

# → 7th ADVANCED TRAINING COURSE ON LAND REMOTE SENSING

4–9 September 2017 | Szent István University | Gödöllő, Hungary





# MONITORING VEGETATION IN A CHANGING CLIMATE

[D5T3a]

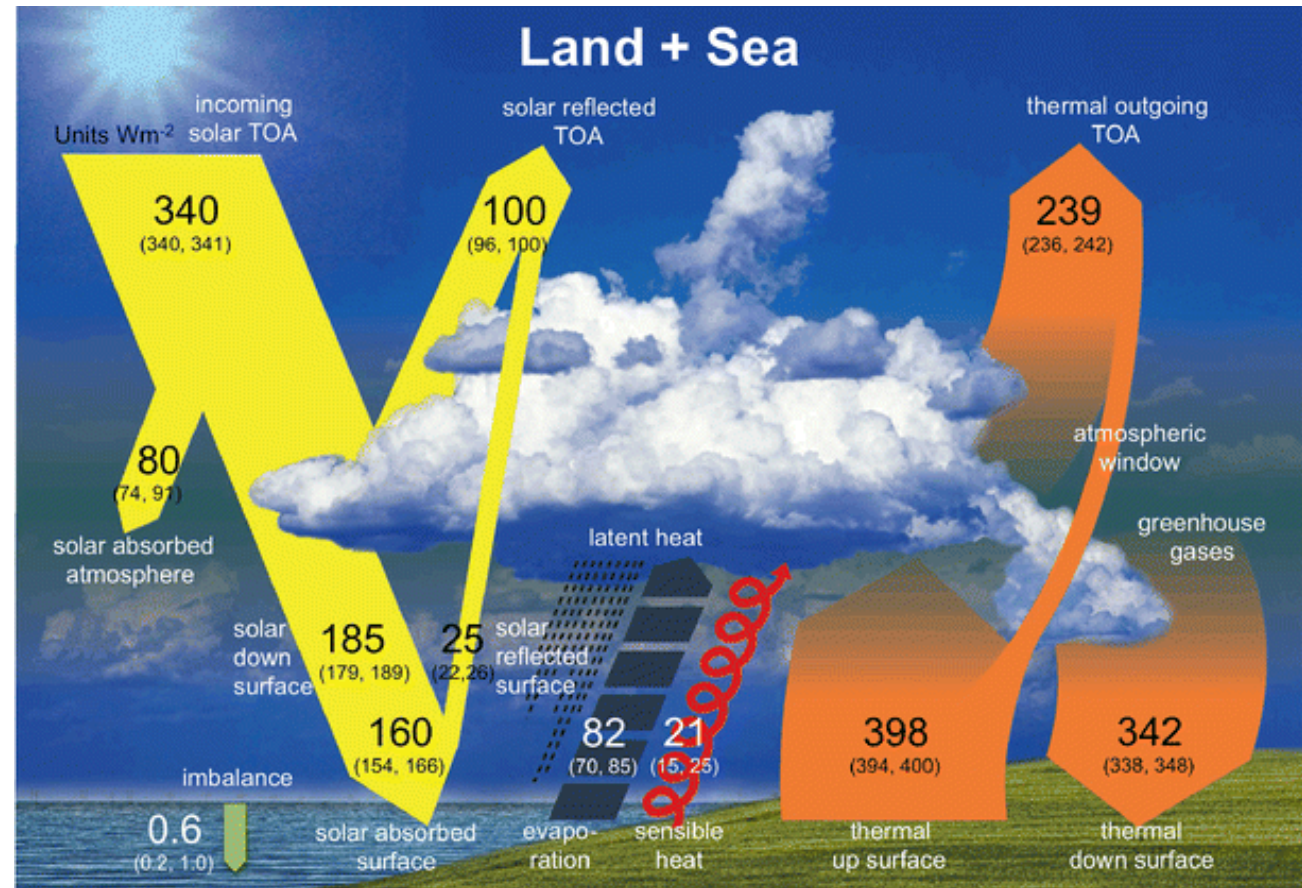
**Gregory Duveiller**

*European Commission Joint Research Centre*



# Why/how is the climate changing?

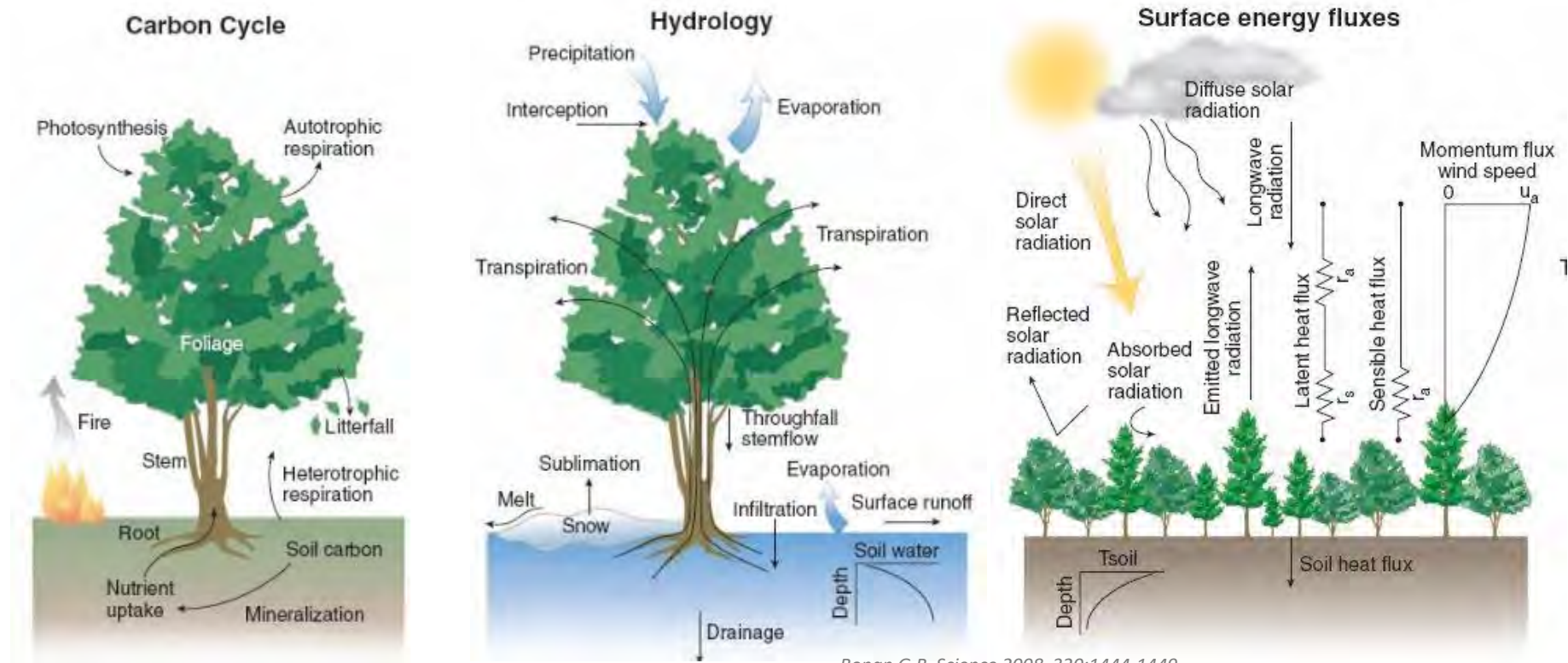
The Global Energy (im)balance



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# Vegetation interacts with climate in various ways



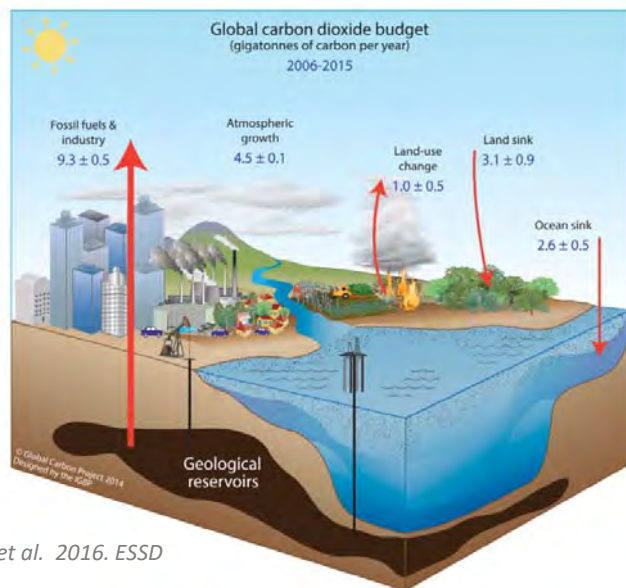
Bonan G.B. Science 2008, 320:1444-1449

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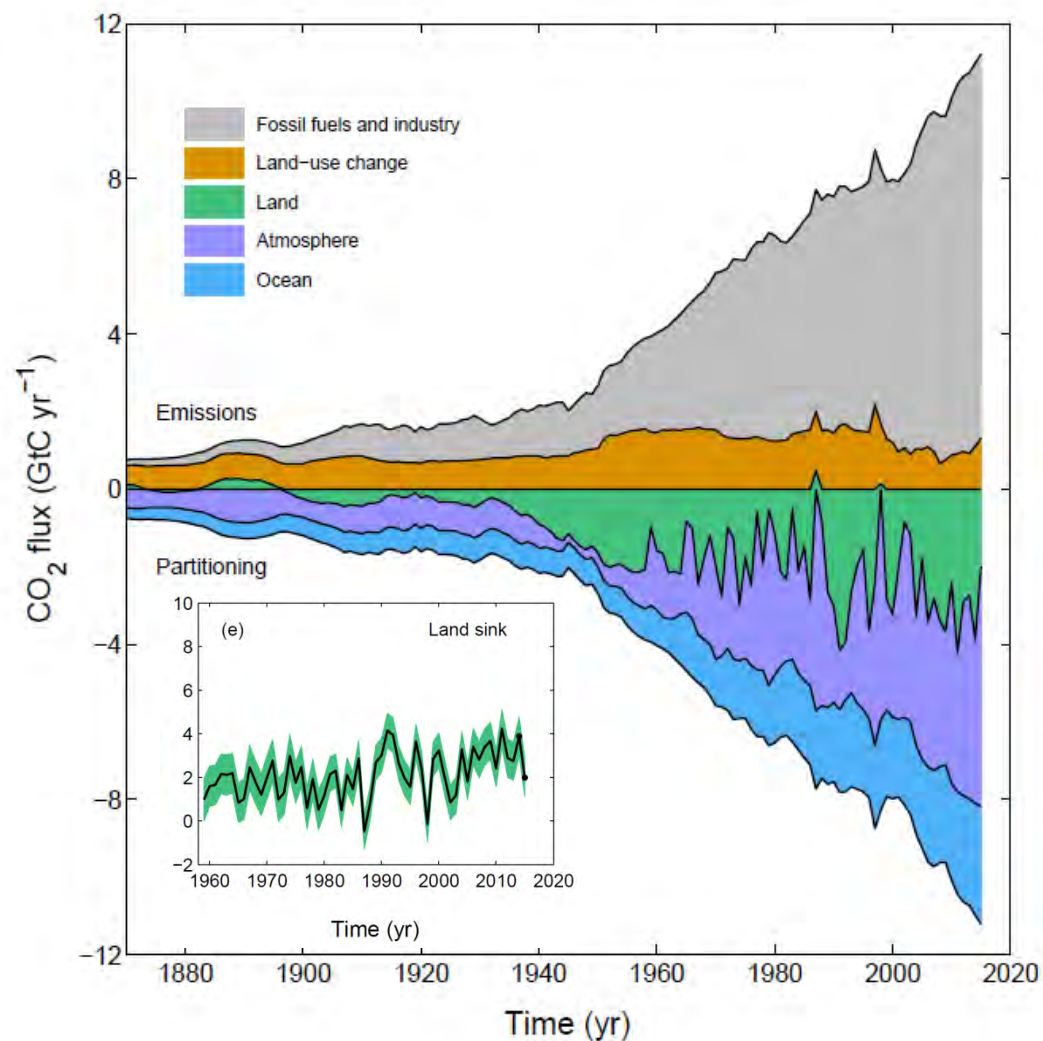


# Carbon fluxes



Le Quere et al. 2016. ESSD

Terrestrial vegetation is the most variable and uncertain component in the global carbon balance



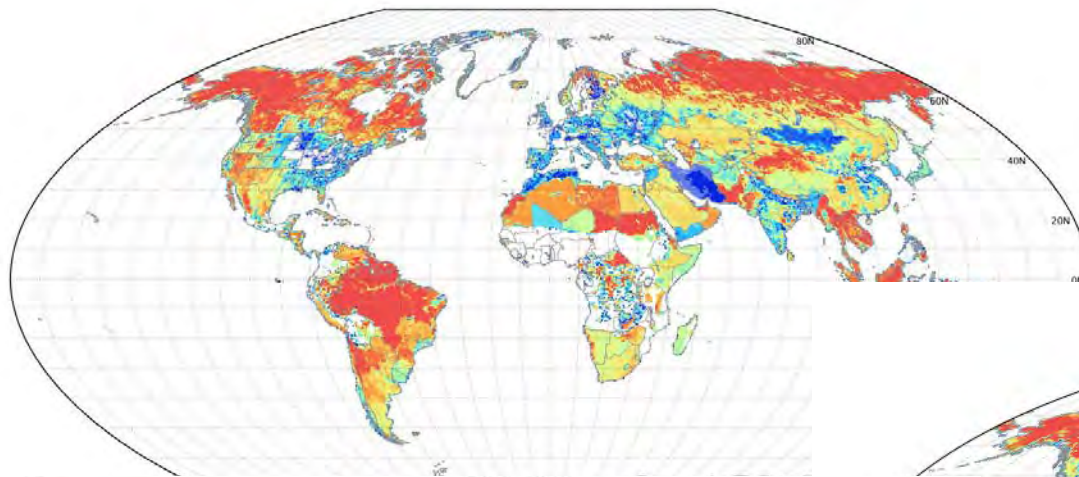
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# The land is changing ...

