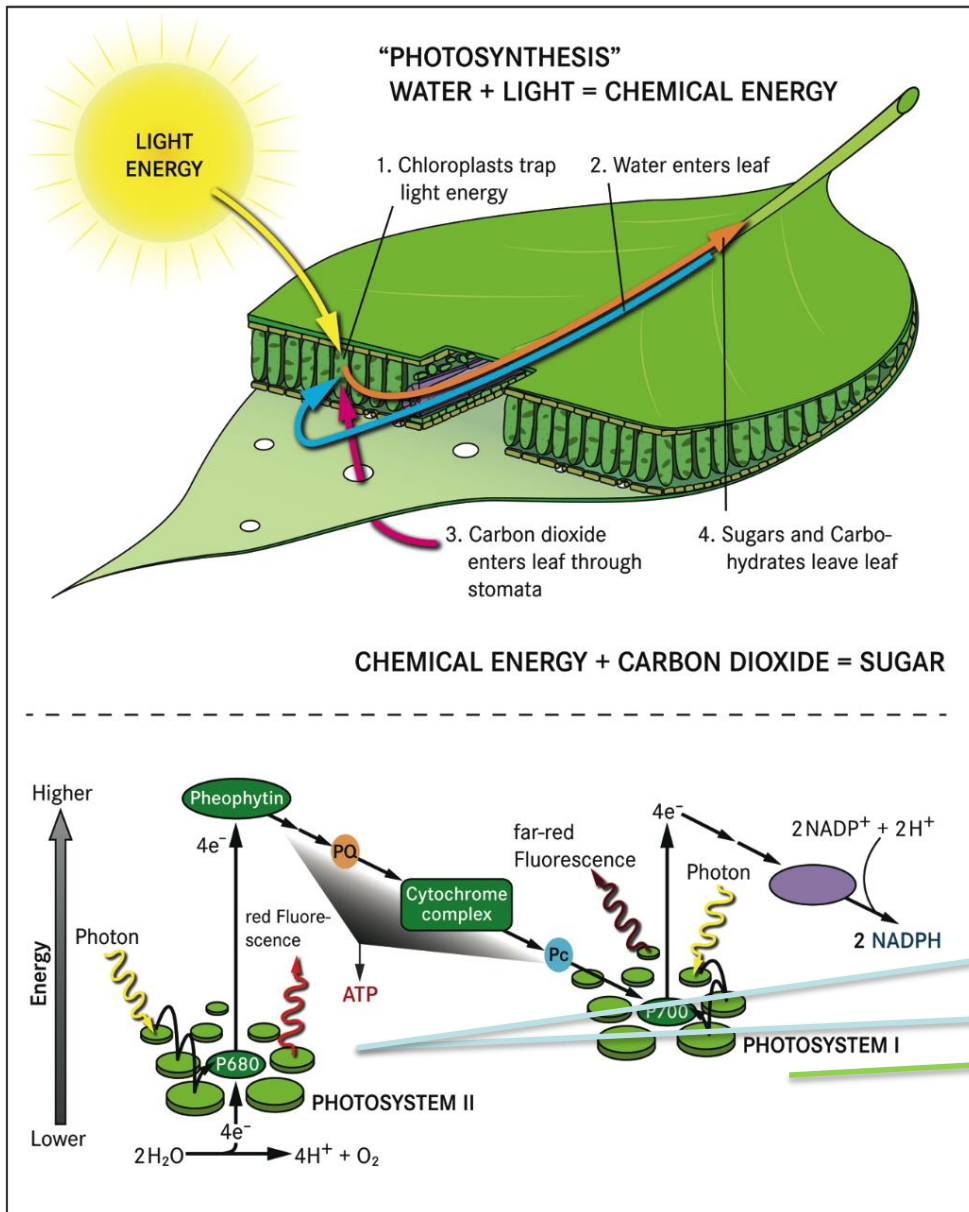
The origin of fluorescence – an indicator for photosynthetic efficiency

- Photosynthesis is a highly regulated process that involves a cascade of electron transfers (*Light reaction*) to fuel carbon fixation (*Calvin cycle*)
- Fluorescence is emitted from the cores of the photosynthetic machinery: Photosystems I and II

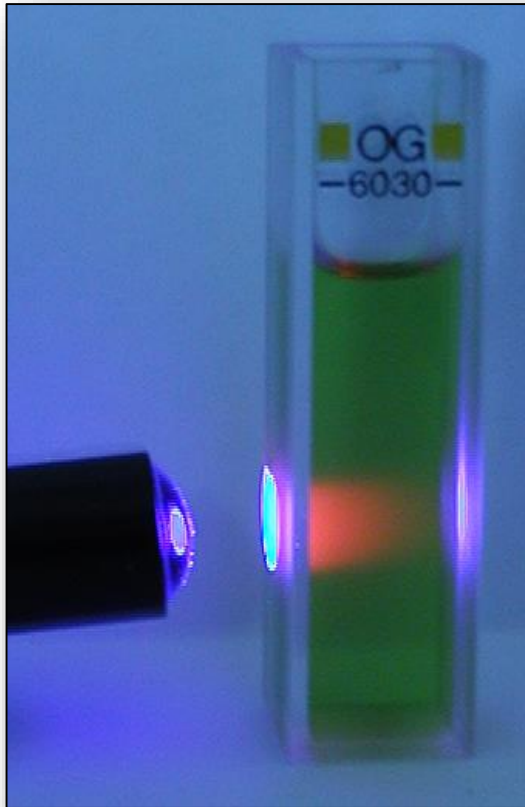


The origin of fluorescence – an indicator for photosynthetic efficiency

- Photosynthesis is a highly regulated process that involves a cascade of electron transfers (*Light reaction*) to fuel carbon fixation (*Calvin cycle*)
- Fluorescence is emitted from the cores of the photosynthetic machinery: Photosystems I and II
- Two-peak feature of fluorescence



Leaf fluorescence – two photosystems and two dynamically adapting signals



Biochimica et Biophysica Acta, 462 (1977) 307–313

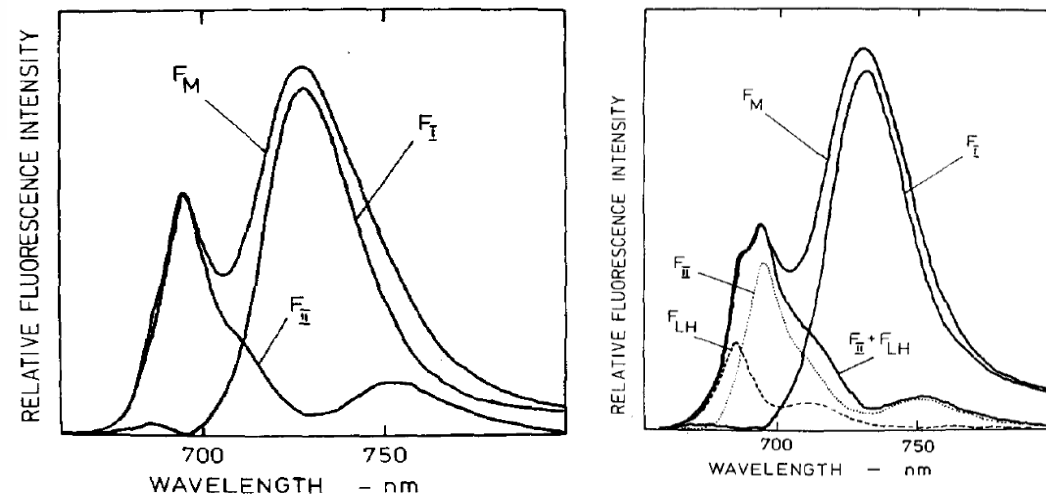
© Elsevier/North-Holland Biomedical Press

BBA 47380

FLUORESCENCE EMISSION SPECTRA OF PHOTOSYSTEM I, PHOTOSYSTEM II AND THE LIGHT-HARVESTING CHLOROPHYLL *a/b* COMPLEX OF HIGHER PLANTS

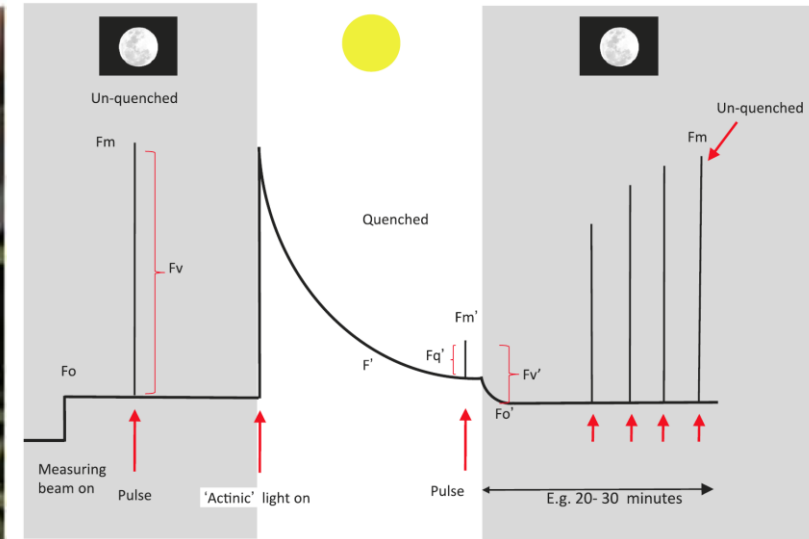
RETO J. STRASSER and WARREN L. BUTLER

Department of Biology, University of California, San Diego, La Jolla, Calif. 92093 (U.S.A.)



Fluorescence techniques are the most widely used approaches to investigate photosynthesis

- Various leaf level instruments available and currently ~750 Papers published per year
- Most methods use active approaches, such as PAM, saturating light pulses or lasers induced fluorescence transients



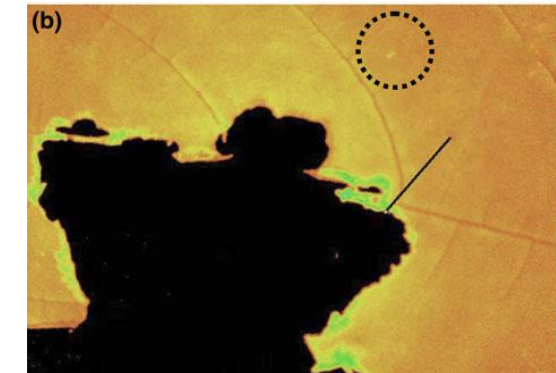
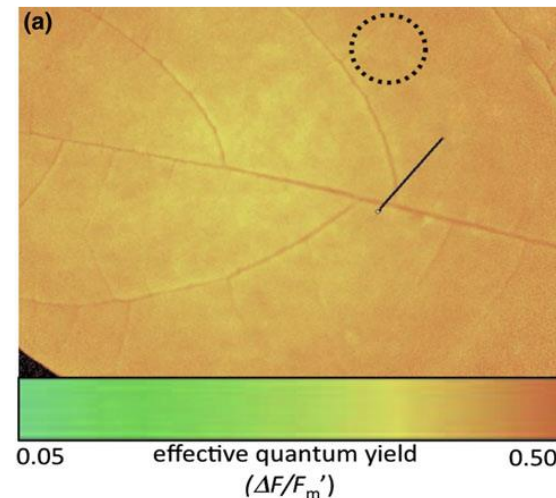
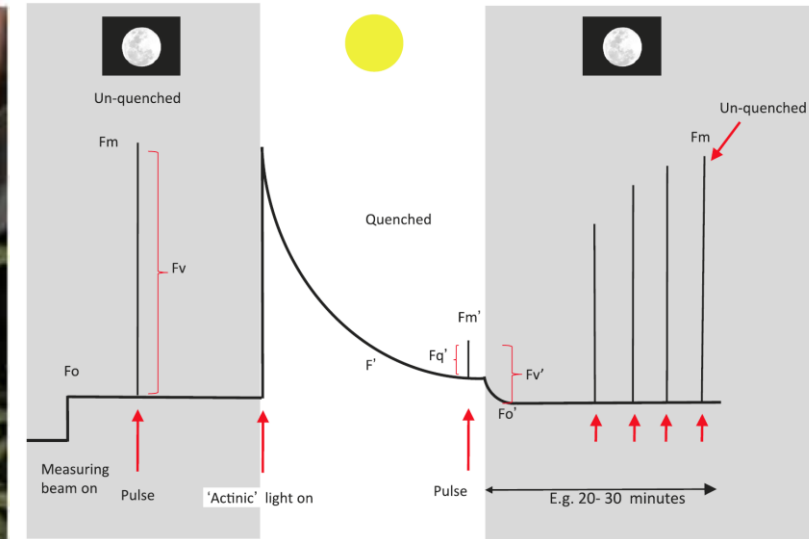
Rascher et al. (2010) Sensing of photosynthetic activity of crops. In Precision Crop Protection - the Challenge and Use of Heterogeneity. Springer Science+Business Media B.V., doi 10.1007/978-90-481-9277-9_6.