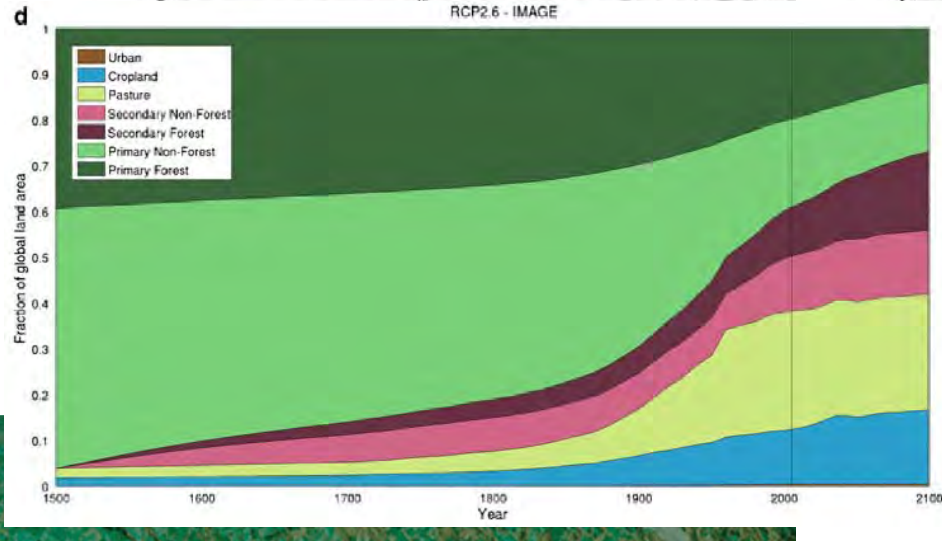
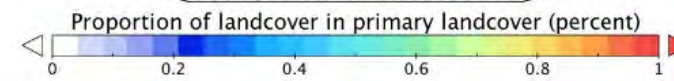
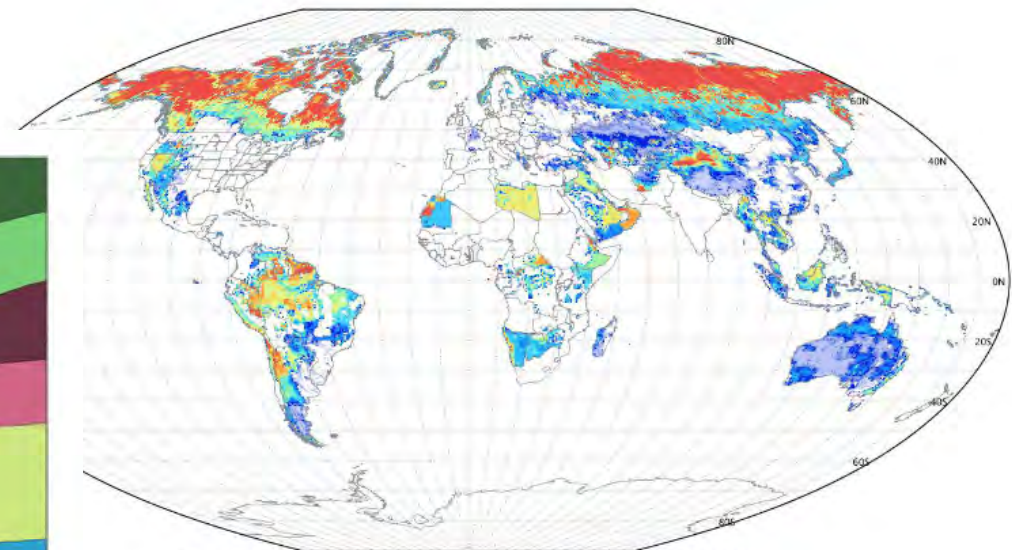
G. C. Hurtt et al 2011. *Clim. Change*

Year 1900

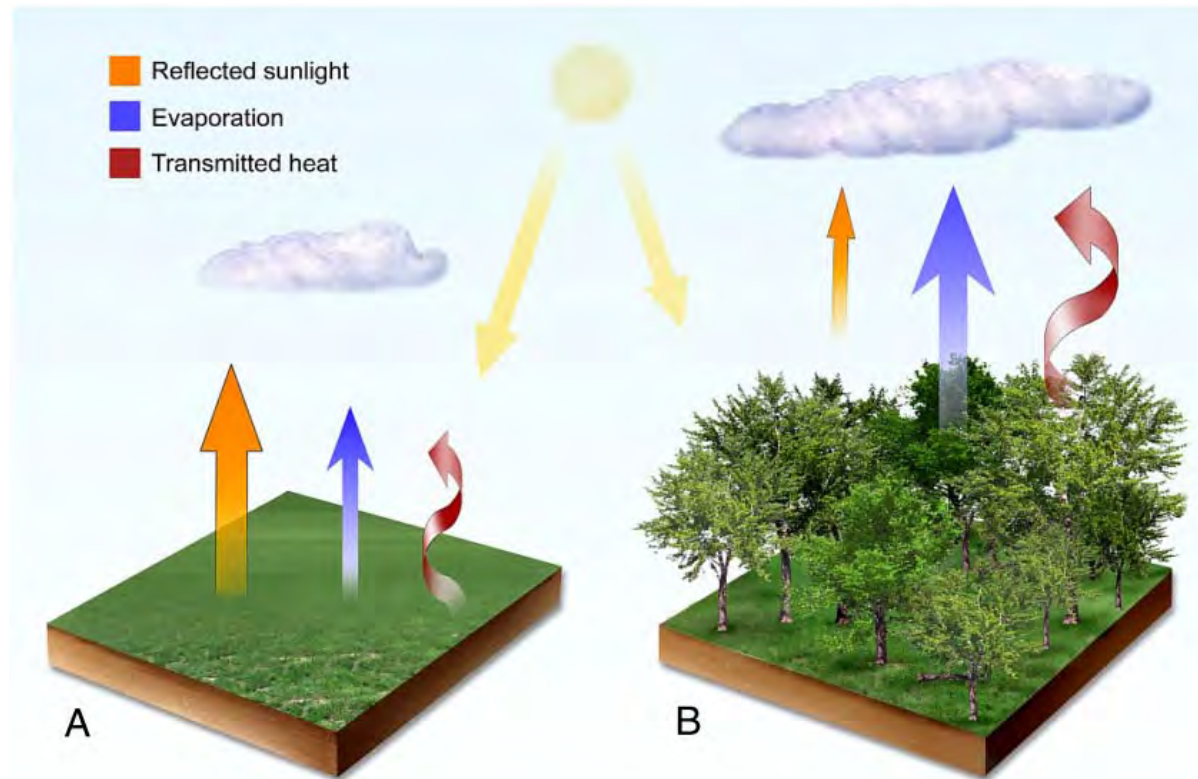


RCP2.6 - IMAGE

Year 2100



## Land cover changes has local biophysical impacts



Source:

*Jackson, R. B., Randerson, J. T., Canadell, J. G., Anderson, R. G., Avissar, R., Baldocchi, D. D., ... Pataki, D. E. (2008). Protecting climate with forests. Environmental Research Letters, 3(4)*



# ALBEDO

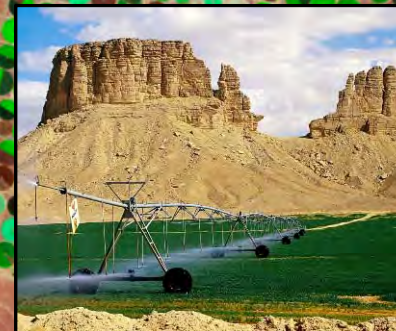
Low albedo → High Absorbed radiation → HIGH Air Temp

High albedo → Low Absorbed radiations → LOW Air Temp

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Water limitation → Low ET → HIGH Air Temp

Irrigation → Higher ET → Lower Air Temp



# EVAPOTRANSPIRATION



## What would we want to extract from RS?

- Vegetation productivity and biomass
- Energy fluxes (LST, albedo, ET)
- Vegetation type and change

**SPACE and TIME**

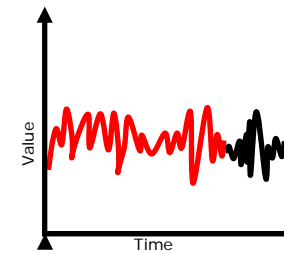
## What do we use these estimates for?

- To better understand land surface processes
  - To monitor changes in land surface at global to local scales
- To benchmark, calibrate and parametrize land surface models

# Requirements for climate data

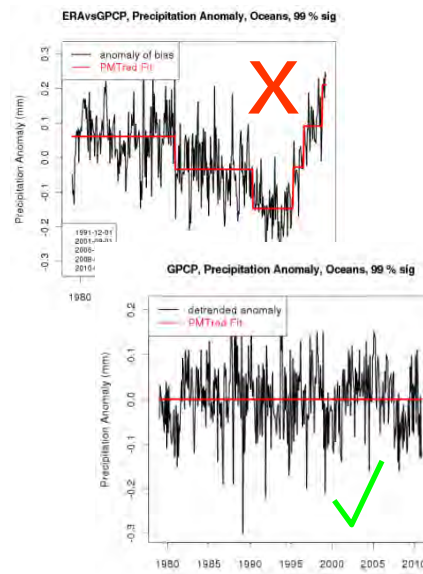
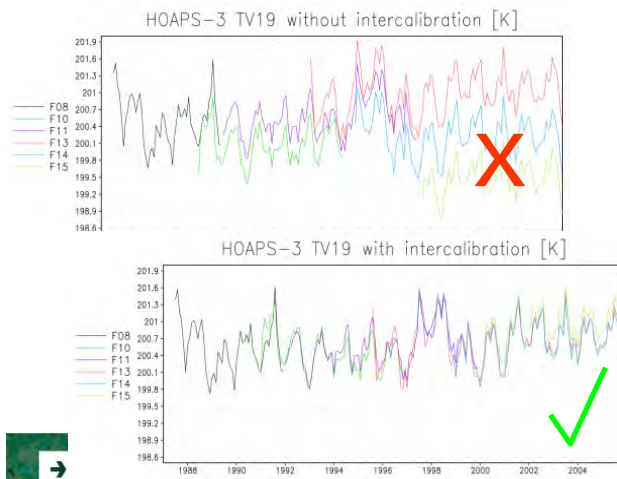
Slide adapted from Jędrzej S. Bojanowsky

Homogeneous

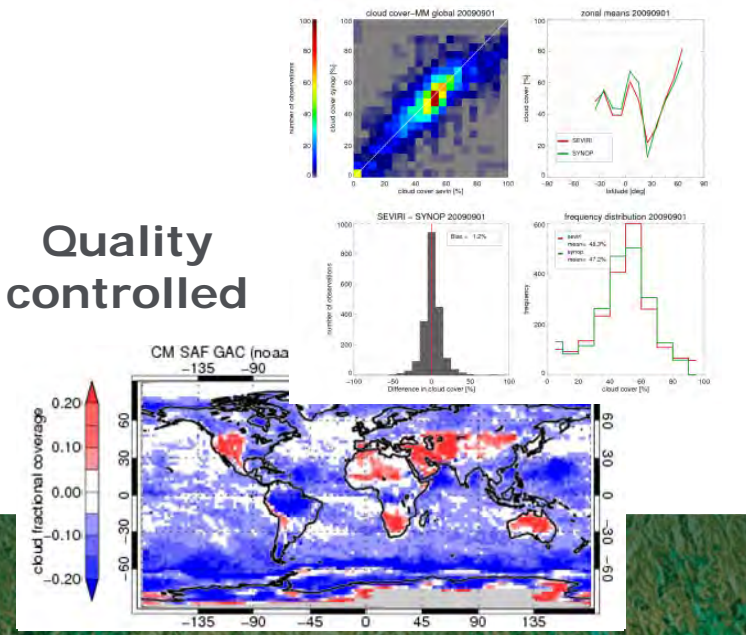


Sufficiently long  
time series

Calibrated



Quality  
controlled



LAND REMOTE SENSING



# ESA Climate Change Initiative (CCI)

The Global Climate Observing System (GCOS) developed the concept of the **Essential Climate Variable** (ECV).

**ECVs:** Physical, chemical or biological variable (or group of linked variables) that critically contributes to the characterisation of Earth's climate.

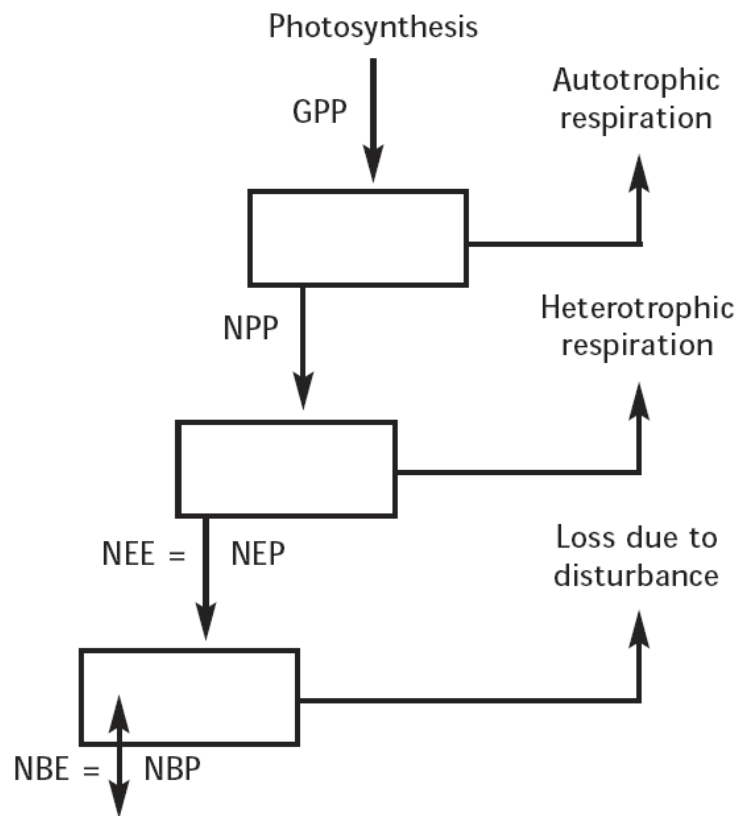
ECVs are defined based on criteria of:

*Relevance, Feasibility and Cost effectiveness*

The CCI program is the response of ESA to GCOS.