Murchie et al. (2018) Annals of Botany, 122, 207-220

Keller et al. (2019) Photosynthesis Research, in press; doi: 0.1007/s11120-018-0594-9.

Fluorescence techniques are the most widely used approaches to investigate photosynthesis

- Various leaf level instruments available and currently ~750 Papers published per year
- Most methods use active approaches, such as PAM, saturating light pulses or lasers induced fluorescence transients
- Fluorescence techniques are widely used to detect plant diseases
- Pest affect photosynthetic machinery and we can use this link for disease detection

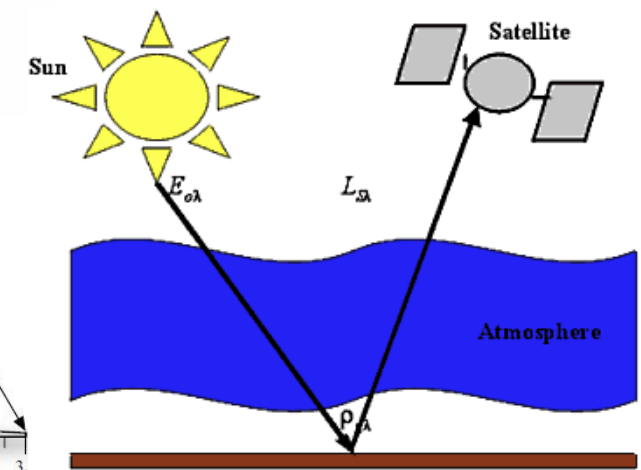
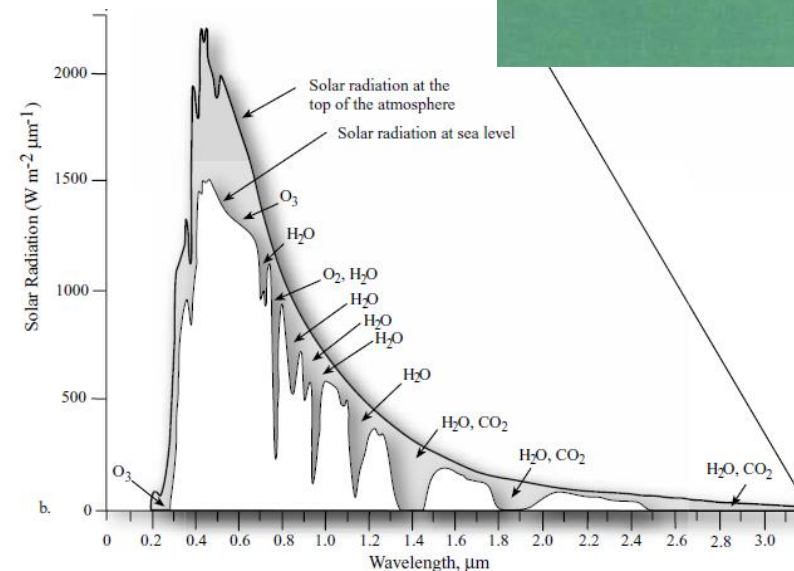
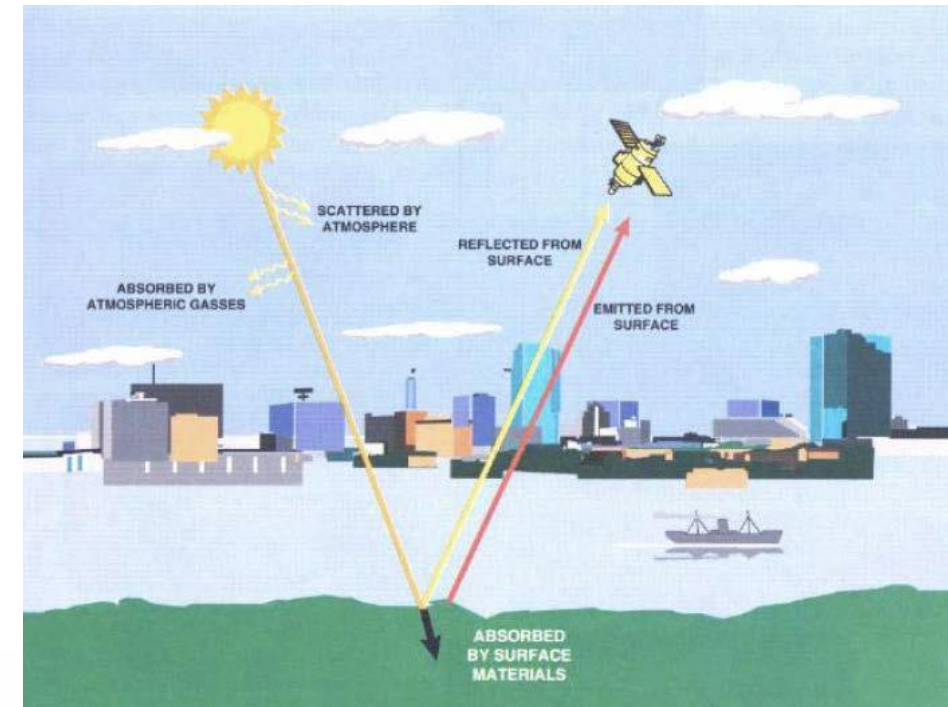


Barron-Gafford et al. (2012) Herbivory of wild *Manduca sexta* causes fast down-regulation of photosynthetic efficiency in *Datura wrightii*: an early signaling cascade visualized by chlorophyll fluorescence. *Photosyn Res*, 113, 249-260.

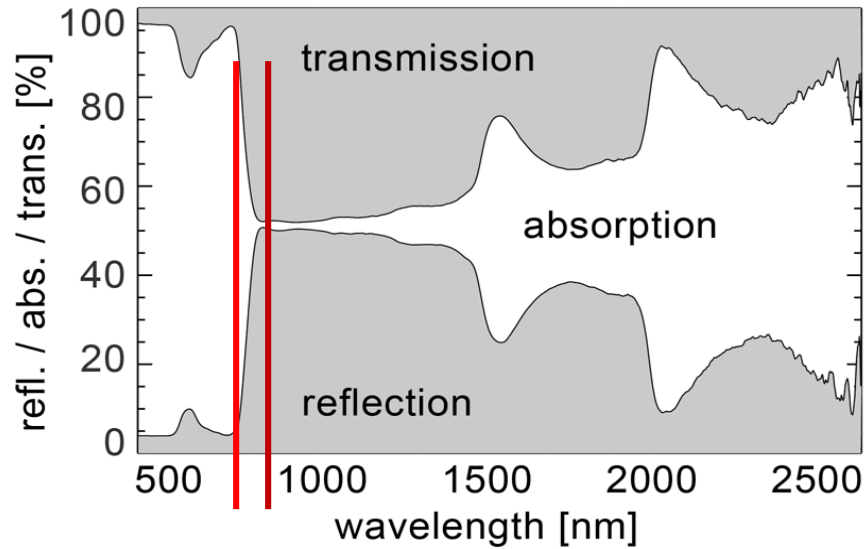
Passive remote sensing – optical remote sensing

- Passive remote sensing inherently needs to correct for the path of light through the atmosphere (sun – surface – sensor)
- Physical modelling (‘atmospheric correction’)
 - Complex atmospheric conditions (clouds, haze, aerosols, etc.) often complicate the atmospheric correction

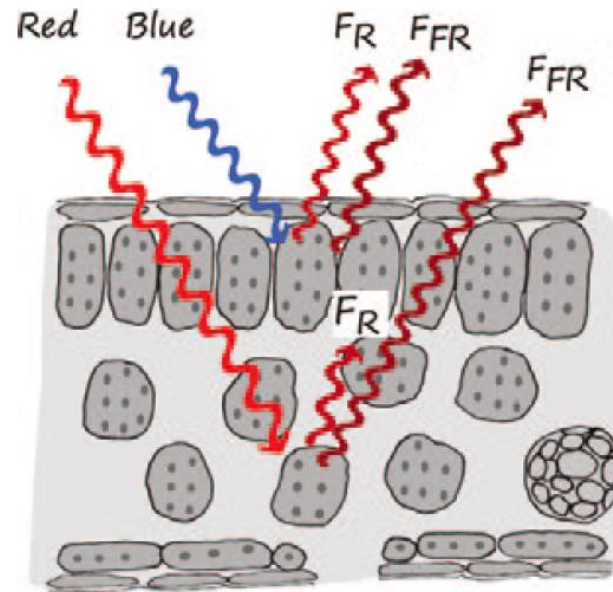
Atmospheric correction, correction for viewing geometry, georegistration and other corrections are normally done by the space agencies. Algorithms have become really good in past years, but they are still not error free



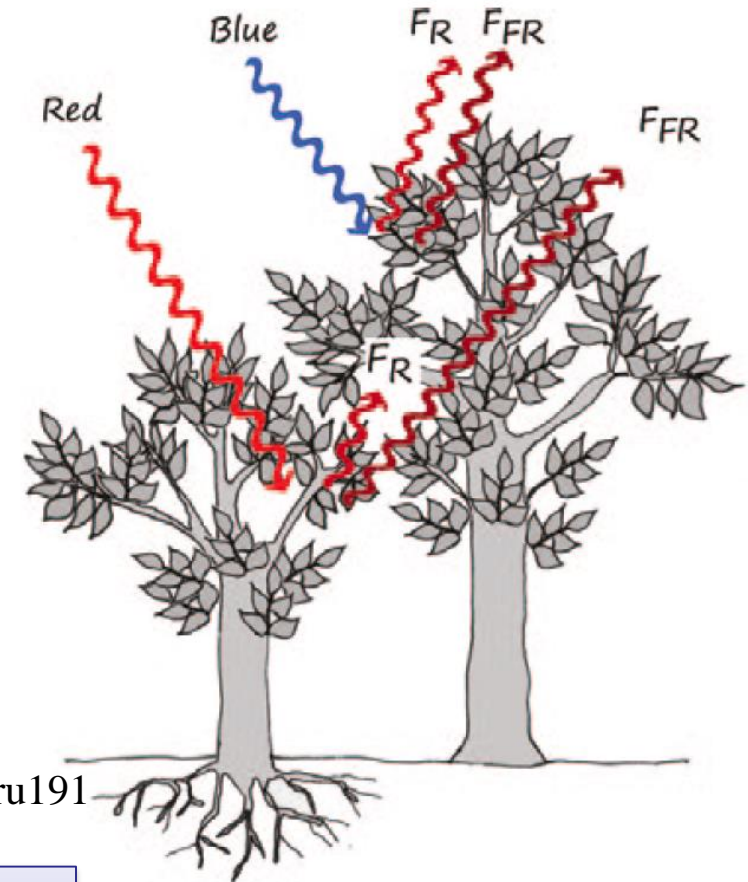
Challenge for disease detection: small features in structurally complex canopies



Rascher et al. (2010) *Precision Crop Protection*, Springer, ISBN: 978-90-481-9276-2, pp 87-100