# Ecological terms commonly used in carbon accounting



*Kirschbaum, M. U. F., et al. "Definitions of some ecological terms commonly used in carbon accounting." Cooperative Research Centre for Carbon Accounting, Canberra (2001): 2-5.*

## **GPP [Gross Primary Production]:**

total amount of carbon fixed in the process of photosynthesis by plants in an ecosystem

## **NPP [Net Primary Production]:**

net production of organic matter by plants in an ecosystem, that is:  $GPP - \text{autotrophic respiration}$

## **NEP [Net Ecosystem Production]:**

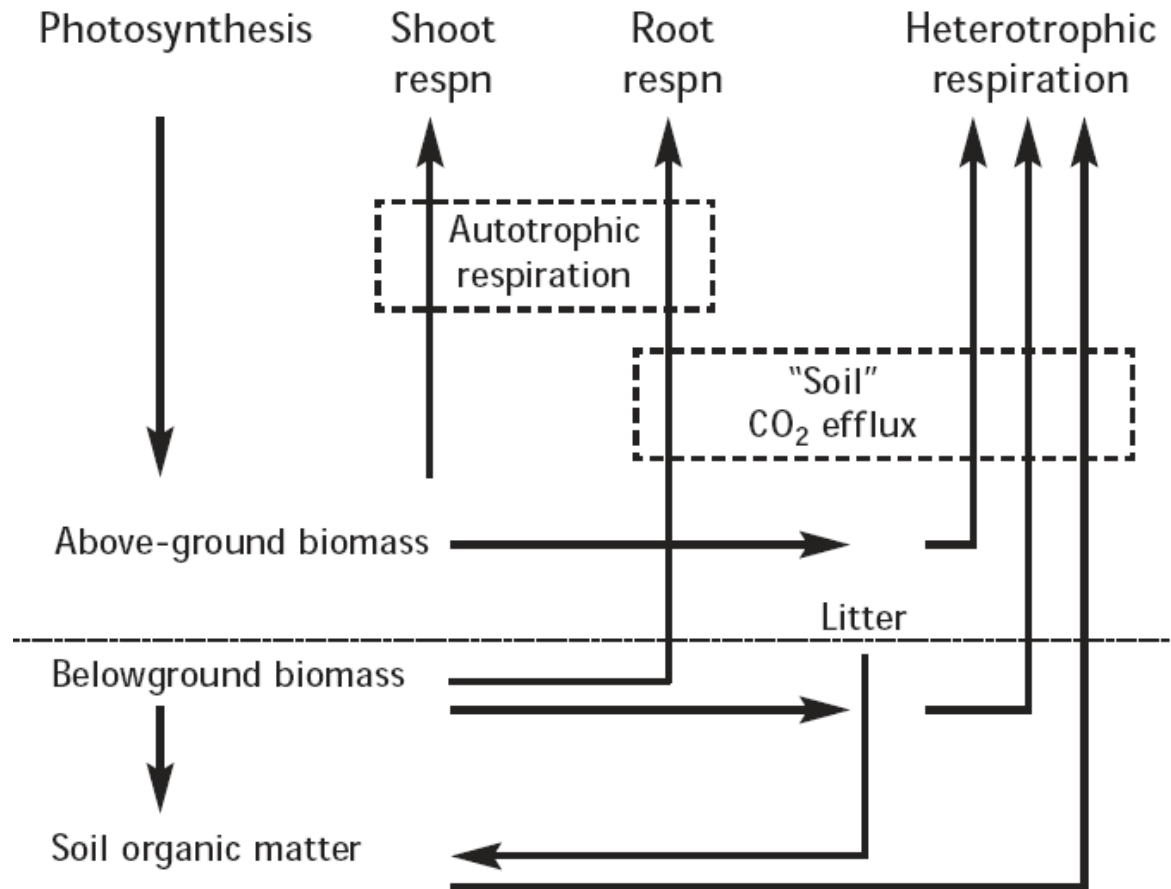
net accumulation of organic matter or carbon by an ecosystem;  $NEP = NPP - \text{heterotrophic resp.}$

## **NBP [Net Biosphere Production]:**

net production of organic matter in a region containing a range of ecosystems (a biome) minus what is lost by disturbance (harvest, forest clearance, and fire, etc.)



Complexity of  
measuring all  
components...



*Kirschbaum, M. U. F., et al. "Definitions of some ecological terms commonly used in carbon accounting." Cooperative Research Centre for Carbon Accounting, Canberra (2001): 2-5.*

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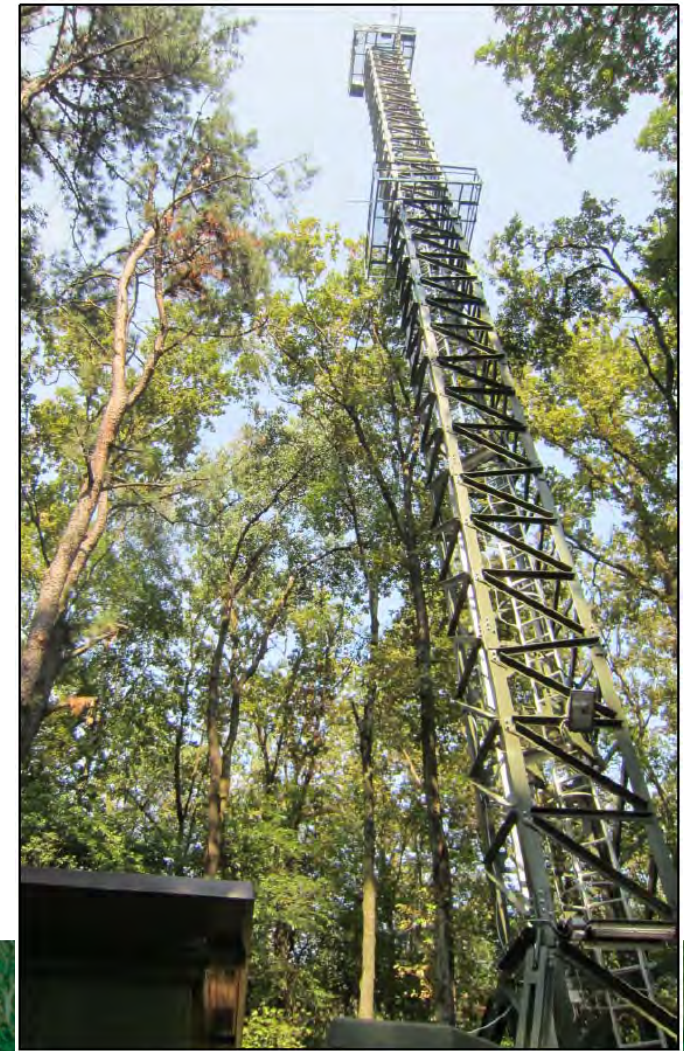
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# Productivity measured from flux-towers

Measures NEE (Net Ecosystem Exchange = NEP)

GPP can be derived, but already contains some  
modelling assumptions to remove respiration

Limited to a very localized area (~1km)



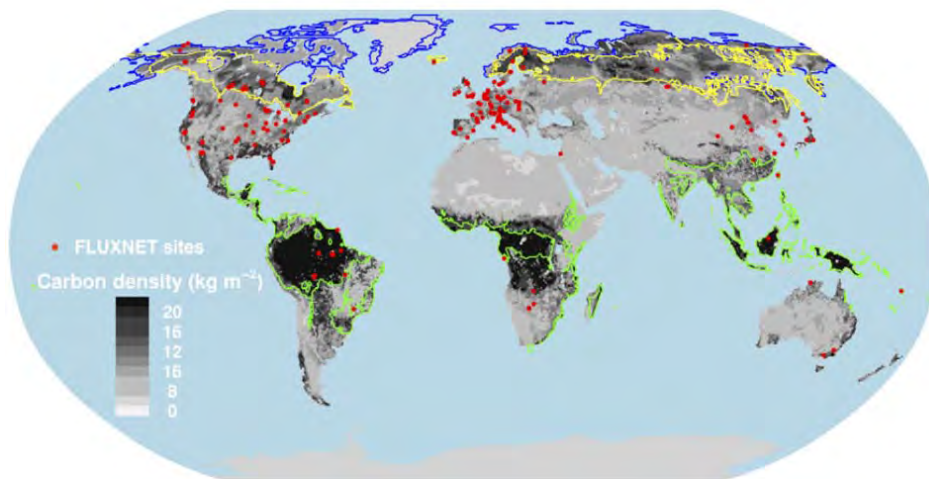
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# Productivity measured from flux-towers

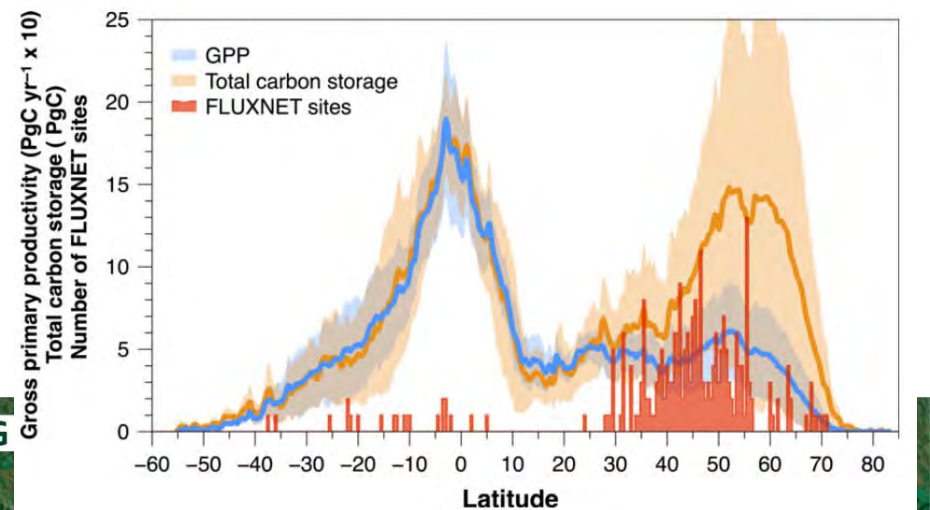
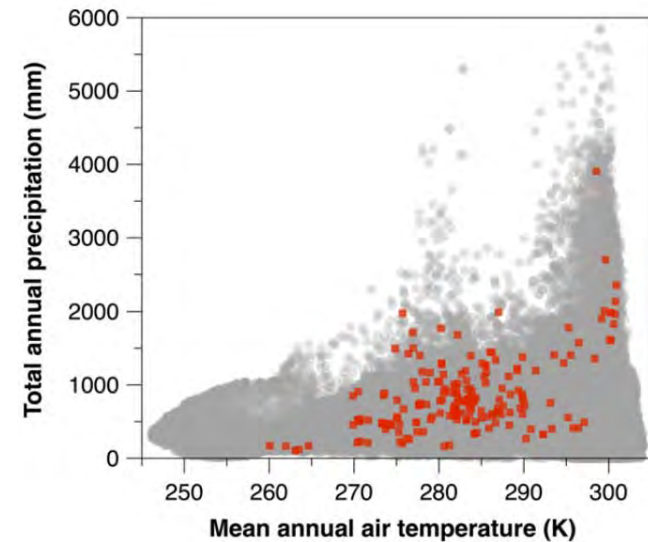
Sub-optimal spatial distribution despite reasonable climatic distribution



Schimel, D., Pavlick, R., Fisher, J. B., Asner, G. P., Saatchi, S. S., Townsend, P., ... Cox, P. (2015). Observing terrestrial ecosystems and the carbon cycle from space. *Global Change Biology*, 21(5), 1762–76. <https://doi.org/10.1111/gcb.12822>

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## Light use efficiency (Monteith approach)



*Monteith, J. L. (1972). Solar radiation and productivity in tropical ecosystems. Journal of Applied Ecology, 9(3), 747–766.*

$$\text{Canopy productivity} = \text{Radiation interception} * \text{Radiation use efficiency}$$

*Simple approach that can be linked to remote sensing observations*



# PAR, APAR and fAPAR

## Photosynthetically Active Radiation (PAR):

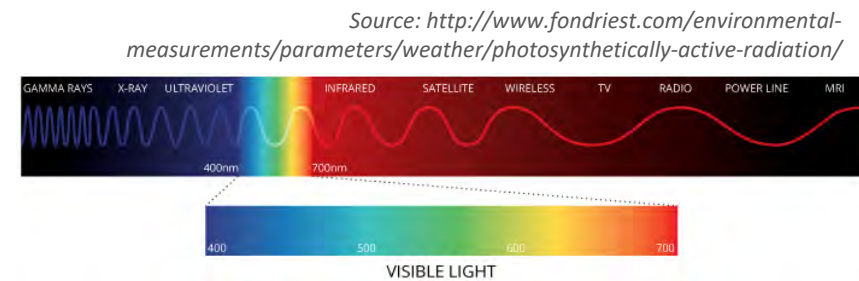
Radiation between 400 and 700 nm that photosynthetic organisms are able to use in the process of photosynthesis.