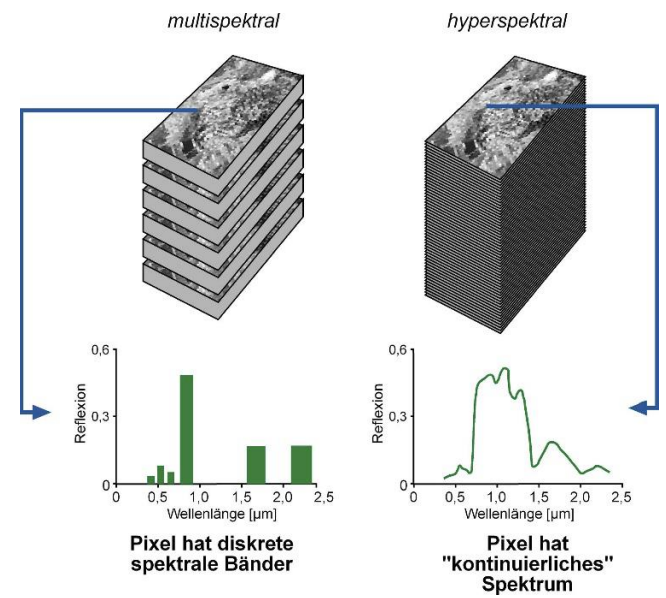
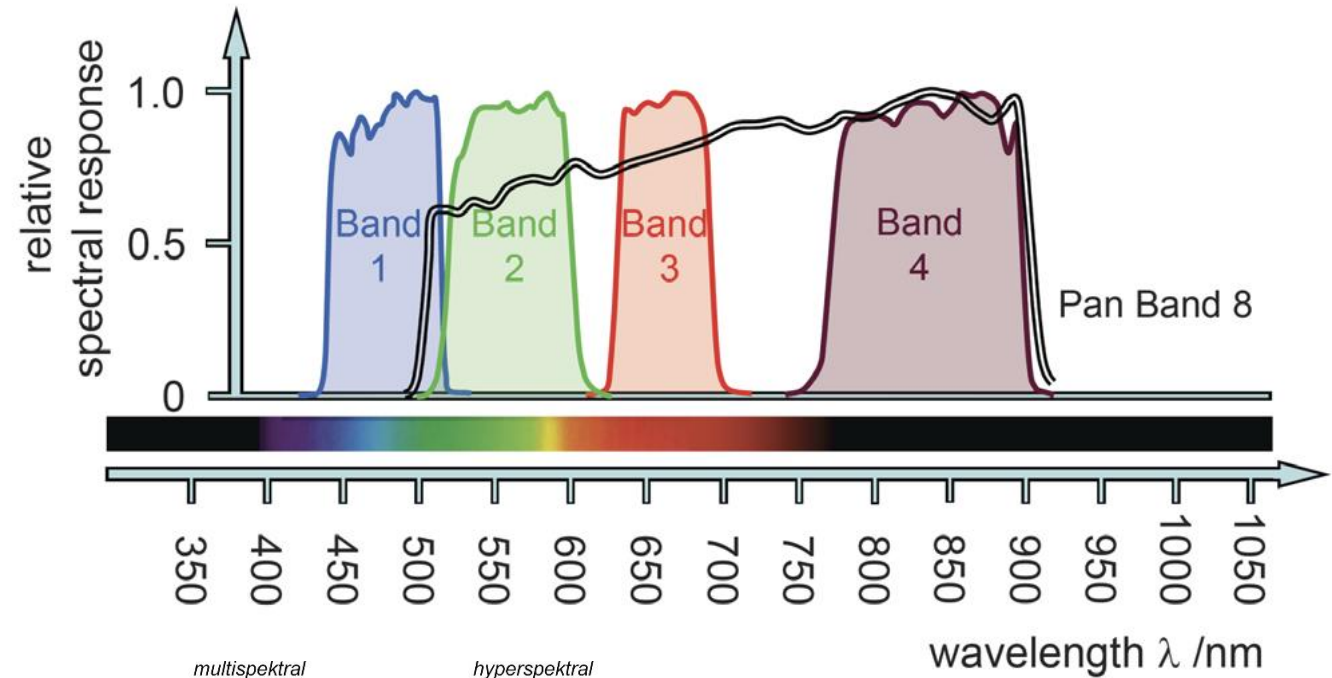
Porcar-Castell et al. (2014) *Journal of Experimental Botany*, doi:10.1093/jxb/eru191



- Disease symptoms are generally very small and only occur on selected organs of a plant.
- These small features, which are visible to the human eye are often obscured by changes in the canopy. Such changes often affect reflectance and fluorescence in a larger degree than disease effects

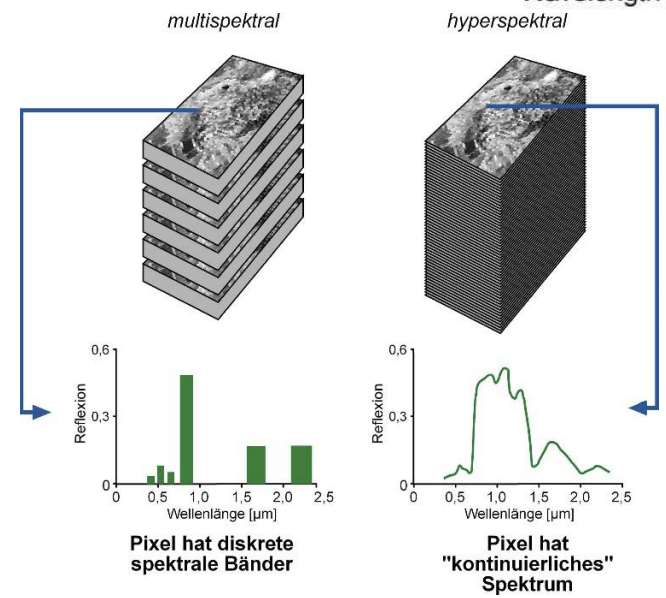
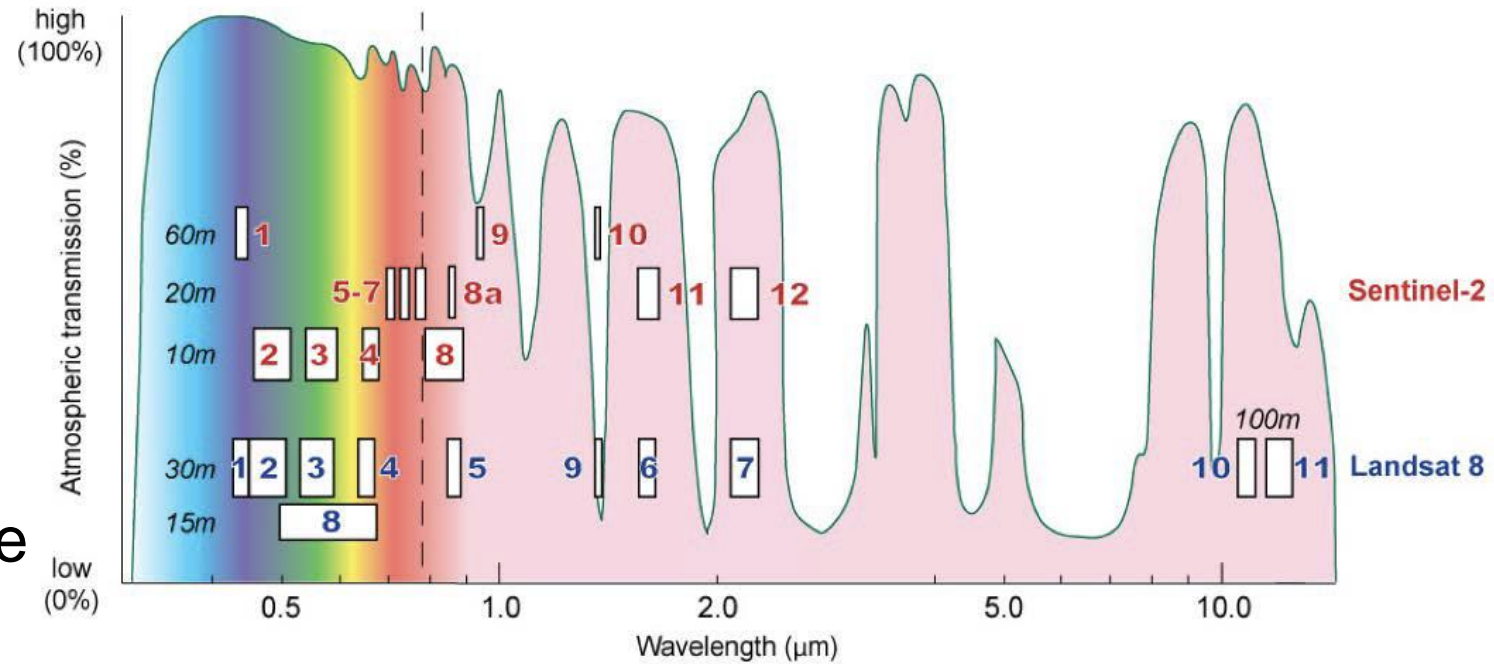
Multispectral vs hyperspectral satellites

- Every satellite or airborne sensor has a specific sensitivity to spectral bands
- Multispectral satellites measure discrete bands (advantage higher sensitivity as more photons are available in these broad bands)
- Hyperspectral satellites measure a continuous spectrum (advantage to resolve also small spectral differences, but lower signal to noise values)



Multispectral vs hyperspectral satellites

- Every satellite or airborne sensor has a specific sensitivity to spectral bands
- Multispectral satellites measure discrete bands (advantage higher sensitivity as more photons are available in these broad bands)
- Hyperspectral satellites measure a continuous spectrum (advantage to resolve also small spectral differences, but lower signal to noise values)



Existing and future satellite for crop monitoring and precision agriculture

Satellite (year)	Spatial resolution	Spectral resolution	Return frequency	Suitability for PA
Landsat 1/2/3 (1972/75/78)	79 m	4 bands	18 d	L
Landsat 5 (1984)	30 m	6 bands	16 d	M
SPOT 1/2/3 (1986/90/93)	20 m	3 bands	2-6 d	M
IRS 1A (1988)	72 m	4 bands	22 d	M
SPOT 4 (1998)	20 m	4 bands	2-6 d	M
IKONOS (1999)	3.2 m	5 bands	3 d	H
Landsat 7(1999)	30 m	7 bands	16 d	M
EO-1 Hyperion (2000)	30 m	220 bands	16 d	H
EO-1 ALI (2000)	30 m	10 bands	16 d	M
QuickBird (2001)	2.62 m	5 bands	1-4 d	H
PROBA/CHRIS (2001)	18-36 m	19 bands	7 d	H
SPOT 5 (2002)	10-20 m	5 bands	1-4 d	M
RapidEye (2008)	6.5 m	5 bands	1 d	H
GeoEye-1 (2008)	1.6 m	6 bands	2-8 d	H
WorldView-2 (2009)	1.84 m	8 bands	1.1 d	H
Pleiadis-1A/B (2011/12)	2 m	5 bands	1 d	H
SPOT 6/7 (2012/14)	6 m	5 bands	1 d	H

Satellite (year)	Spatial resolution	Spectral resolution	Return frequency	Suitability for PA
Landsat 8 (2013)	30 m	9 bands	16 d	M
SkySat-1/2 (2013/14)	2 m	5 bands	2 d	H
WorldView-3 (2014)	1.24-3.7 m	17 bands	1 d	H
Flock-1 1-28 (2014)	3-5 m	3 bands	1 d	M
Sentinel-2A (2015)	10-60 m	13 bands	10 d	M
WorldView-4 (2016)	1.24 m	5 bands	1 d	H
SkySat-3/4/5/6/7 (2016)	2 m	5 bands	1 d	M
Sentinel-2B (2017)	10-60 m	13 bands	10 d	M
Venµs (2017)	5.3 m	12 bands	2 d	H
PRISMA (2018)	20-30 m	249 bands	16 d	H
EnMAP (2019)	30 m	242 bands	27 d	H
HISUI MS (2019)	5 m	4 bands	60 d	M
HISUI HS (2019)	30 m	185 bands	60 d	H
Landsat 9 (2020)	30 m	9 bands	16 d	M
HyspIRI (2022)	30 m	214 bands	16 d	H
FLEX (2022)	300 m	-	30 d	M
Sentiel-2C/D (???)	10-60 m	13 bands	10 d	M

existing / past

future

Existing and future satellite for crop monitoring and precision agriculture

Satellite (year)	Spatial resolution	Spectral resolution	Return frequency	Suitability for PA
Landsat 1/2/3 (1972/75/78)	79 m	4 bands	18 d	L
Landsat 5 (1984)	30 m	6 bands	16 d	M
SPOT 1/2/3 (1986/90/93)	20 m	3 bands	2-6 d	M
IRS 1A (1988)	72 m	4 bands	22 d	M
SPOT 4 (1998)	20 m	4 bands	2-6 d	M
IKONOS (1999)	3.2 m	5 bands	3 d	H
Landsat 7(1999)	30 m	7 bands	16 d	M
EO-1 Hyperion (2000)	30 m	220 bands	16 d	H
EO-1 ALI (2000)	30 m	10 bands	16 d	M
QuickBird (2001)	2.62 m	5 bands	1-4 d	H
PROBA/CHRIS (2001)	18-36 m	19 bands	7 d	H
SPOT 5 (2002)	10-20 m	5 bands	1-4 d	M
RapidEye (2008)	6.5 m	5 bands	1 d	H
GeoEye-1 (2008)	1.6 m	6 bands	2-8 d	H
WorldView-2 (2009)	1.84 m	8 bands	1.1 d	H
Pleiadis-1A/B (2011/12)	2 m	5 bands	1 d	H
SPOT 6/7 (2012/14)	6 m	5 bands	1 d	H

Satellite (year)	Spatial resolution	Spectral resolution	Return frequency	Suitability for PA
Landsat 8 (2013)	30 m	9 bands	16 d	M
SkySat-1/2 (2013/14)	2 m	5 bands	2 d	H
WorldView-3 (2014)				
Flock-1 1-28 (2014)				
Sentinel-2A (2015)				
WorldView-4 (2016)				
SkySat-3/4/5/6/7 (2016)				
Sentinel-2B (2017)				
Venµs (2017)				
PRISMA (2018)	20-30 m	249 bands	16 d	H
EnMAP (2019)	30 m	242 bands	27 d	H
HISUI MS (2019)	5 m	4 bands	60 d	M
HISUI HS (2019)	30 m	185 bands	60 d	H
Landsat 9 (2020)	30 m	9 bands	16 d	M
HypSIrI (2022)	30 m	214 bands	16 d	H
FLEX (2022)	300 m	-	30 d	M
Sentiel-2C/D (???)	10-60 m	13 bands	10 d	M

Technical feasibility defines the possibility, scientific request define the balance between
 (i) spatial resolution
 (ii) number of spectral bands, and
 (iii) revisiting time (return frequency)

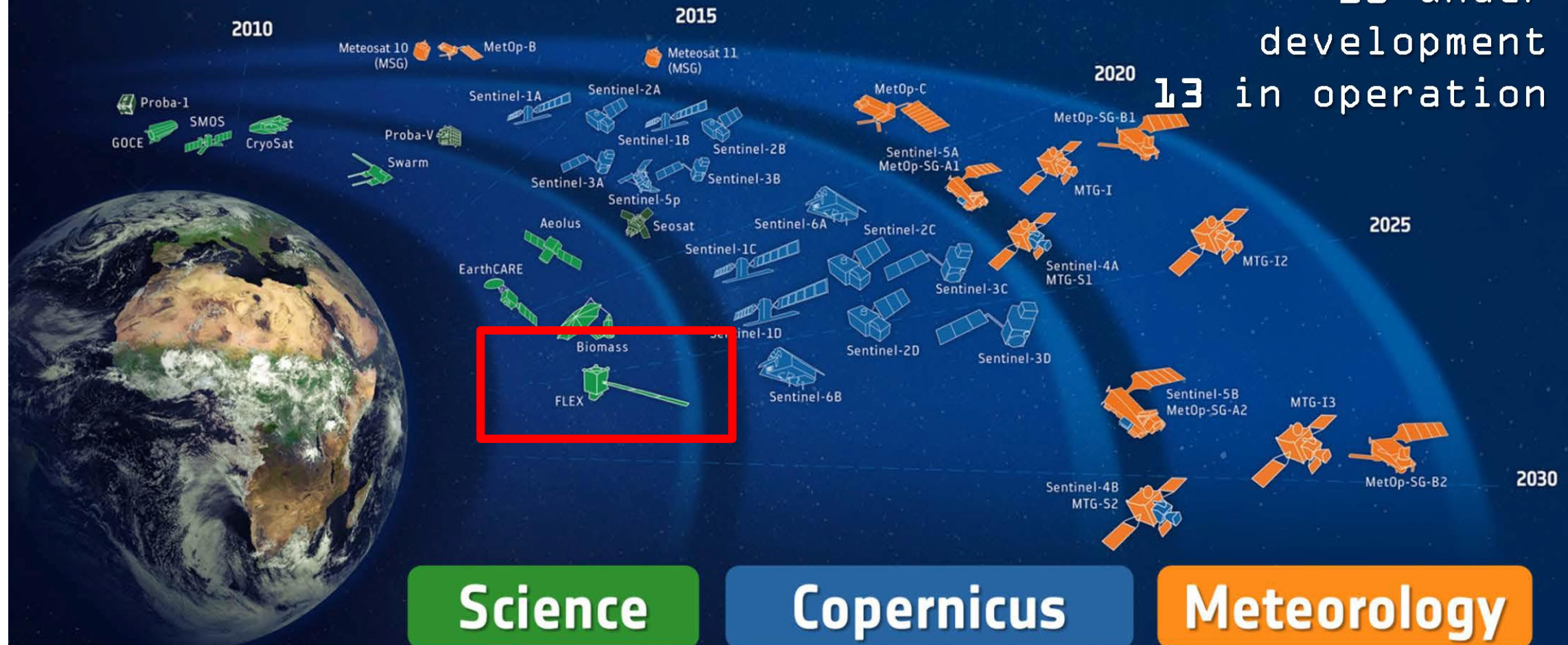
existing / past

future

ESA-DEVELOPED EARTH OBSERVATION MISSIONS



Satellites
28 under
development
13 in operation



FLEX Satellite Mission will become the 8th Earth Explorer of ESA

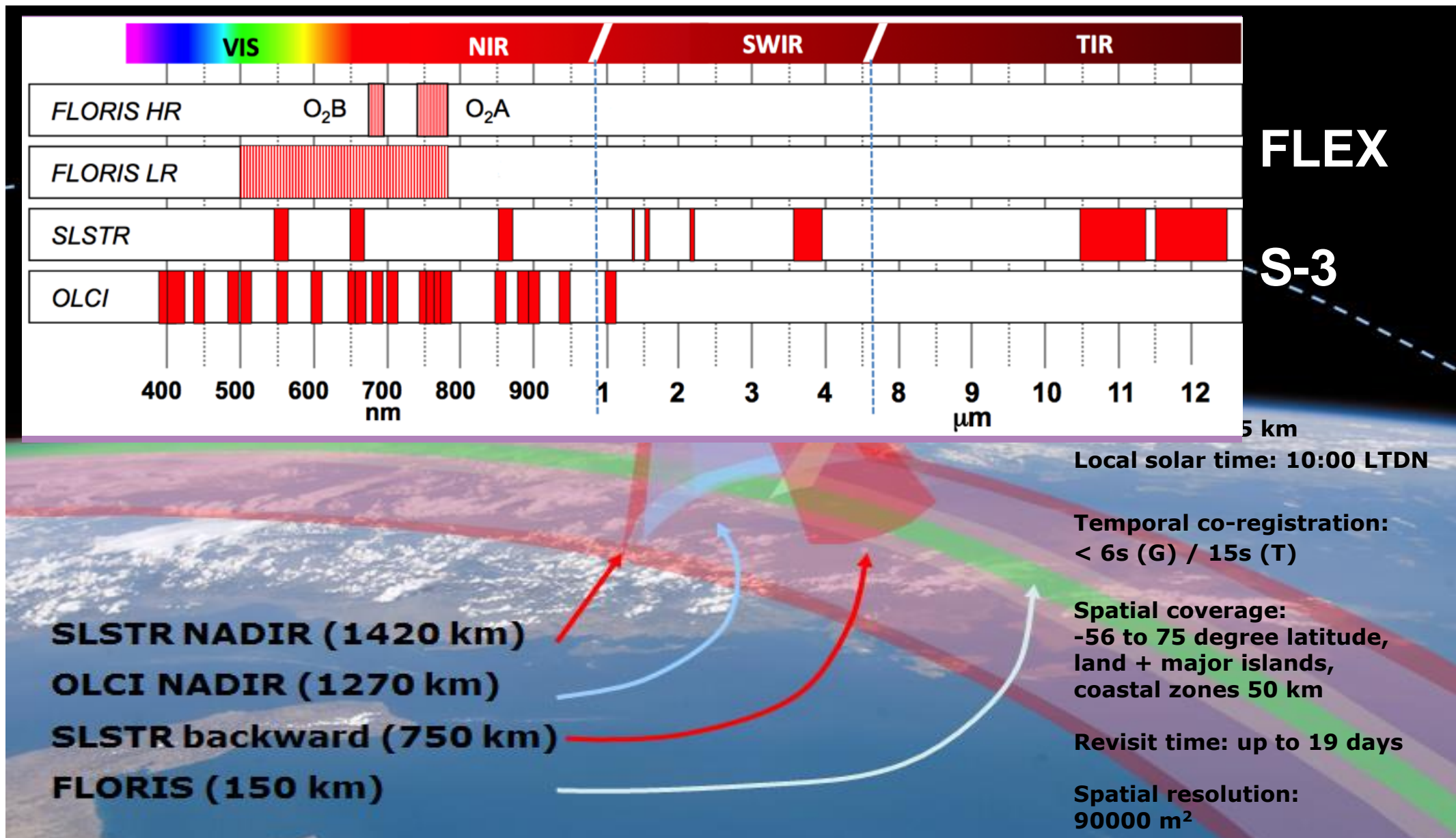
FLEX will quantify **actual photosynthetic activity** of terrestrial ecosystems

FLEX will provide **physiological indicators** for vegetation health status

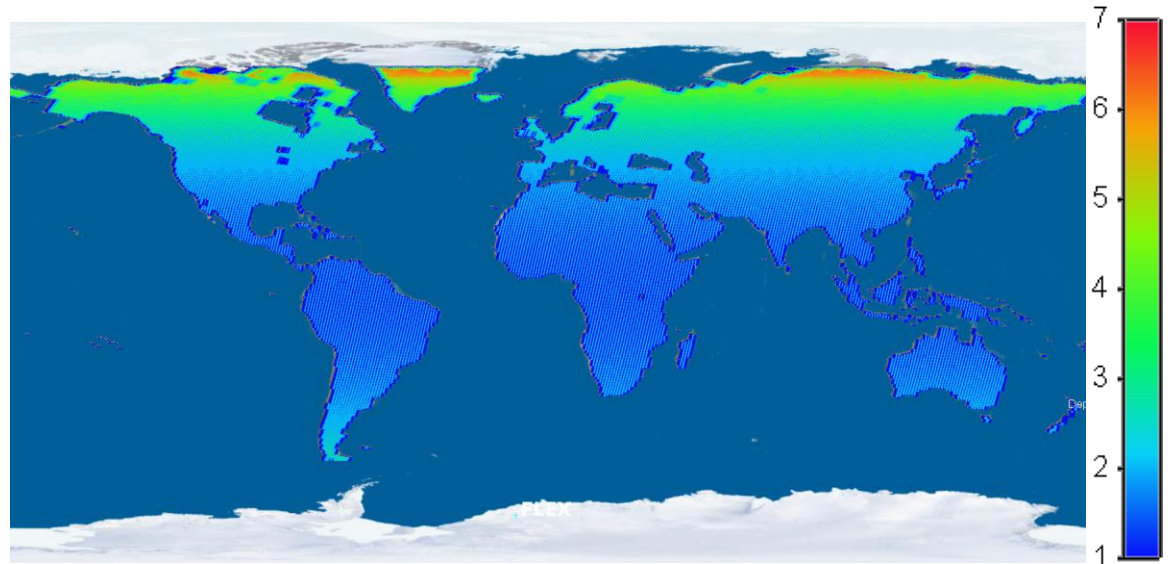
by direct measurements of **vegetation fluorescence** at 300x300 meters every 10-25 days



FLEX Satellite Mission will be launched mid 2025 and become the 8th Earth Explorer of ESA



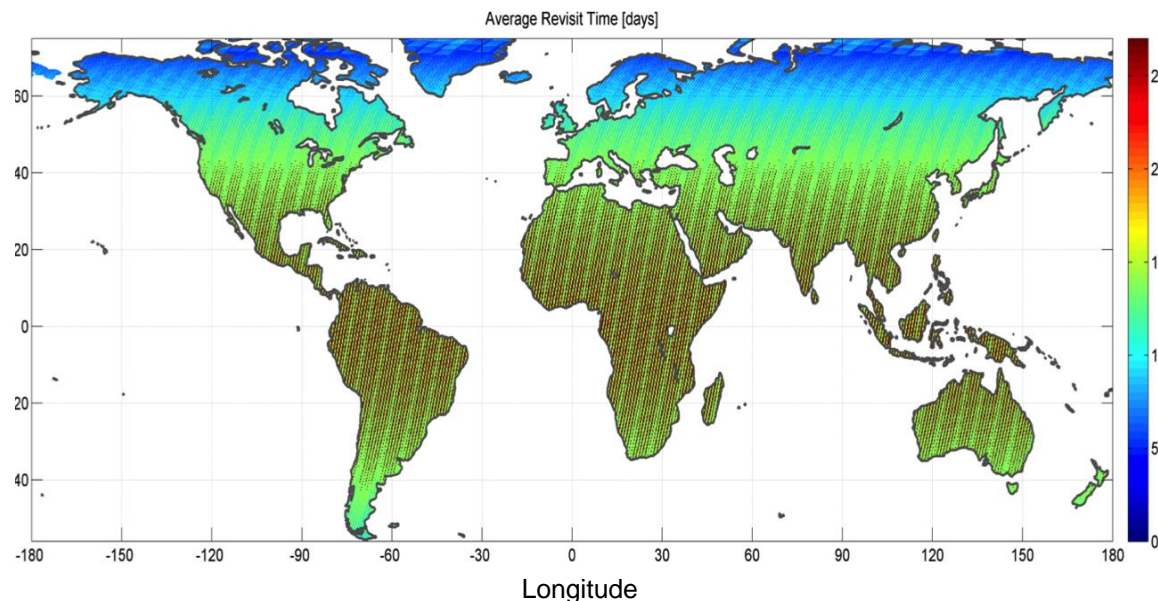
FLEX Satellite Mission – a tandem concept with Sentinel-3