Coincides with visible light [Units:  $\mu\text{mol photons m}^{-2} \text{s}^{-1}$ ]

## Absorbed PAR (APAR):

Quantity of PAR absorbed by the plants

Often considered equal to  
*intercepted* PAR



## Fraction of APAR (fAPAR):

Normalized variable between 0 and 1

$fAPAR = APAR/PAR$

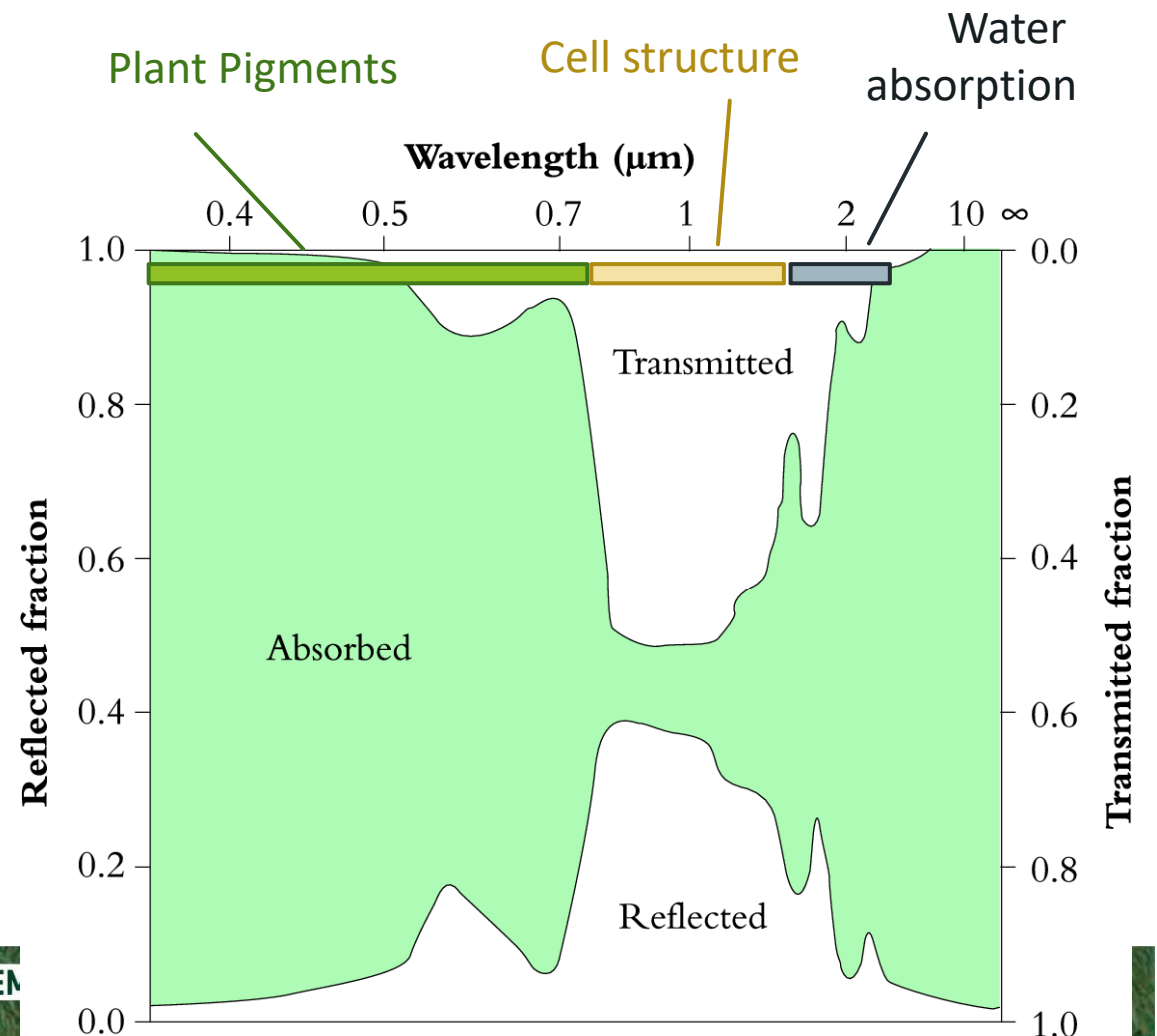
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# Spectral properties of vegetation



Source: *Plants in Action*, published by the Australian Society of Plant Scientists, <http://plantsinaction.science.uq.edu.au/edition1/>

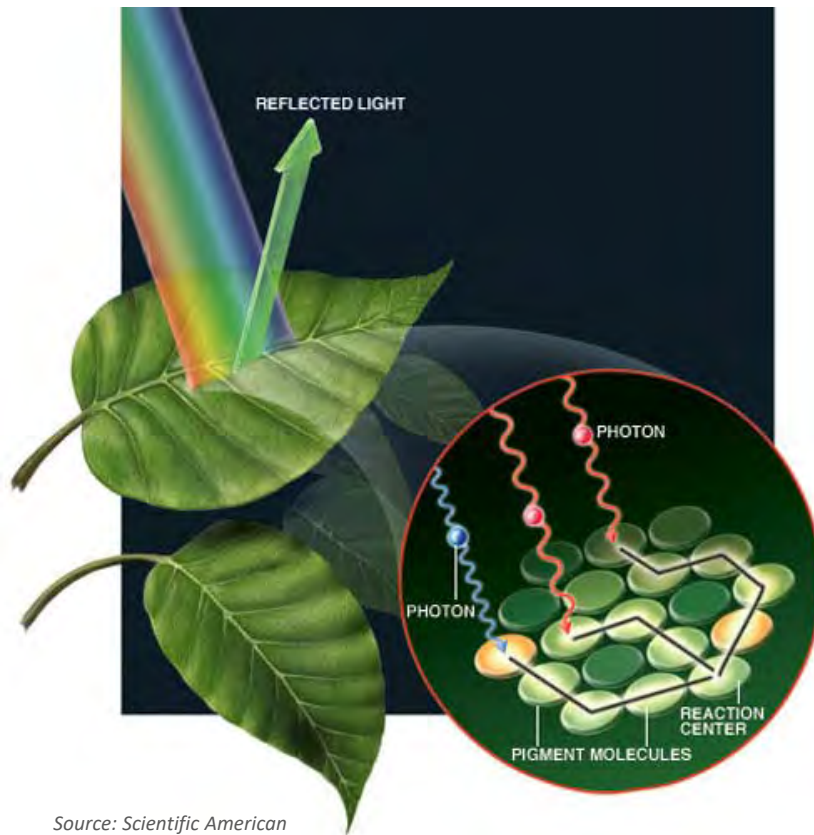


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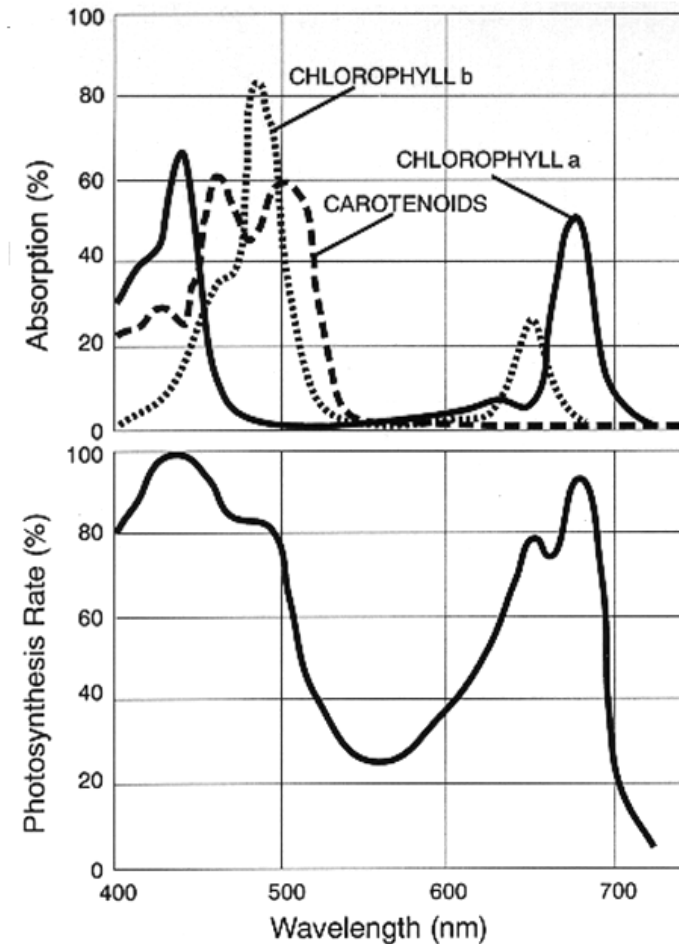
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# Plant pigments



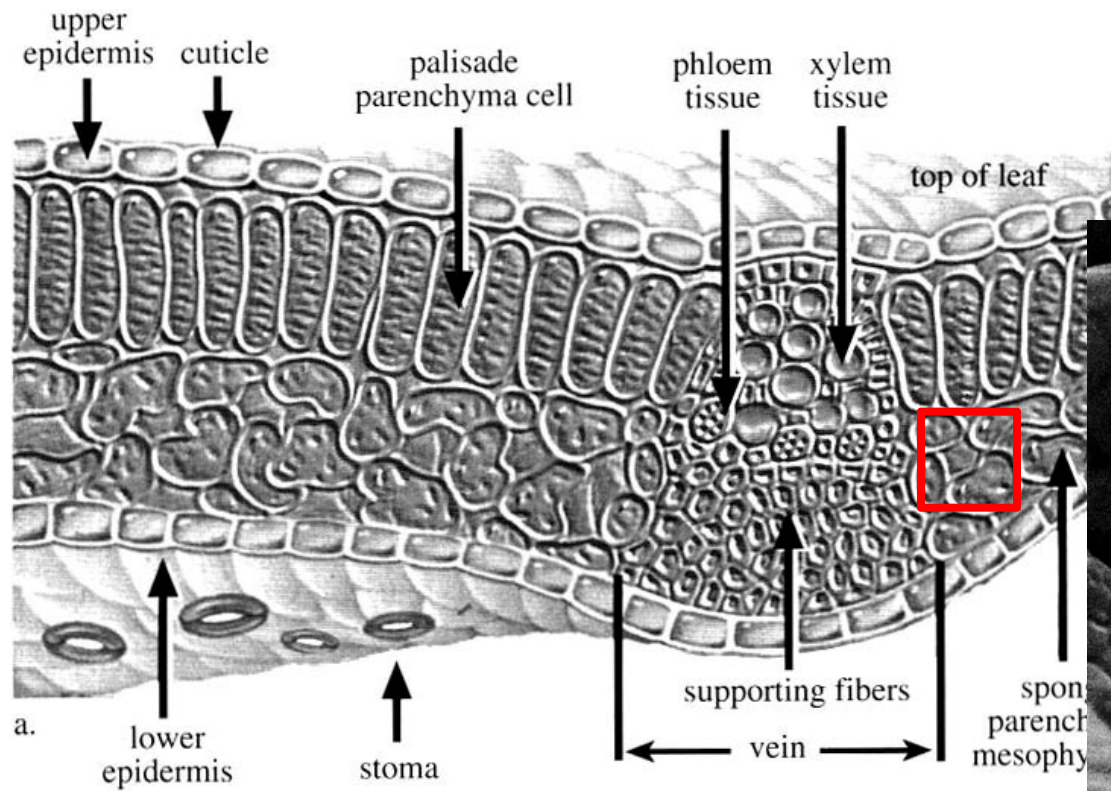
Source: <http://www.life.uiuc.edu/govindjee/paper/gov.html>, from "Concepts in Photobiology: Photosynthesis and Photomorphogenesis", Edited by GS Singhal, G Renger, SK Sopory, K-D Irrgang



Source: Scientific American

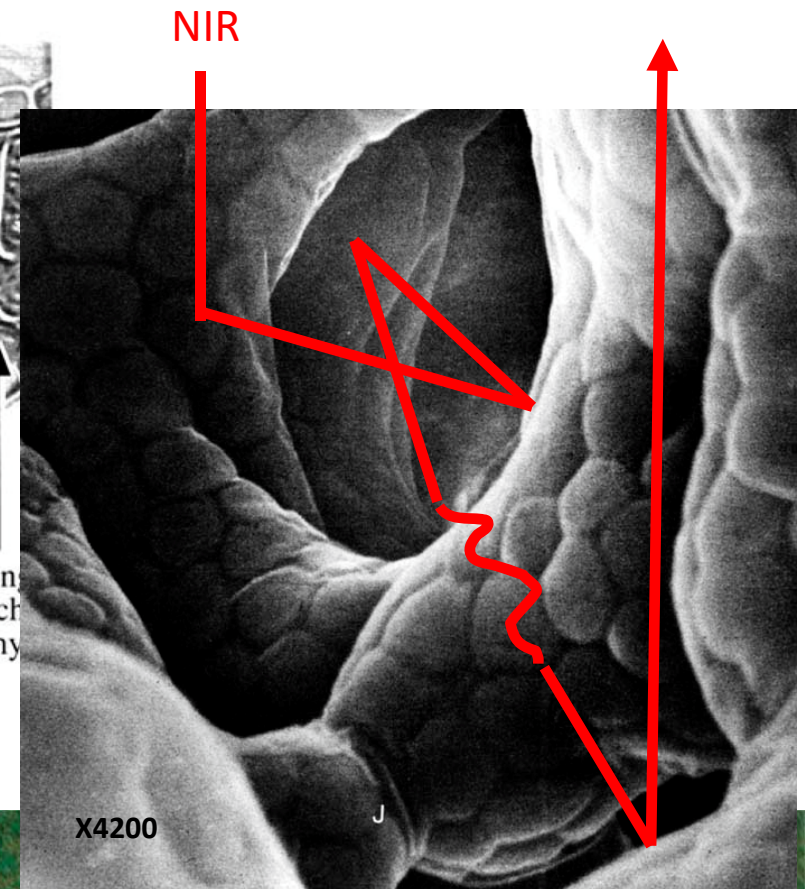
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NSING



Cell structure strongly scatters NIR radiation to prevent cell damage

## Cell Structure

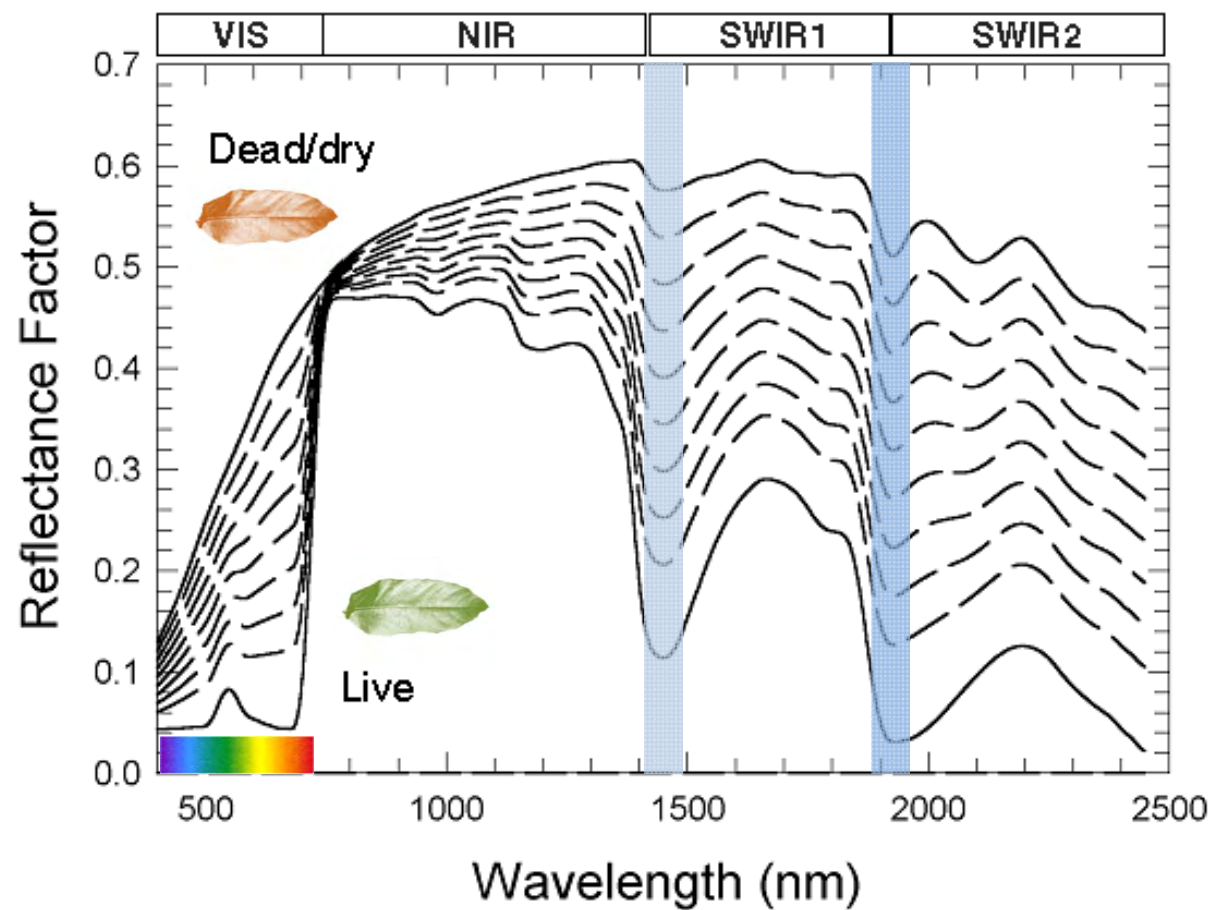


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## Absorption from water in the plants

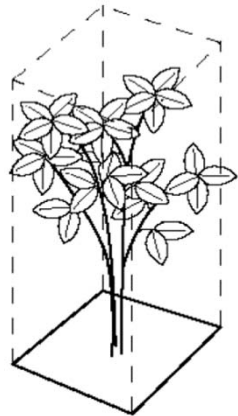


Source:  
<http://www.exelisvis.com/docs/NonPhotosyntheticVegetation.html>  
[incorrectly cited as coming from:  
Asner, G.P., 1998. RSE.]