Number of acquisitions within a repeat cycle

Average revisit time (days)

- FLEX will acquire images of all land between 56° S to 75° N, including major islands and coastal areas
- 300 x 300 meter pixels
- Launch is scheduled for 2023
- Full coverage every 27 days



Average revisit time:

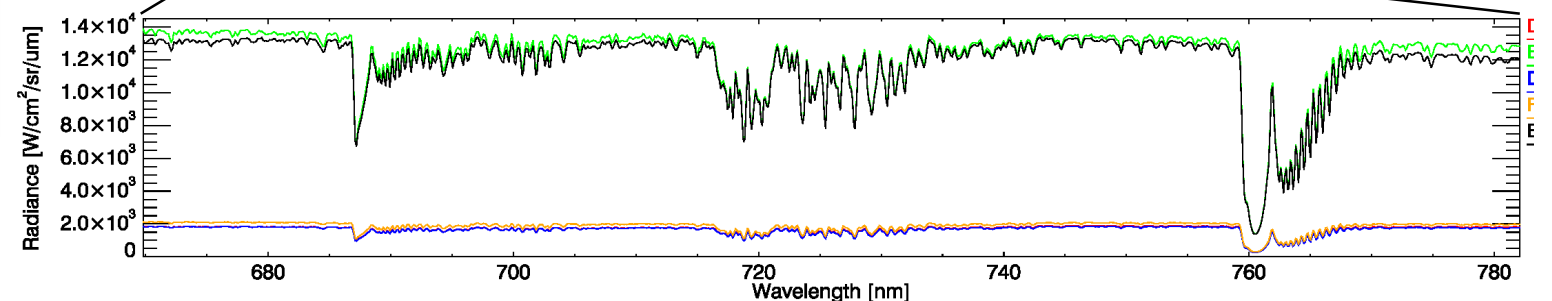
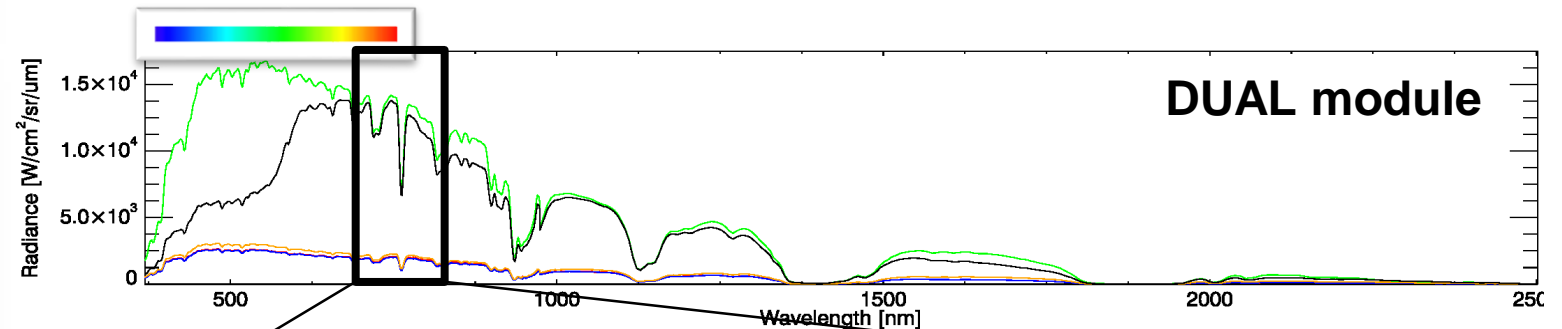
- 27 days at Equator
- ~ 20 days at Tropics
- ~ 10 to 15 days over Europe and Canada
- ~ 5 to 10 days over boreal forests

HyPlant: A high-resolution airborne imaging spectrometer with FLEX like measurement characteristics



Rascher et al. (2015) *Global Change Biology*, 21, 4673–4684

- **DUAL module** (380 – 2500 nm)
VIS/NIR: 3-4 nm FWHM, 1.7 nm SSI, SNR > 510
SWIR: 13 nm FWHM, 5.5 nm SSI, SNR > 1100
- **FLUO module** (670 – 780 nm)
0.25 nm FWHM, 0.11 nm SSI, SNR > 250
- Various improvement and now consolidated version (HyPlant_3)



HyPlant complemented by thermal imager (TASI) and LIDAR system (since 2018 campaign)

TASI-600

- Hyperspectral thermal sensor (8 – 11.5 μm)
- Field of view aligned with HyPlant sensor
- Operated in synchrony with HyPlant

LIDAR (Riegl LMS-Q780)

- Long range laser scanner
- Full-waveform echo digitalization and analysis

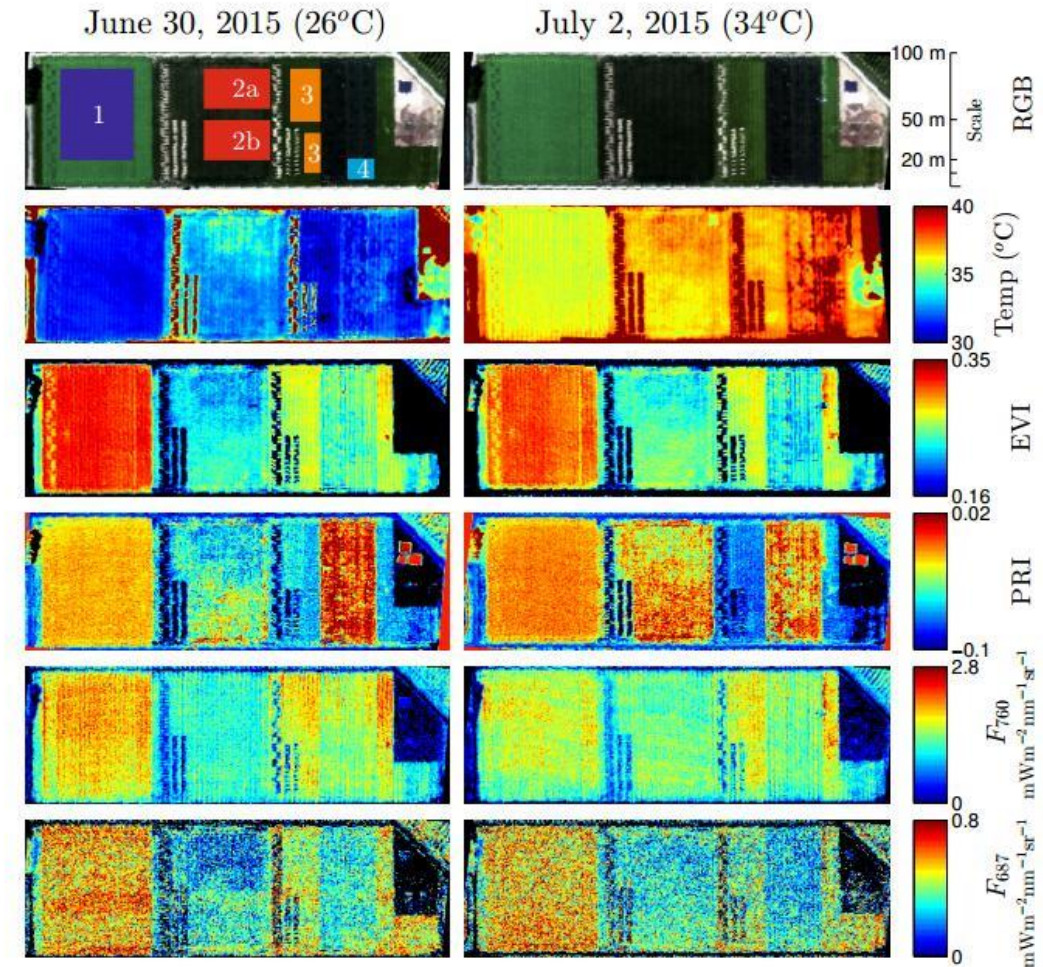
Sensor	Riegl LMS-Q780
Max. Pulse Repetition Rate [kHz]	400
Max. Operating Altitude [m]	5800
Wavelength [nm]	1064
Max. Laser Beam Divergence [mrad]	0.25
FOV [°]	60
Eye Safety Class	Laser Class 3B
Min. Operating Altitude	50m



Sensor	TASI-600
Spectral Region	LWIR
Sensor Type	Pushbroom Hyperspectral TIR
Spectral Bands	32
Spectral Range [nm]	8 000 – 11 500
Number of Spatial Pixels	600
Max. Spectral Resolution [nm]	110
FOV [°]	40
IFOV [°]	0.07
Dynamic Range	14-bits (16384:1)
Burst Data Rate	5 Mpix/sec
NEDT	TASI-600/32: 0.11° C @ 100° C
Spectral Smile	TASI-600/32: < ± 0.25 pixels
Keystone Distortion	TASI-600/32: < ± 0.25 pixels

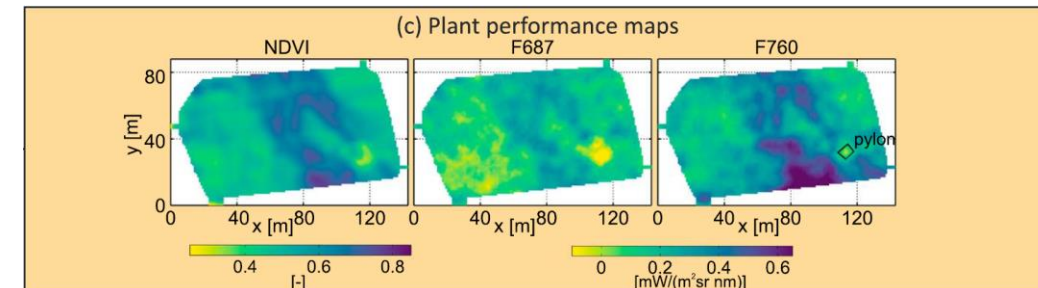
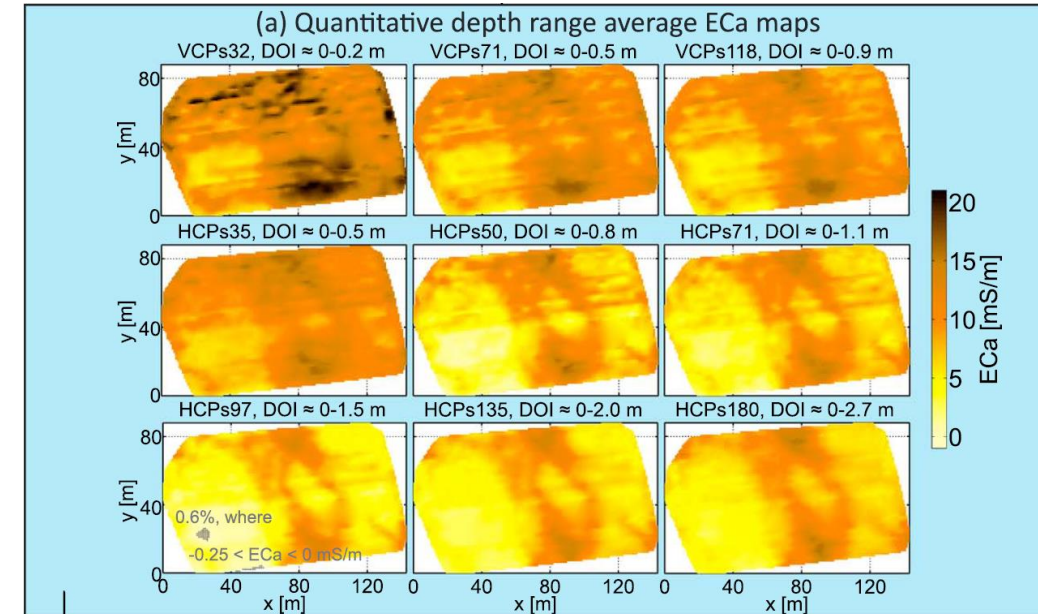
HyPlant campaigns: Measuring and understanding the spatial dynamics of solar-induced fluorescence