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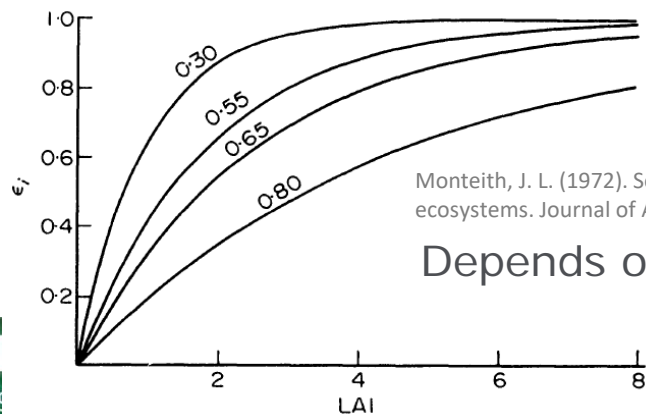
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# Leaf Area Index (LAI)



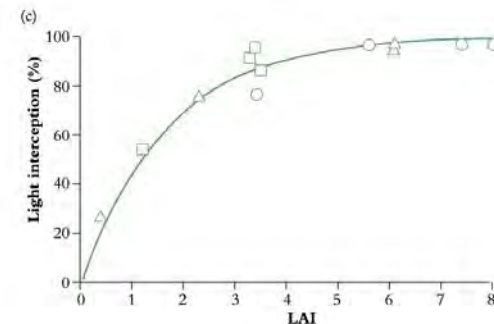
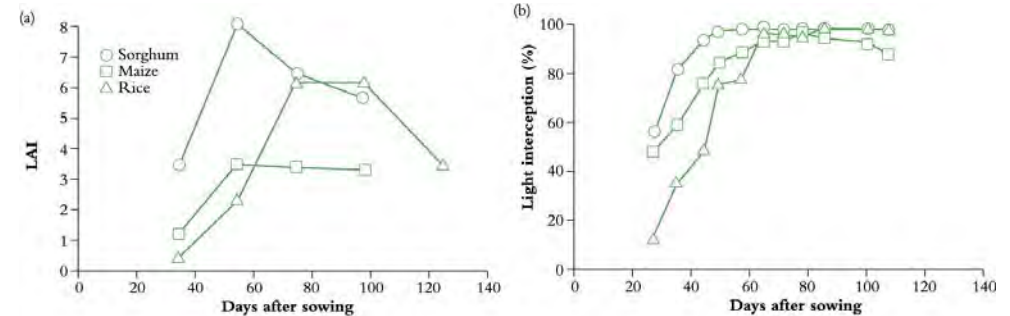
Defined as half the total developed area of green leaves per unit of ground horizontal surface area [units: m<sup>2</sup> m<sup>-2</sup>]

Interface between atmosphere and vegetation.



Monteith, J. L. (1972). Solar radiation and productivity in tropical ecosystems. *Journal of Applied Ecology*, 9(3), 747–766.

Depends on leaf geometry



Useful to describe light interception:  $I = I_0 e^{-kLAI}$

REMOTE SENSING



# Measuring 'greenness'

## Normalized Difference Vegetation Index

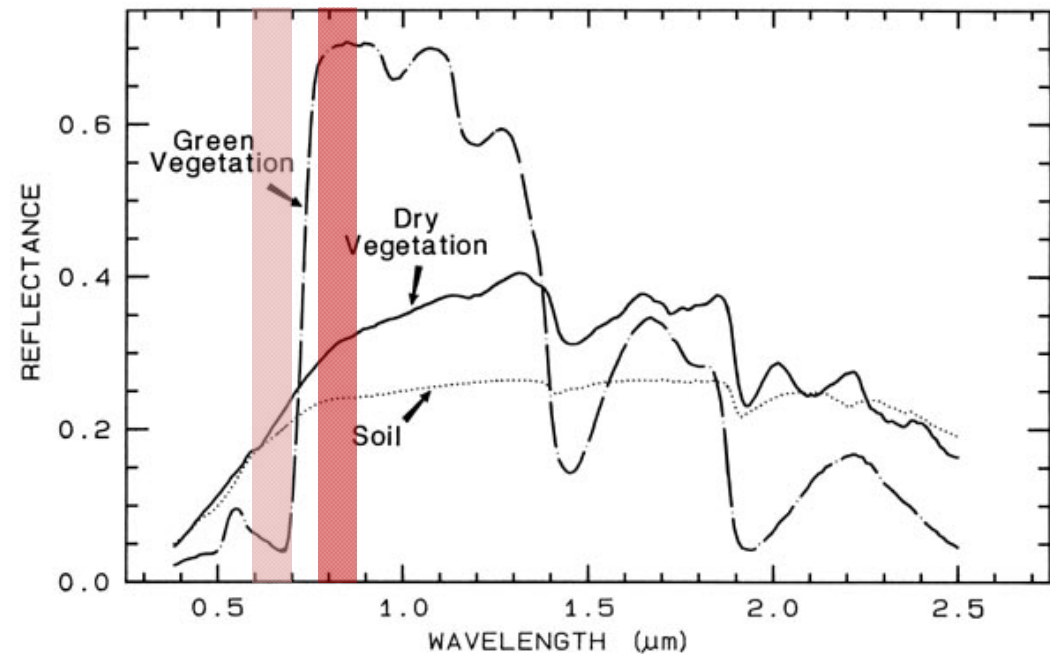
$$\text{NDVI} = \frac{\text{NIR} - \text{Red}}{\text{NIR} + \text{Red}}$$

Exploits particular spectral properties of vegetation

Partly independent of viewing geometry

Proposed by Rouse et al. 1974

Popularized by Tucker since 1980



# Vegetation indices

- Convenient way to resume information
- Exploit the particular spectral properties of vegetation
- Depend on spectral response of the sensors (which changes even for bands with same names)
- Potentially unlimited number of combinations

Differenced Vegetation Index	DVI
Misra Soil Brightness Index	MSBI
Misra Green Vegetation Index	MGVI
Misra Yellow Vegetation Index	MYVI
Misra Non Such Index	MNSI
Perpendicular Vegetation Index	PVI
Ashburn Vegetation Index	AVI
Greenness Above Bare Soil	GRABS
Multi-Temporal Vegetation Index	MTVI
Greenness Vegetation and Soil Brightness	GVSB
Adjusted Soil Brightness Index	ASBI
Adjusted Green Vegetation Index	AGVI
Transformed Vegetation Index	TVI
Differenced Vegetation Index	DVI
Normalized Difference Greenness Index	NDGI
Redness Index	RI
Normalized Difference Vegetation Index	NDVI
Perpendicular Vegetation Index	PVI
Soil Adjusted Vegetation Index	SAVI
Transformed SAVI	TSAVI
Transformed SAVI	TSAVI
Atmospherically Resistant Vegetation Index	ARVI
Global Environment Monitoring Index	GEMI
Transformed Soil Atmospherically Resistant Vegetation Index	TSARVI
Modified SAVI	MSAVI
Angular Vegetation Index	AVI

$$\begin{aligned}
 & (2.4\text{MSS7} - \text{MSS5}) \\
 & (0.406\text{MSS4} + 0.600\text{MSS5} + 0.645\text{MSS6} + 0.243\text{MSS7}) \\
 & (-0.386\text{MSS4} - 0.530\text{MSS5} + 0.535\text{MSS6} + 0.532\text{MSS7}) \\
 & (0.723\text{MSS4} - 0.597\text{MSS5} + 0.206\text{MSS6} - 0.278\text{MSS7}) \\
 & (0.404\text{MSS4} - 0.039\text{MSS5} - 0.505\text{MSS6} + 0.762\text{MSS7}) \\
 & \sqrt{(\rho_{\text{sol}} - \rho_{\text{veg}})_R^2 + (\rho_{\text{sol}} - \rho_{\text{veg}})_{\text{NIR}}^2} \\
 & (2.0\text{MSS7} - \text{MSS5}) \\
 & (\text{GVI} - 0.09178\text{SBI} + 5.58959) \\
 & (\text{NDVI}(\text{date } 2) - \text{NDVI}(\text{date } 1)) \\
 & \frac{\text{GVI}}{\text{SBI}} \\
 & (2.0 \text{ YVI}) \\
 & \text{GVI} - (1 + 0.018\text{GVI})\text{YVI} - \text{NSI}/2 \\
 & \frac{(\text{NDVI} + 0.5)}{|\text{NDVI} + 0.5|} \sqrt{|\text{NDVI} + 0.5|} \\
 & \frac{(\text{NIR} - \text{R})}{(\text{G} - \text{R})} \\
 & \frac{(\text{G} - \text{R})}{(\text{G} + \text{R})} \\
 & \frac{(\text{R} - \text{G})}{(\text{R} + \text{G})} \\
 & \frac{(\text{NIR} - \text{R})}{(\text{NIR} + \text{R})} \\
 & \frac{(\text{NIR} - a\text{R} - b)}{\sqrt{a^2 + 1}} \\
 & \frac{(\text{NIR} - \text{R})}{(\text{NIR} + \text{R} + \text{L})} (1 + \text{L}) \\
 & \frac{[a(\text{NIR} - a\text{R} - b)]}{(\text{R} + a\text{NIR} - ab)} \\
 & \frac{[a(\text{NIR} - a\text{R} - b)]}{[\text{R} + a\text{NIR} - ab + X(1 + a^2)]} \\
 & \frac{(\text{NIR} - \text{RB})}{(\text{NIR} + \text{RB})} \\
 & \text{RB} = \text{R} - \gamma(\text{B} - \text{R}) \\
 & \text{GEMI} = \eta(1 - 0.25\eta) - \frac{(\text{R} - 0.125)}{(1 - \text{R})} \\
 & \eta = \frac{[2(\text{NIR}^2 - \text{R}^2) + 1.5\text{NIR} + 0.5\text{R}]}{(\text{NIR} + \text{R} + 0.5)} \\
 & \frac{[a_{rb}(\text{NIR} - a_{rb}\text{RB} - b_{rb})]}{[\text{RB} + a_{rb}\text{NIR} - a_{rb}b_{rb} + X(1 + a_{rb}^2)]} \\
 & \frac{2\text{NIR} + 1 - \sqrt{(2\text{NIR} + 1)^2 - 8(\text{NIR} - \text{R})}}{2} \\
 & \tan^{-1} \left\{ \frac{\lambda_3 - \lambda_2}{\lambda_2} [\text{NIR} - \text{R}]^{-1} \right\} + \tan^{-1} \left\{ \frac{\lambda_2 - \lambda_1}{\lambda_2} [\text{G} - \text{R}]^{-1} \right\}
 \end{aligned}$$

Richardson and Wiegand, 1977
Misra et al., 1977
Misra et al., 1977
Misra et al., 1977
Misra et al., 1977
Richardson and Wiegand, 1977
Ashburn, 1978
Hay et al., 1979
Yazdani et al., 1981
Badhwar, 1981
Jackson et al., 1983
Jackson et al., 1983
Perry and Lautenschlager, 1984
Clevers, 1986
Chamard et al., 1991
Escadafal and Huete, 1991
Rouse et al., 1974
Jackson et al., 1980
Huete, 1988
Baret et al., 1989
Baret and Guyot, 1991
Kaufman and Tanré, 1992
Pinty and Verstraete, 1992
Bannari et al., 1994
Qi et al., 1994
Plummer et al., 1994

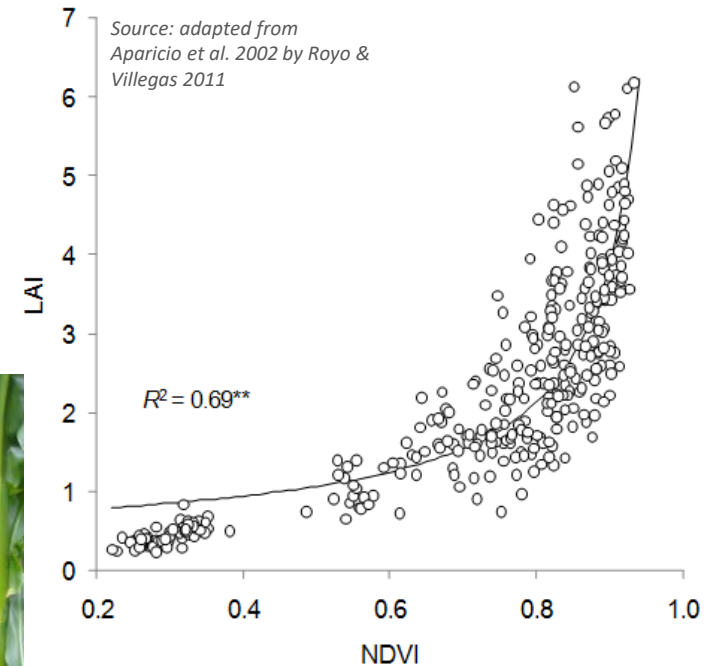
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# Retrieving biophysical variables (fAPAR, LAI) from RS

## Empirical methods

- Establishment of a statistical relationship between VI or  $\rho$  and field measured biophysical variables
- Require intensive field measurements for calibration and validation
- Relation is typically limited to large geographic extent





# Retrieving biophysical variables (fAPAR, LAI) from RS

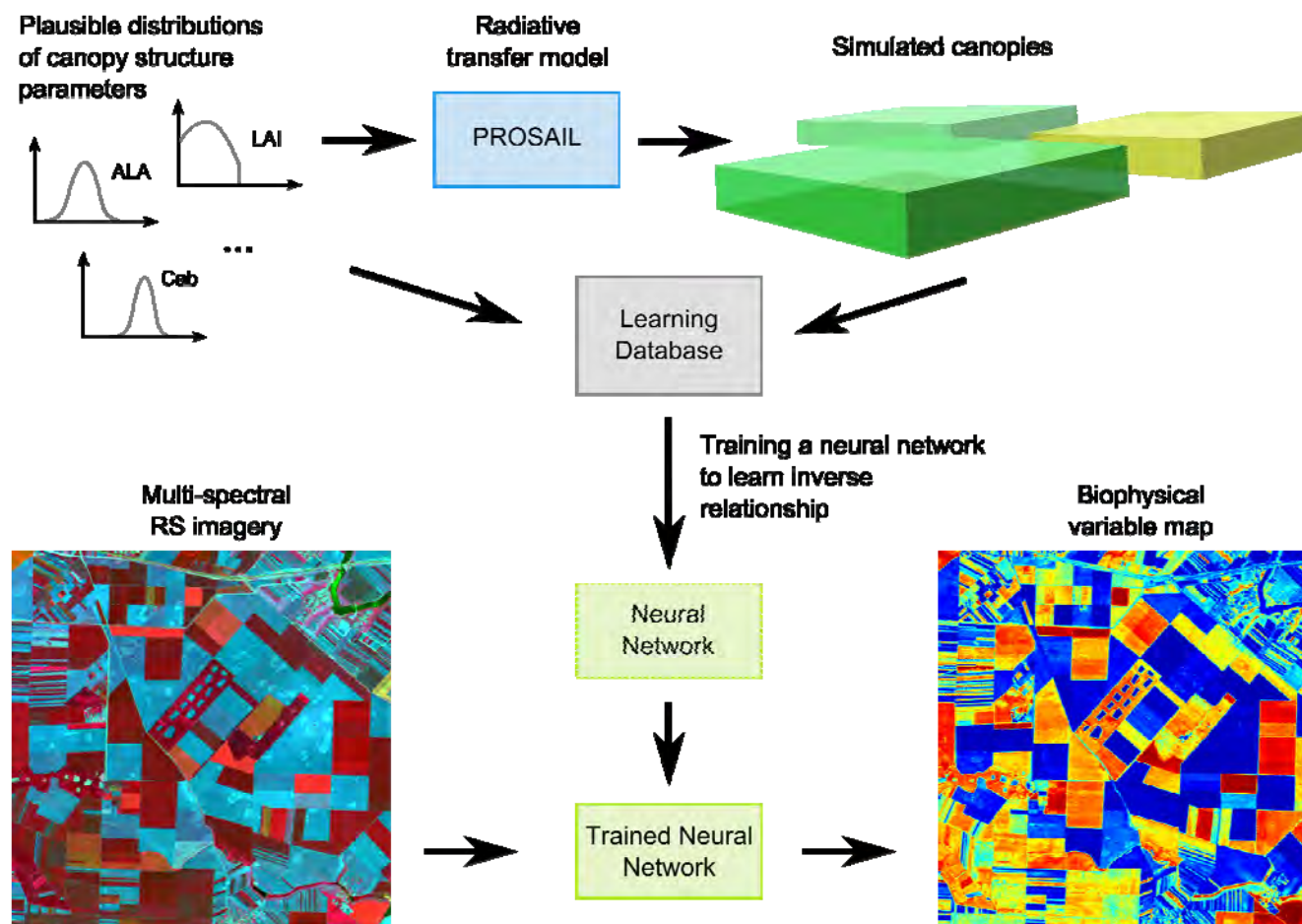
## Physical methods

- Replacement of field measurements by radiative transfer models (RTMs)



- Mathematical inversion necessary, but difficult because it is an ill-posed problem
- Method is transportable across landscapes as long as RTM is valid

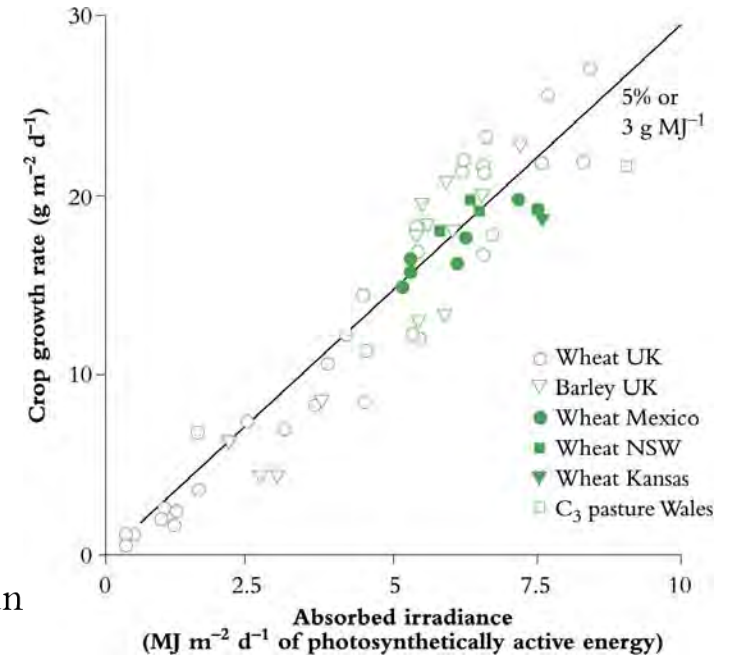
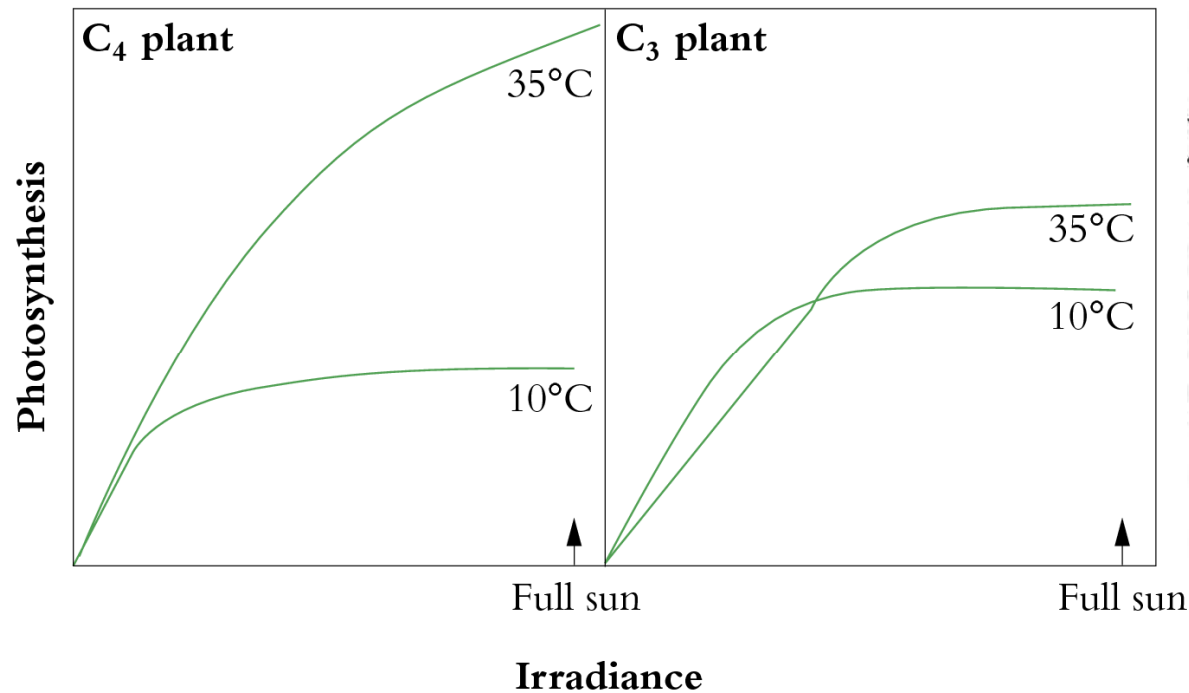
# Hybrid method with neural network inversion of RTM



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# Light use efficiency

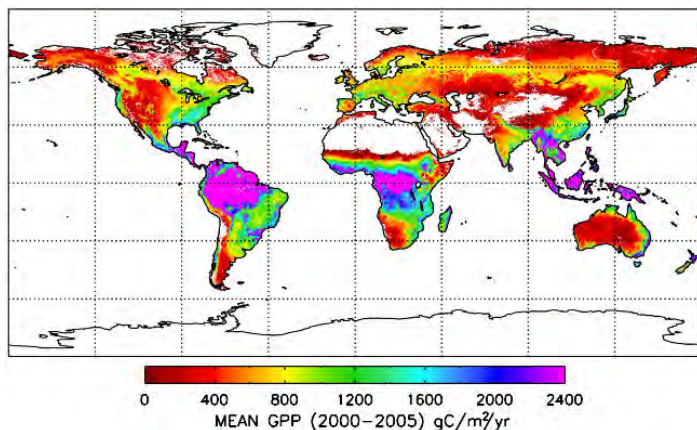


Source: *Plants in Action*, published by the Australian Society of Plant Scientists, <http://plantsinaction.science.uq.edu.au/edition1/>

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Source: <http://www.ntsug.umt.edu/project/modis/mod17.php>

## MODIS GPP/NPP Project (MOD17)

*Land cover/ biome maps*



*Set for different vegetation types and climatic conditions*

*Meteorology*



*Down-regulated if water-stressed and/or cold temperature condition*

$PAR * fAPAR$



Canopy  
productivity

=

Radiation  
interception

\*

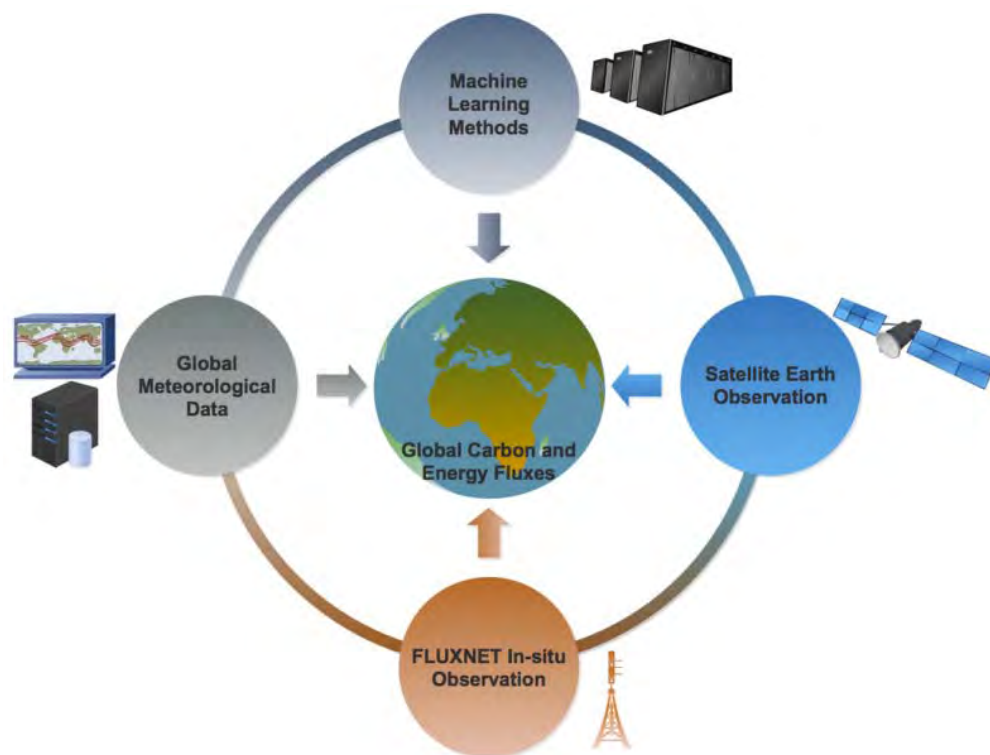
Radiation use  
efficiency



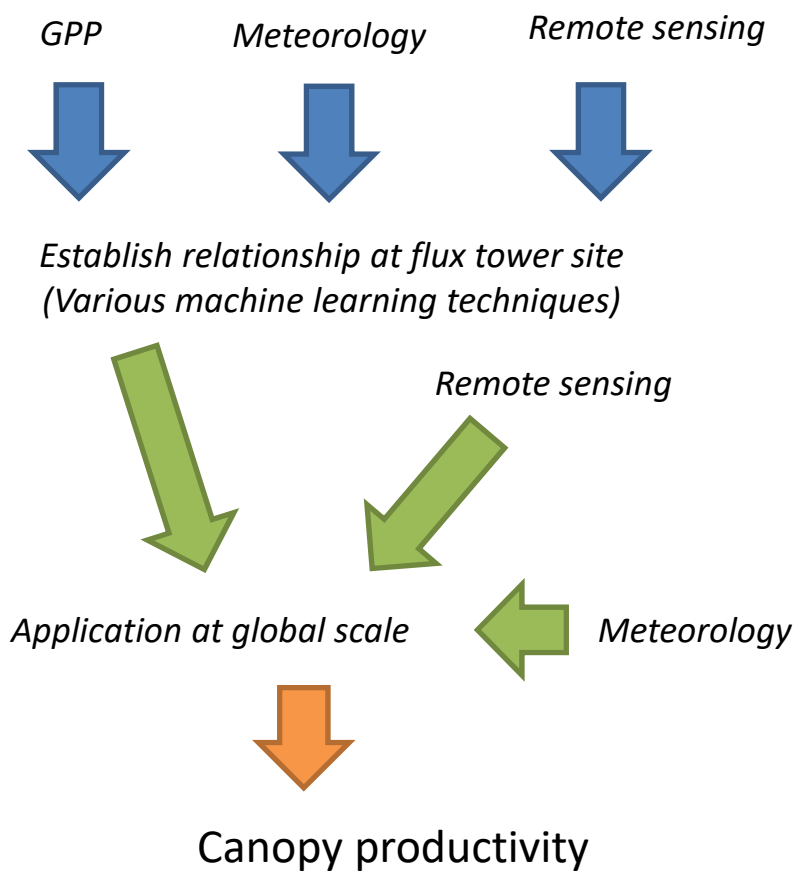
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# FLUXCOM