- Vegetation stress during summer heat wave
[Yang et al (2019) Rem. Sens. Environ., doi:
10.1016/j.rse.2018.11.039]



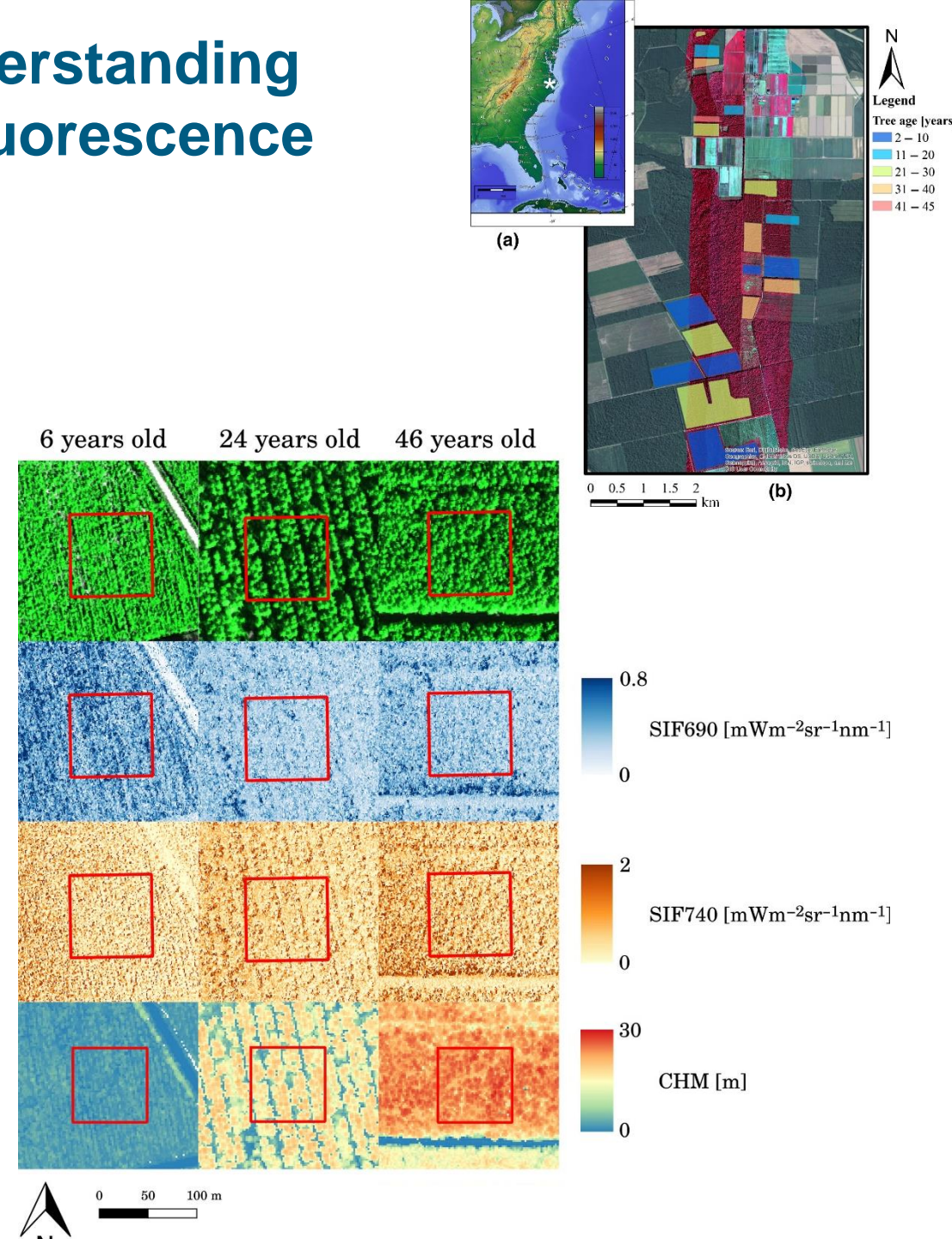
HyPlant campaigns: Measuring and understanding the spatial dynamics of solar-induced fluorescence

- Vegetation stress during summer heat wave
[Yang et al (2019) Rem. Sens. Environ., doi:
10.1016/j.rse.2018.11.039]
- Water availability of deeper soil layers are mapped in fluorescence signal
[von Hebel et al (2018) Geophys. Res Lett., 45,]



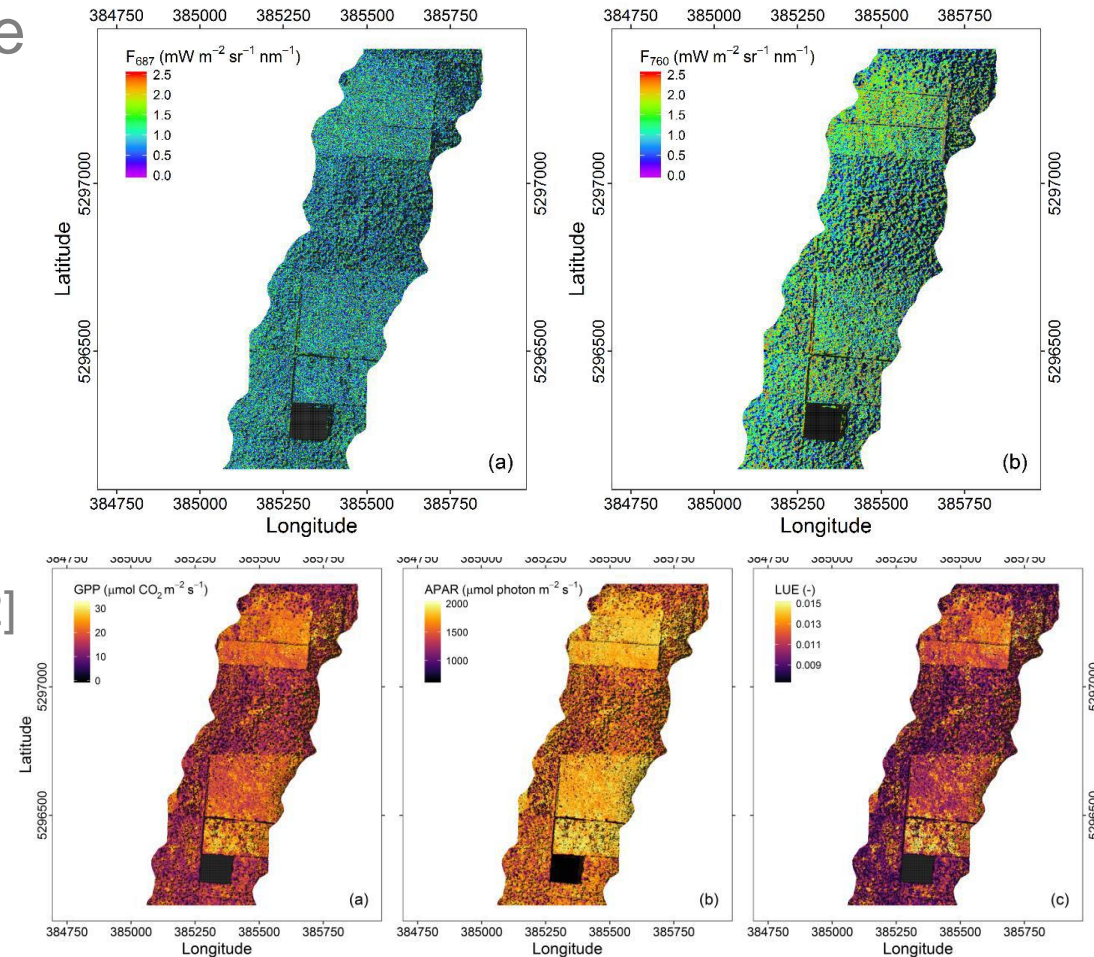
HyPlant campaigns: Measuring and understanding the spatial dynamics of solar-induced fluorescence

- Vegetation stress during summer heat wave
[Yang et al (2019) Rem. Sens. Environ., doi: 10.1016/j.rse.2018.11.039]
- Water availability of deeper soil layers are mapped in fluorescence signal
[von Hebel et al (2018) Geophys. Res Lett., 45,]
- F_{690} reflects tree age, while F_{740} is constant in Loblolly pine stands of different age
[Middleton et al (2017) Remote Sensing, 9, article no. 612]
[Colombo et al. (2018) Global Change Biology, 24, 2980-2996]



HyPlant campaigns: Measuring and understanding the spatial dynamics of solar-induced fluorescence

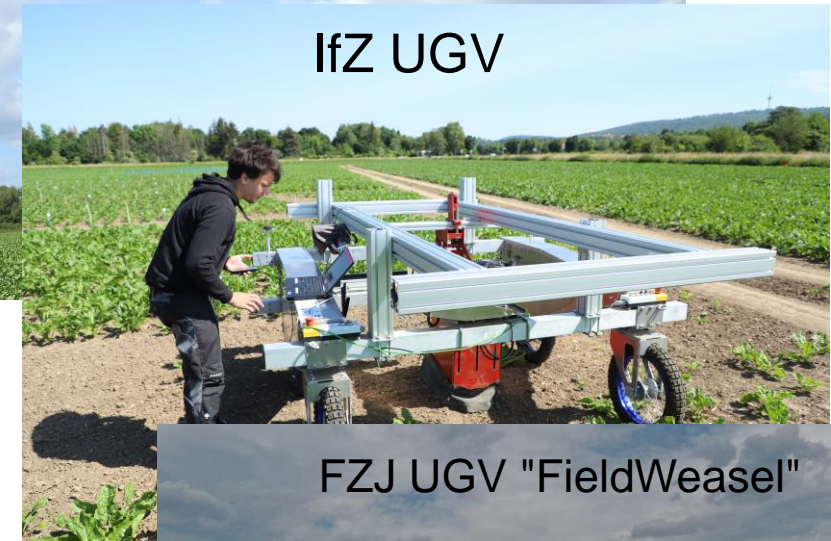
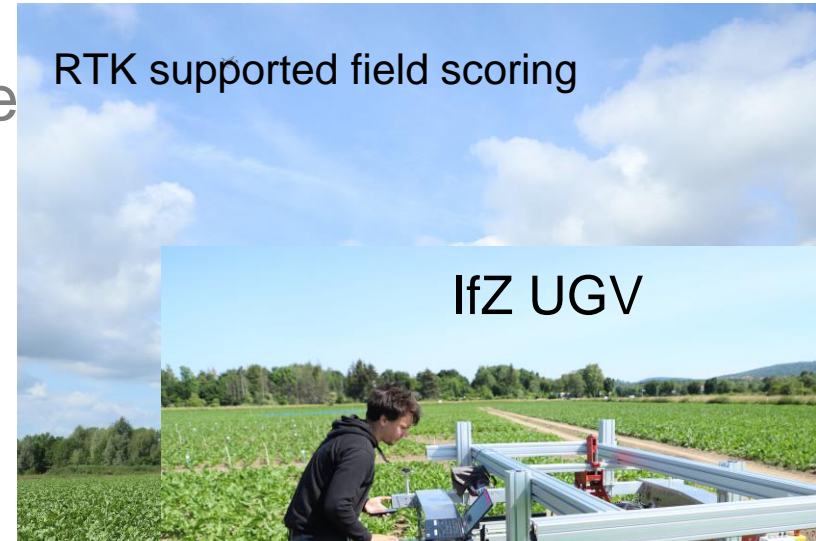
- Vegetation stress during summer heat wave
[Yang et al (2019) Rem. Sens. Environ., doi: 10.1016/j.rse.2018.11.039]
- Water availability of deeper soil layers are mapped in fluorescence signal
[von Hebel et al (2018) Geophys. Res Lett., 45,]
- F_{690} reflects tree age, while F_{740} is constant in Loblolly pine stands of different age
[Middleton et al (2017) Remote Sensing, 9, article no. 612]
[Colombo et al. (2018) Global Change Biology, 24, 2980-2996]
- F_{687} is not related with GPP and APAR; F_{760} is positively, but nonlinearly related to GPP and APAR in the spatial domain.
[Tagliabue et al. (2019) Rem. Sens. Environ., 231, article no. 111272]



HyPlant campaigns: Measuring and understanding the spatial dynamics of solar-induced fluorescence