Source: <http://www.fluxcom.org/>

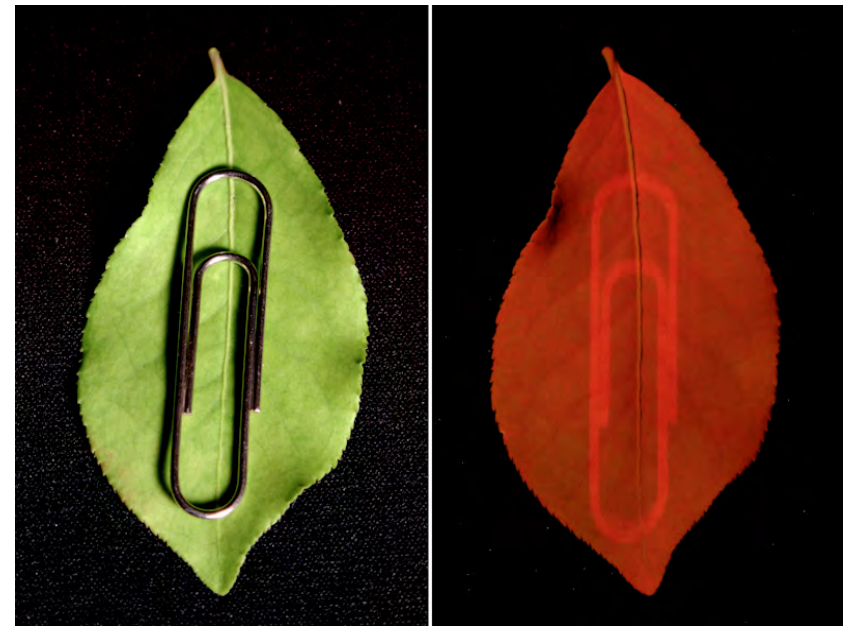


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## Exploring other avenues ...

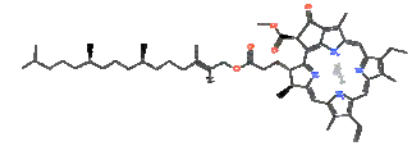
- (Sun-induced) chlorophyll fluorescence (SIF) emitted by the photosynthetic machinery
- Responds **instantaneously** to perturbations in the environmental conditions such as light and water stress
- This allows to translate effects of stress which do not necessarily cause a **reduction of Chl or LAI**
- Can provide early and direct diagnostic of functional status of vegetation...  
**proxy for photosynthetic activity**



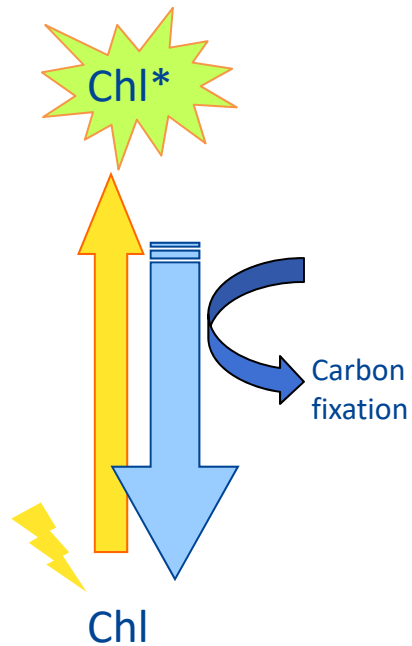
<http://www.nightsea.com/articles/fluorescence-photography-illuminates-chlorophyll/>



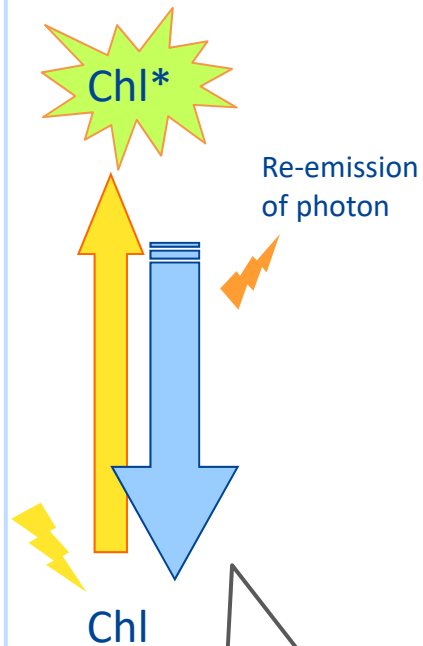
# The various fates of excited chlorophyll...



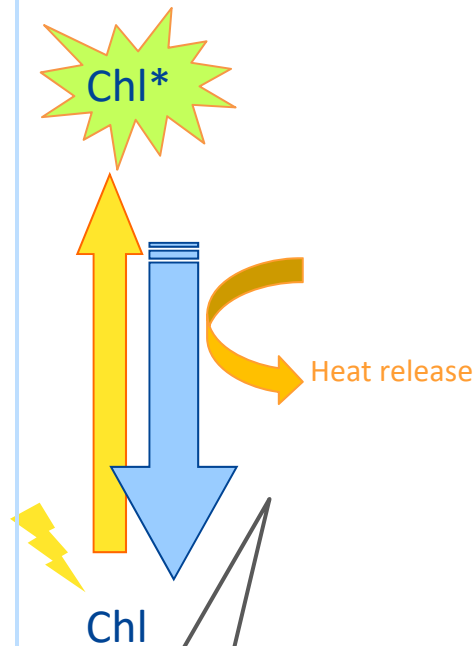
## Photochemistry



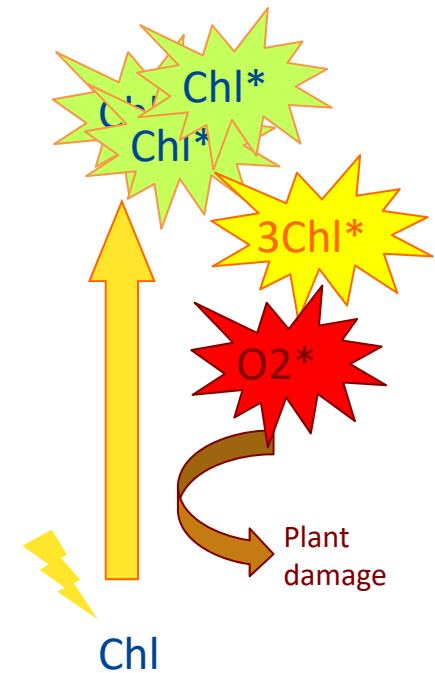
## Fluorescence



## NPQ



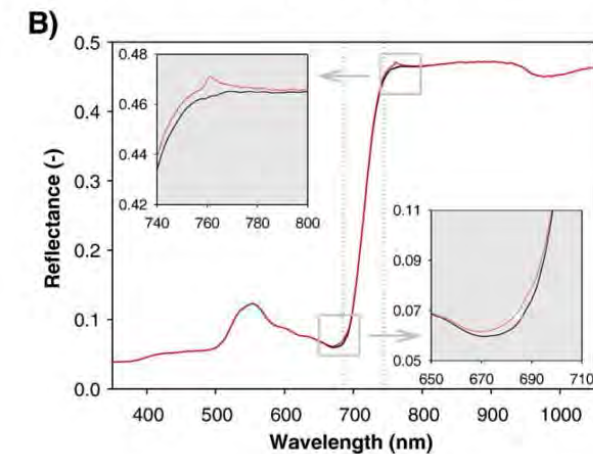
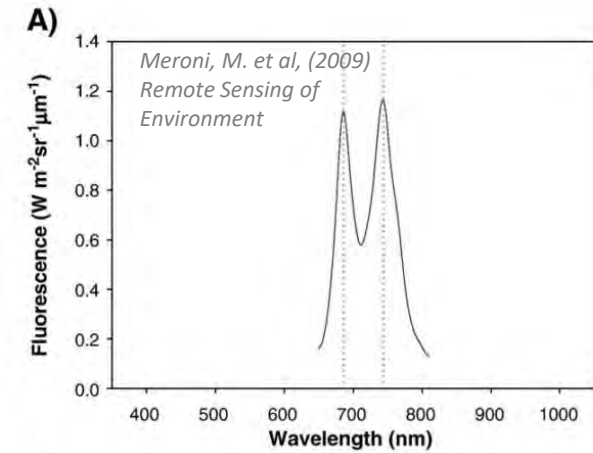
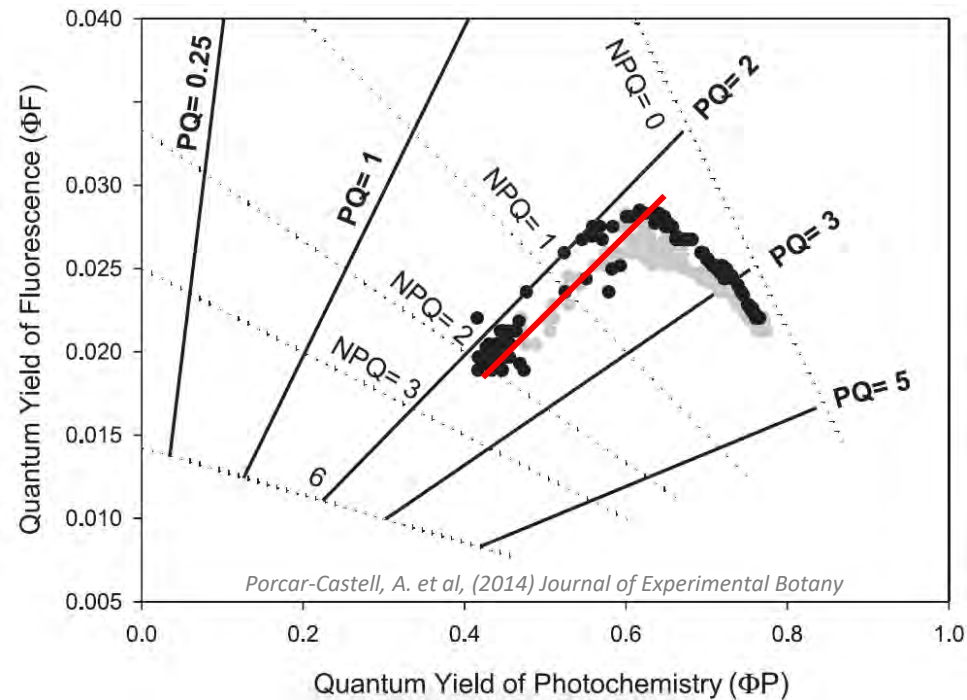
## Oxidative damage



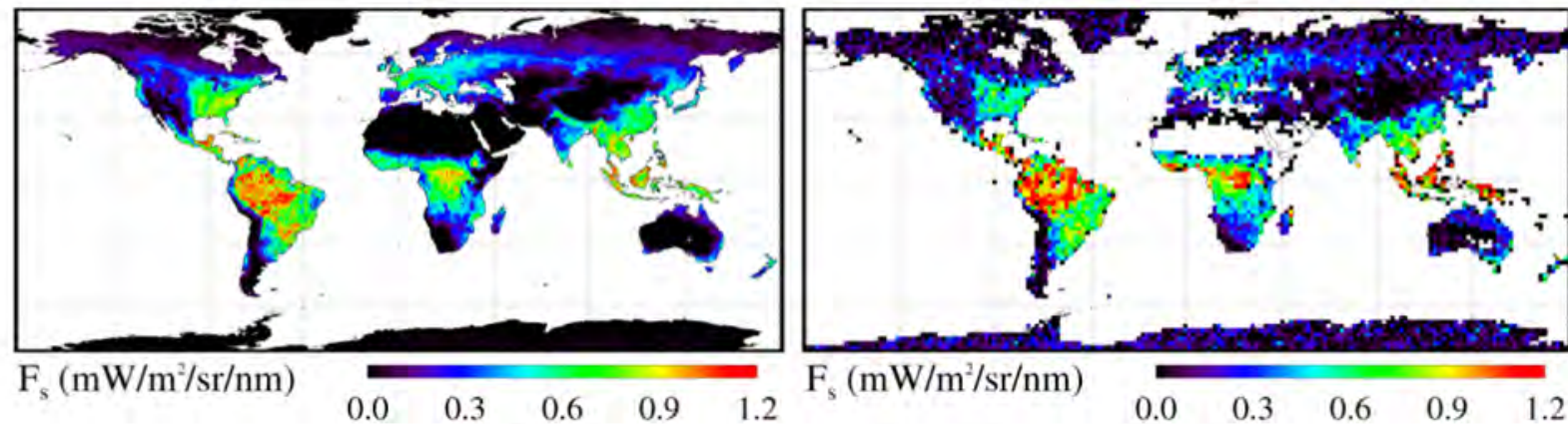
# The challenge of retrieving SIF from satellite RS

Only 1-5% of the reflected signal !!!

Fluorescence is proportional to photosynthesis only in some conditions



Several global datasets have appeared



(c) GOME-2 (left) and GOSAT (right), Annual 2009

Potentially useful, even if only a better 'green APAR'

Coarse spatial resolution: 0.5 degrees (but downscaled product exists)

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## Other avenues worth exploring...

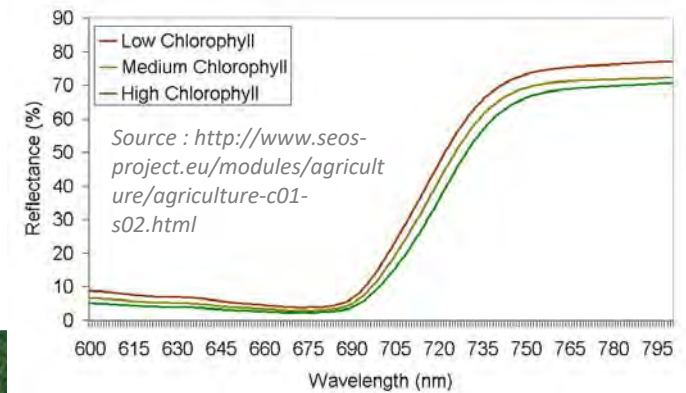
**NIRv:** Index multiplying NIR times NDVI seems to provide high correlation with SIF/GPP (*Badgley, Field, Berry, Sci. Adv. 2017*), advantage of having longer archive & higher res.