- Vegetation stress during summer heat wave
[Yang et al (2019) Rem. Sens. Environ., doi: 10.1016/j.rse.2018.11.039]
- Water availability of deeper soil layers are mapped in fluorescence signal
[von Hebel et al (2018) Geophys. Res Lett., 45,]
- F_{690} reflects tree age, while F_{740} is constant in Loblolly pine stands of different age
[Middleton et al (2017) Remote Sensing, 9, article no. 612]
[Colombo et al. (2018) Global Change Biology, 24, 2980-2996]


- No dedicated disease mapping / detection / identification campaigns or projects, yet.
- We are just preparing for first data acquisition over a disease field trail in Göttingen (cooperation with A. Mahlein)



On the potential to early and specifically map diseases from ground-, air-, and spaceborne platforms

- Proximity sensing: Advanced technologies (sensors and data processing techniques) nowadays open the possibility for reliable disease detection
- But we
 - need to get beyond vegetation indices, i.e. use the full information, which we have available from novel sensors
 - Be clear what are direct signals emerging from the disease and what are correlated effects that obscure the disease signal

Available online at www.sciencedirect.com




ELSEVIER

ScienceDirect

Current Opinion in
Plant Biology

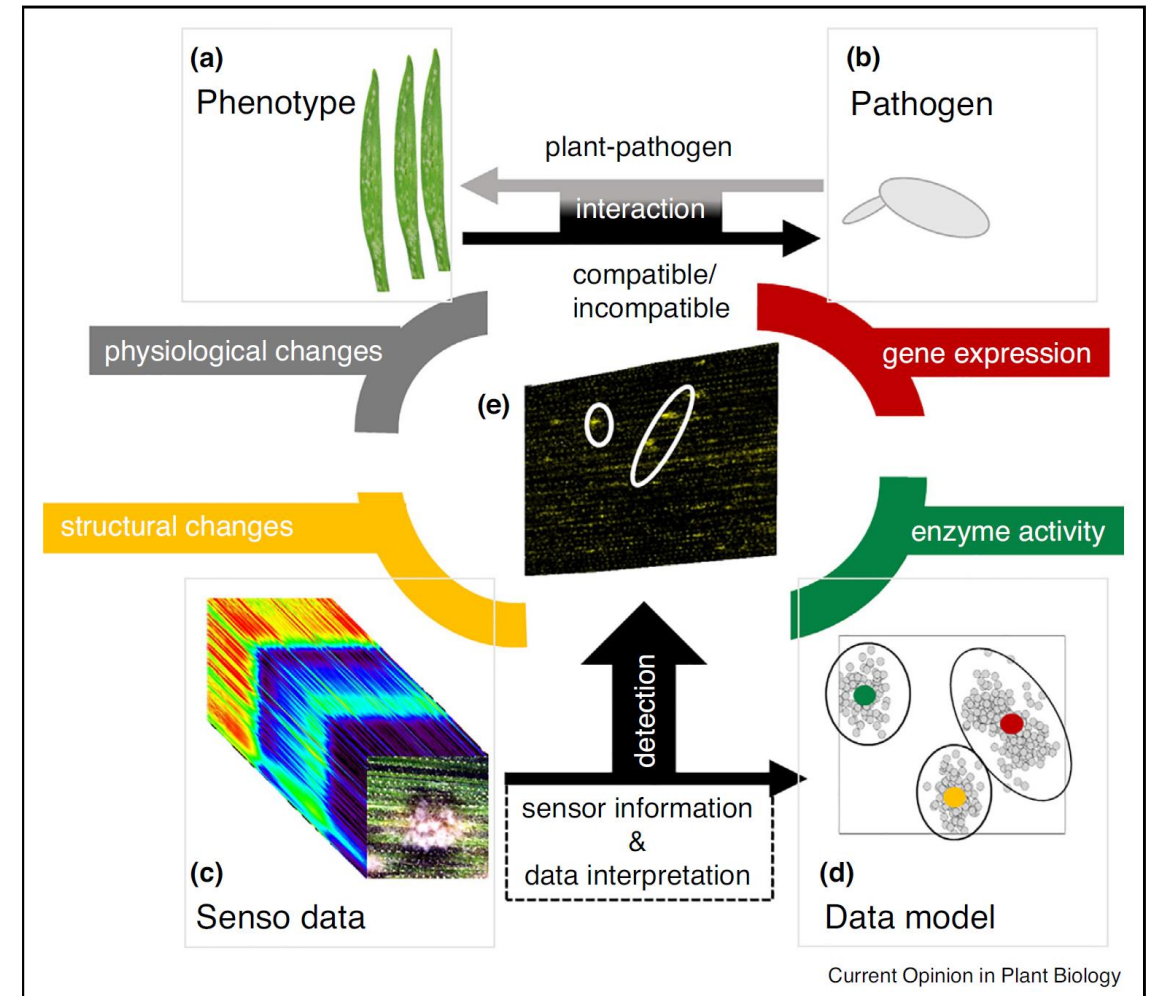
Quantitative and qualitative phenotyping of disease resistance of crops by hyperspectral sensors: seamless interlocking of phytopathology, sensors, and machine learning is needed!

Anne-Katrin Mahlein^{1,2}, Matheus Thomas Kuska², Stefan Thomas², Mirwaes Wahabzada², Jan Behmann², Uwe Rascher³ and Kristian Kersting⁴



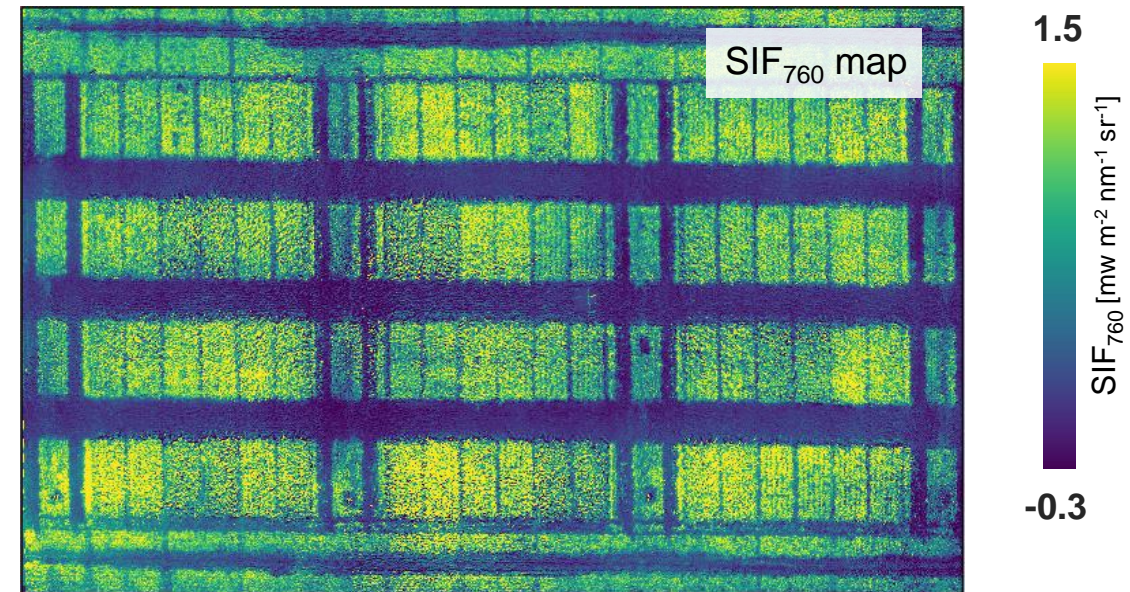
On the potential to early and specifically map diseases from ground-, air-, and spaceborne platforms

- ❑ Proximity sensing: Advanced technologies (sensors and data processing techniques) nowadays open the possibility for reliable disease detection
- ❑ But we
 - ❑ need to get beyond vegetation indices, i.e. use the full information, which we have available from novel sensors
 - ❑ Be clear what are direct signals emerging from the disease and what are correlated effects that obscure the disease signal



On the potential to early and specifically map diseases from ground-, air-, and spaceborne platforms

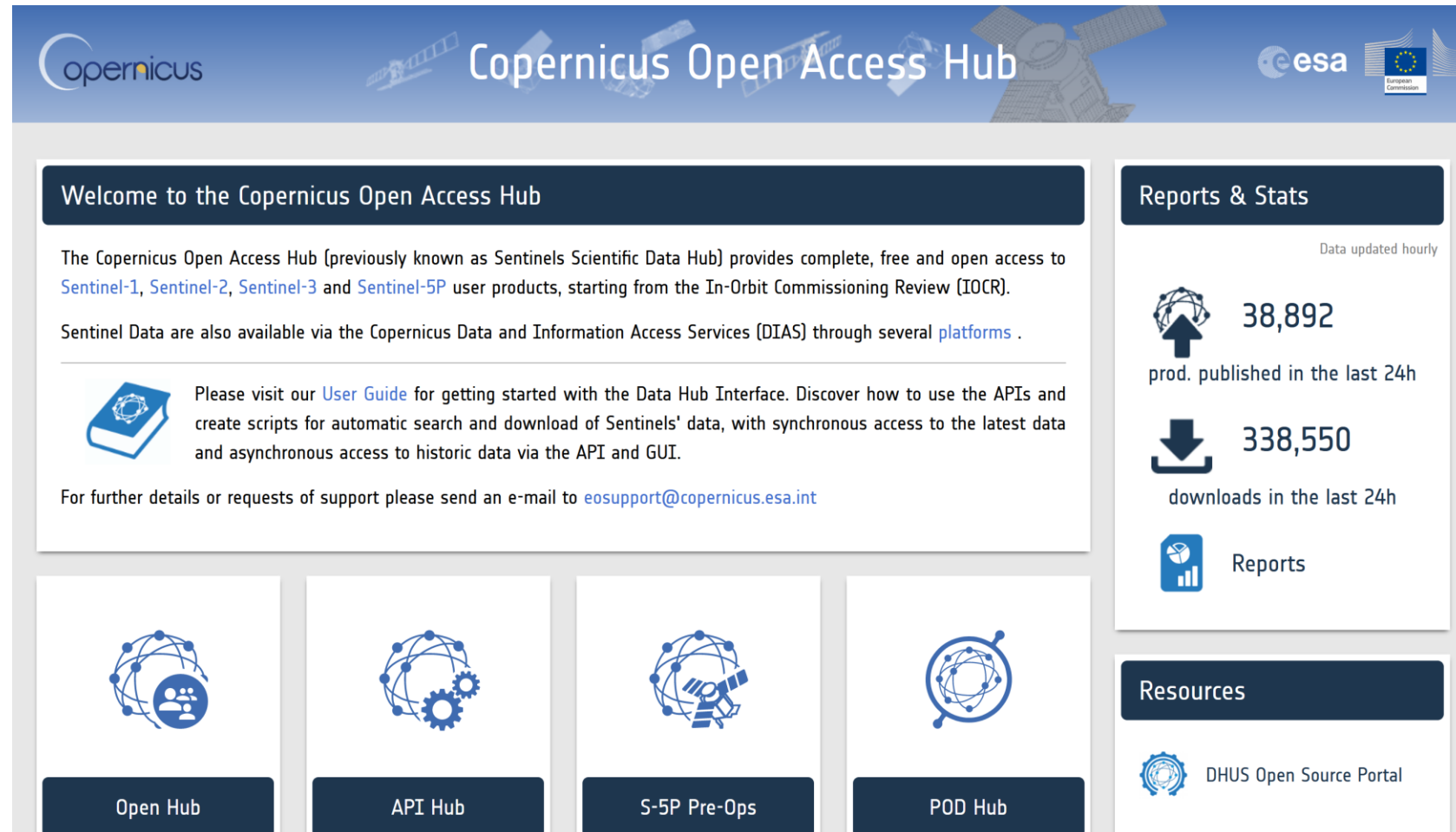
- ❑ Airborne sensing: UAV-based are booming, while airborne sensors are on the decline.
- ❑ Same considerations as before, additionally
 - ❑ Data quality of UAV based sensors is often limiting
 - ❑ Often naïve use of e.g. machine learning
 - ❑ Airborne sensors are important to close the gap to satellites
 - ❑ New measurement options which potential for disease detection may still be explored



Kneer et al (2023) A snapshot imaging system for the measurement of solar-induced chlorophyll fluorescence – addressing the challenges of high-performance spectral imaging for mapping SIF. *IEEE – Sensors, online first*; 10.1109/JSEN.2023.3297054

On the potential to early and specifically map diseases from ground-, air-, and spaceborne platforms

- ❑ Satellite sensing: Satellite data are nowadays of high quality and easy to use. But still a trade-off between spectral performance spatial resolution and revisiting time

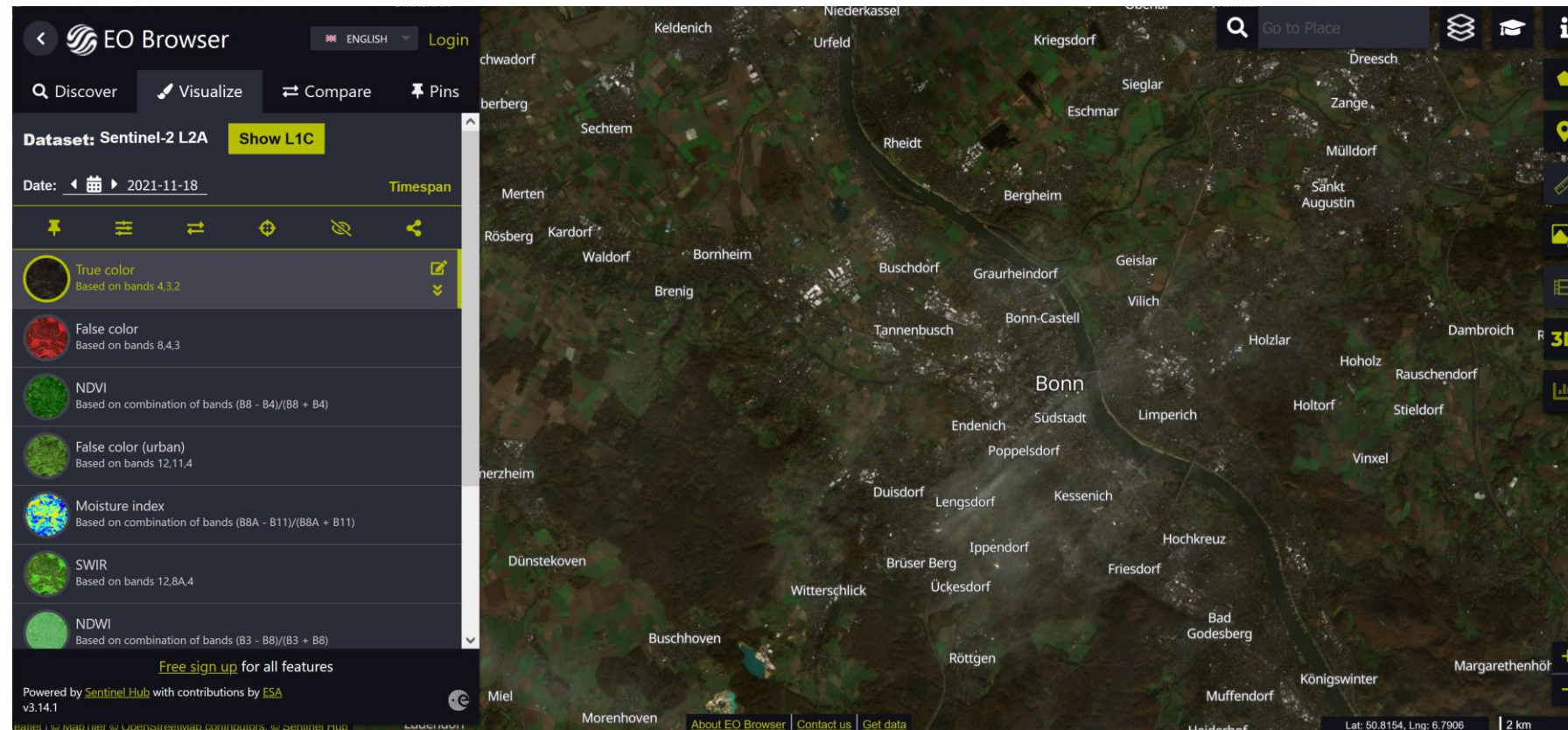


The screenshot shows the Copernicus Open Access Hub website. The header includes the Copernicus logo, the text 'Copernicus Open Access Hub', and the ESA and European Commission logos. The main content area features a welcome message, a description of the hub's services, and a link to the user guide. A sidebar on the right displays 'Reports & Stats' with a data update frequency of 'Data updated hourly', showing 38,892 products published and 338,550 downloads in the last 24 hours. Below the main content are four navigation buttons: 'Open Hub', 'API Hub', 'S-5P Pre-Ops', and 'POD Hub'. The footer of the sidebar includes a 'Resources' section with a link to the 'DHUS Open Source Portal'.

<https://scihub.copernicus.eu/>

On the potential to early and specifically map diseases from ground-, air-, and spaceborne platforms

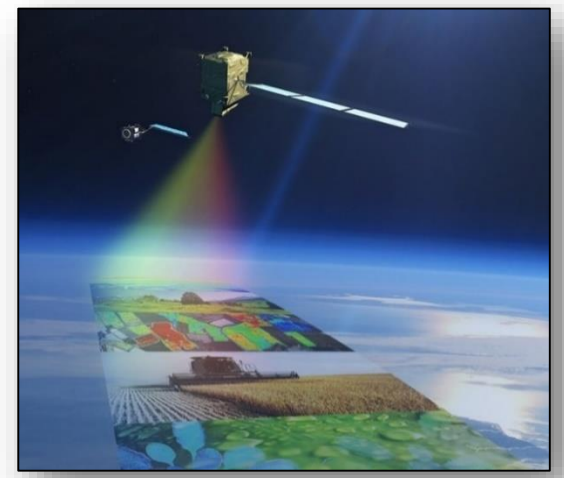
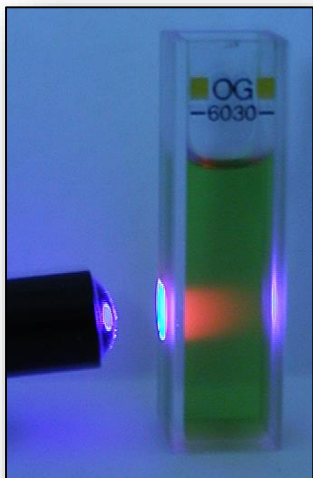
- ❑ Satellite sensing: Satellite data are nowadays of high quality and easy to use. But still a trade-off between spectral performance spatial resolution and revisiting time



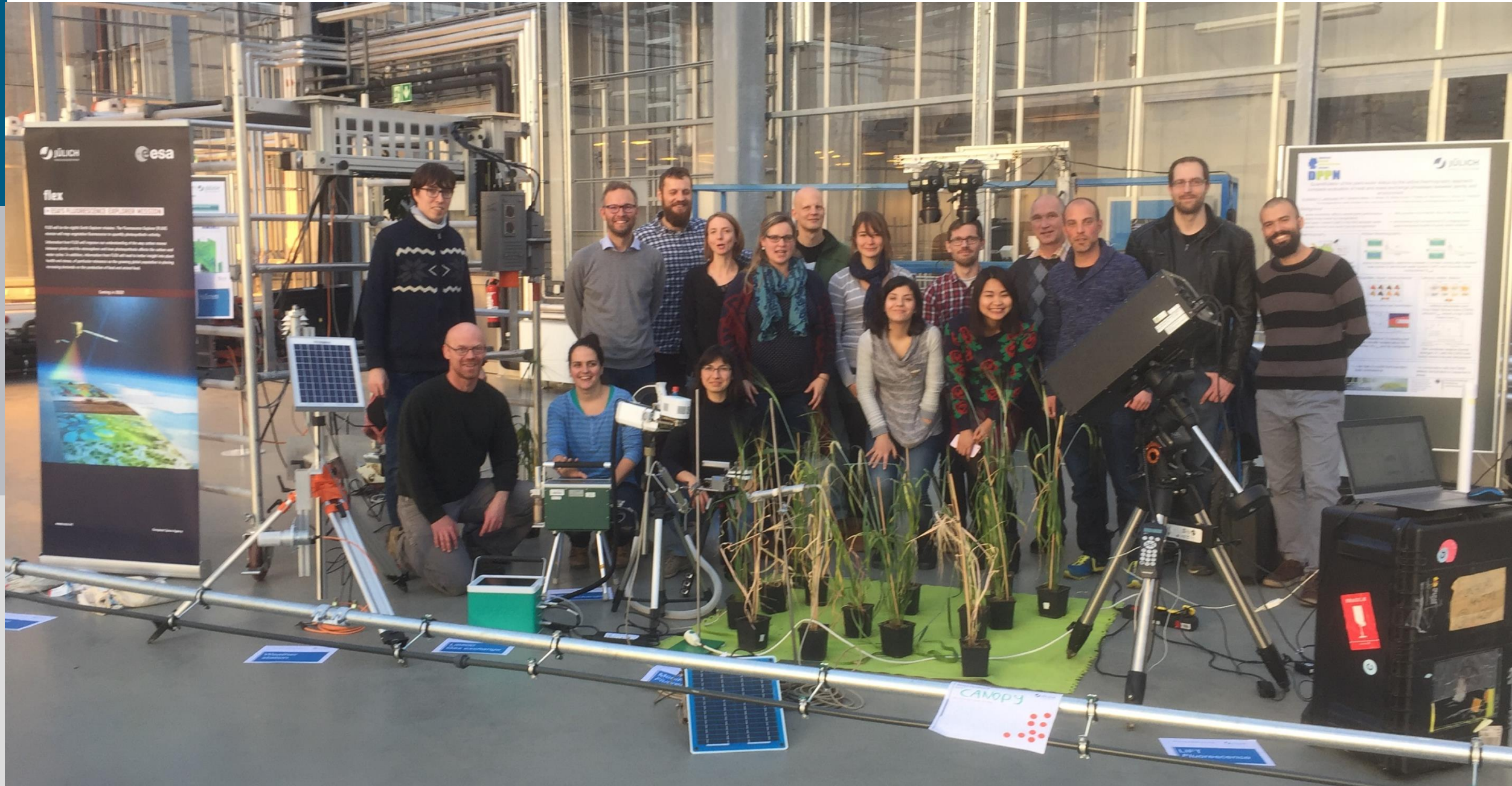
<https://apps.sentinel-hub.com/eo-browser/>

Summary – u.rascher@fz-juelich.de / 0170-2219199

- ❑ Disease symptoms are complex and already at the close range (proximity sensing) a meaningful combination of spectrally resolved data, covering the right range and advanced data processing is needed
- ❑ Small scale symptoms are often obscured on the larger scale and hidden correlations may hamper diseases detection from the distance
- ❑ Combination of spectrally resolved reflectance, fluorescence and thermal approaches shall be further exploited
- ❑ Combination of ground based systems (high temporal resolution and specificity) with airborne & satellite data (large coverage) is needed



Many thanks to my group



Many thanks to the numerous partners



University of Zurich
UZH



Consiglio Nazionale delle Ricerche

University of Cologne



universität**bonn**

Deutsche Forschungsgemeinschaft
DFG



Centre de Recherche Public Gabriel Lippmann



JÜLICH
FORSCHUNGSZENTRUM



P & M Technologies
Innovations in Plant Science & Technology

FAU

FRIEDRICH-ALEXANDER UNIVERSITÄT ERLANGEN-NÜRNBERG

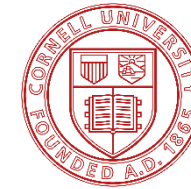
esa

Freie Universität Berlin



CzechGlobe

cnes



HELMHOLTZ GEMEINSCHAFT

CARNEGIE INSTITUTION
DEPARTMENT OF GLOBAL ECOLOGY
Extending the Frontiers of Science



HELMHOLTZ CENTRE FOR ENVIRONMENTAL RESEARCH - UFZ

Embrapa



Institut Pierre Simon Laplace



UNIVERSITY OF WOLLONGONG

FONDAZIONE Clima e Sostenibilità



Bundesministerium für Bildung und Forschung

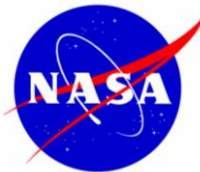


Bundesministerium für Ernährung und Landwirtschaft

UNIVERSITÀ DEGLI STUDI DI MILANO
BICOCCA



UNIVERSITY OF TWENTE



Goddard
SPACE FLIGHT CENTER



UNIVERSITAT DE VALÈNCIA

cost
EUROPEAN COOPERATION IN SCIENCE AND TECHNOLOGY

GFZ
Helmholtz Centre POTSDAM

Heinrich Heine
HEINRICH HEINE UNIVERSITÄT DÜSSELDORF

MICHIGAN STATE UNIVERSITY



WAGENINGEN
UNIVERSITY & RESEARCH

LMU

LÜDWIG-MAXIMILIANS-UNIVERSITÄT MÜNCHEN

BILL & MELINDA GATES foundation