**Photochemical Reflectance Index (PRI):** Normalized difference between leaf reflectance at 531 nm and a reference wavelength (~550 nm) (*Gamon et al. 1992*)

- ⇒ Related to xanthophyll cycle
- ⇒ can serve as proxy for LUE

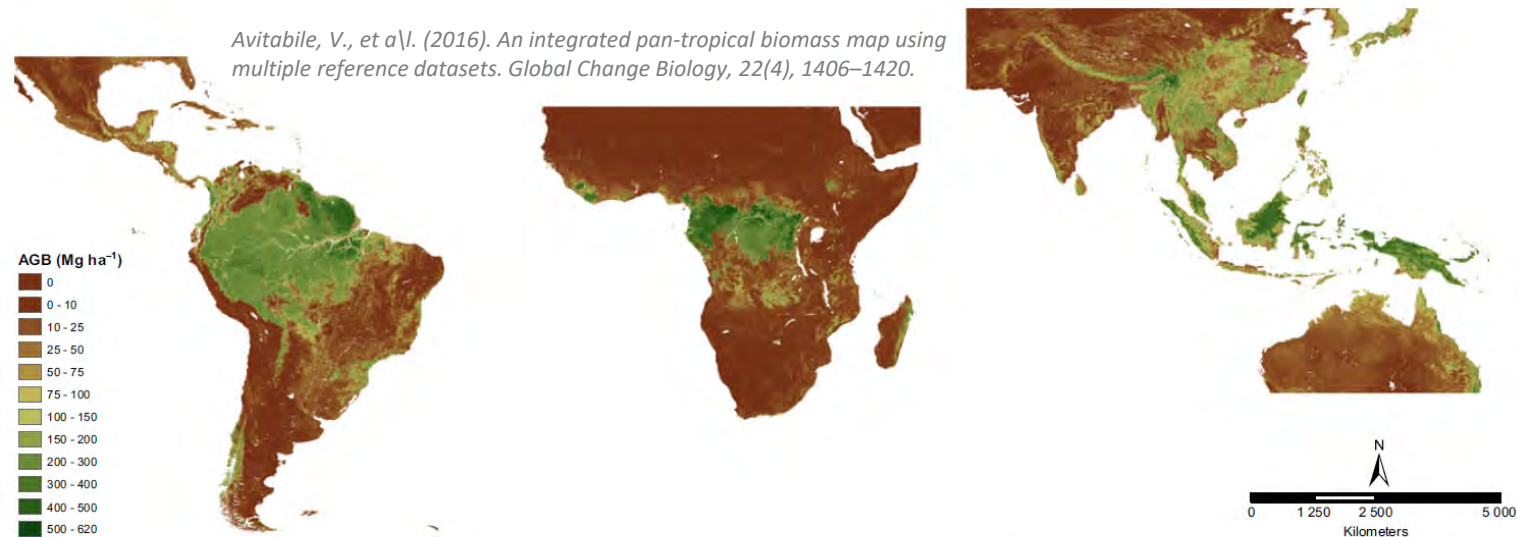
**Bands in the RED EDGE:** region of rapid change in reflectance of vegetation between red and near infrared (690-730 nm)

- ⇒ recognized as key for improving chlorophyll retrieval
- ⇒ Sentinel-2 has 2 bands in the red edge



## (Above-Ground) Biomass [AGB] from space

- Requires space-borne lidar (GLAS) to estimate AGB from tree height
- Assume global or continental allometric relationships (AGB varies only with stand height)
- Fusion of 2 maps along with ground AGB estimates:



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Fig. 3 Fused map, representing the distribution of live woody aboveground biomass (AGB) for all land cover types at 1-km resolution for the tropical region.

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## VOD – Vegetation Optical Depth

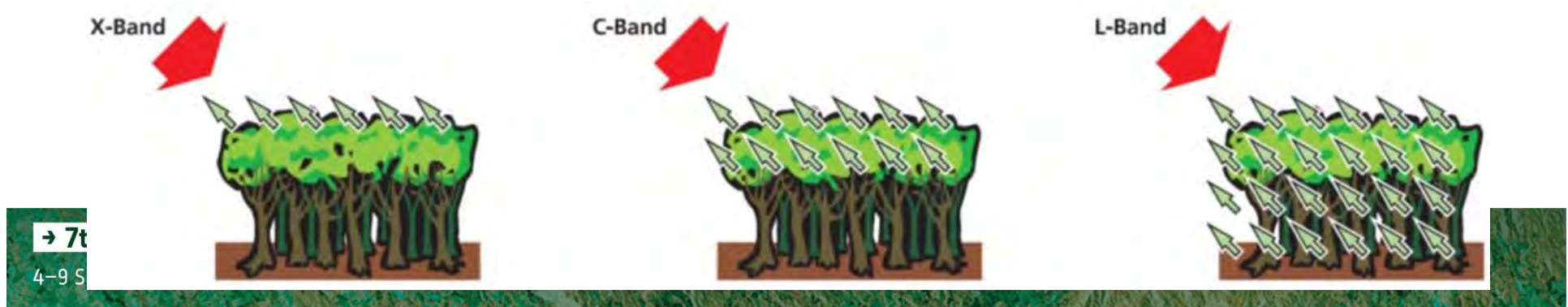
Passive microwave sensors are used for estimating soil moisture

But they need to 'model' and estimate 'noise' from the vegetation above

This 'noise' is in fact useful to relate to wet canopy structure

Not all bands penetrate as much...

*Source: [http://www.dlr.de/hr/en/desktopdefault.aspx/tabid-8113/14171\\_read-35852/](http://www.dlr.de/hr/en/desktopdefault.aspx/tabid-8113/14171_read-35852/)*



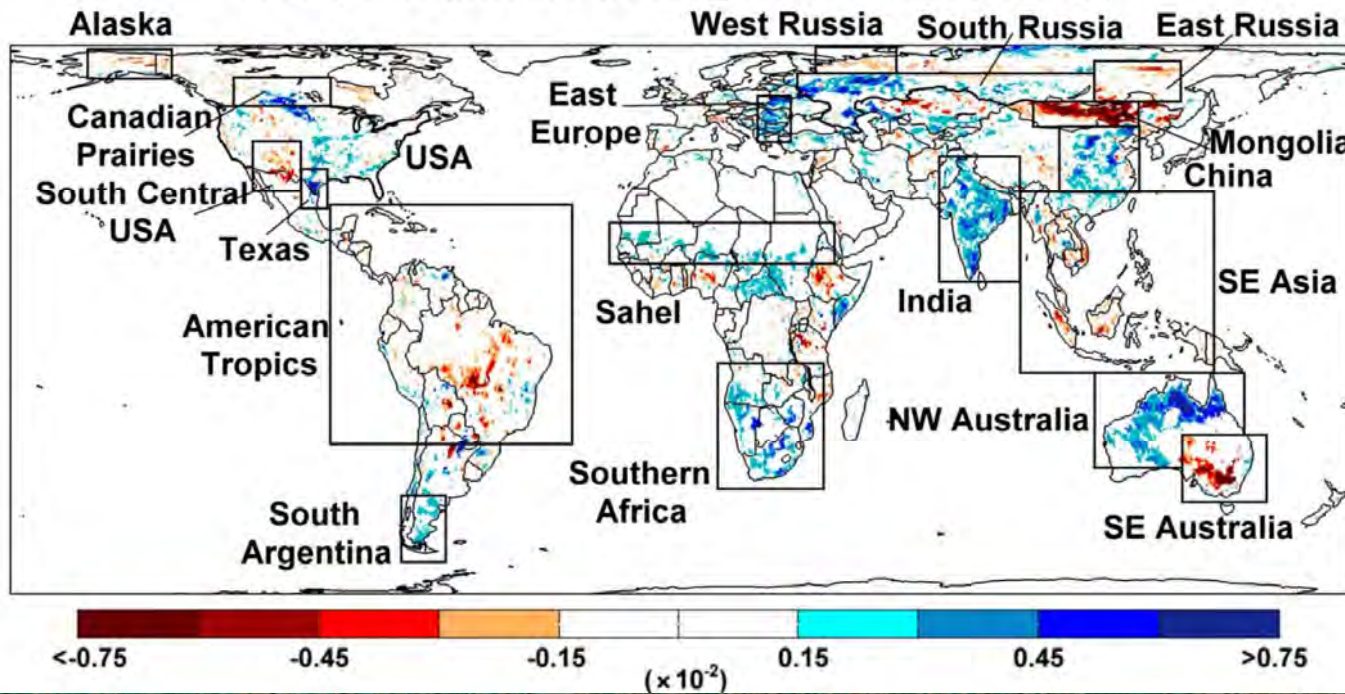


## VOD from C-band

Long time series of above-ground biomass change estimated from VOD

But arguably does not penetrate enough...

**Trends in annual average VOD (1988-2008)**

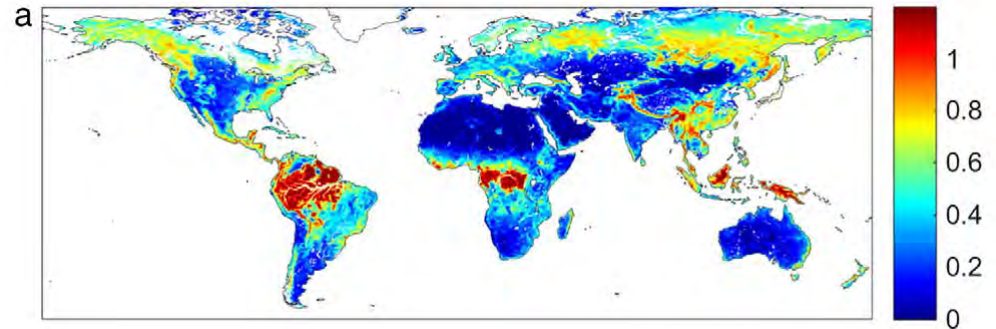
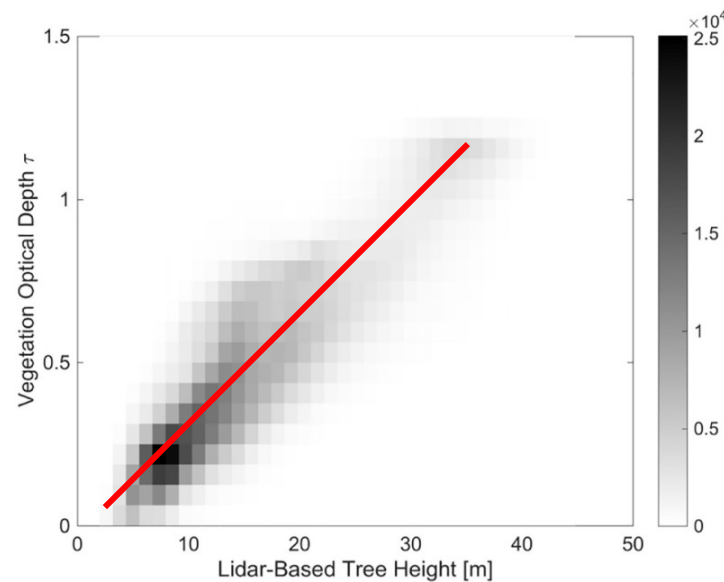


*Liu, Y.Y., et al (2013), Global Ecology and Biogeography*

## VOD from L-band sensors (SMOS, SMAP)

Show some promise of better results.

But time series are very short (couple of years)...



Konings, A. G., Piles, M., Das, N., & Entekhabi, D. (2017). L-band vegetation optical depth and effective scattering albedo estimation from SMAP. *Remote Sensing of Environment*, 198, 460–470. <https://doi.org/10.1016/j.rse.2017.06.037>

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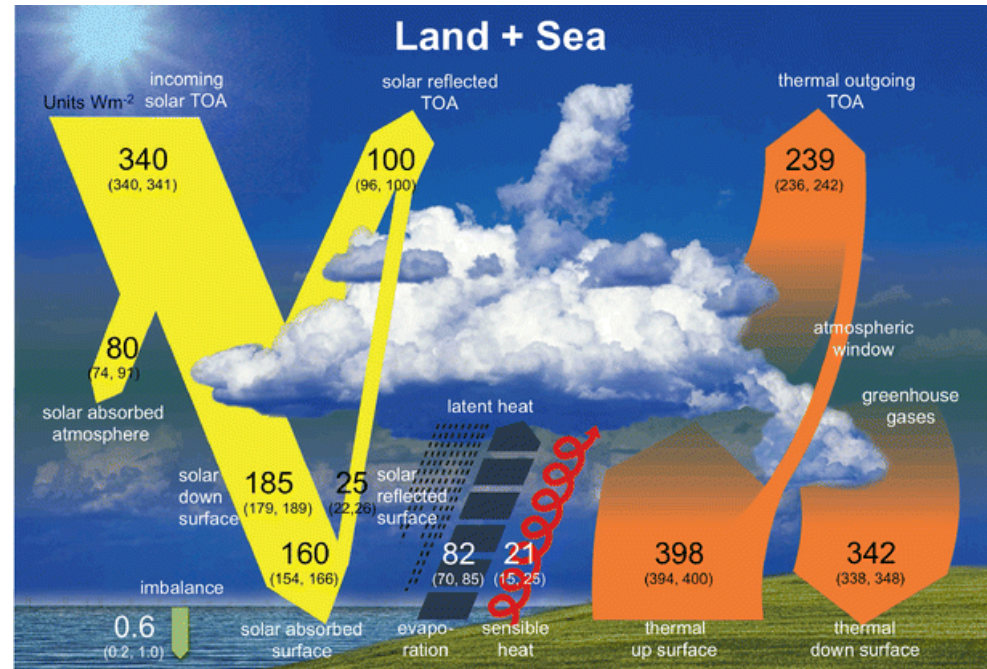
# The surface energy balance (SEB)

Radiative vs non-radiative  
(turbulent) fluxes

$$SW + LW = LE + H + G$$

$$SW_{\text{down}} - SW_{\text{up}} + LW_{\text{down}} - LW_{\text{up}} = LE + H + G$$

$$SW_{\text{down}} (1 - \alpha) + LW_{\text{down}} - LW_{\text{up}} = LE + H + G$$



Wild et al. 2015. *Clim. Dynamics*

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# Surface Albedo

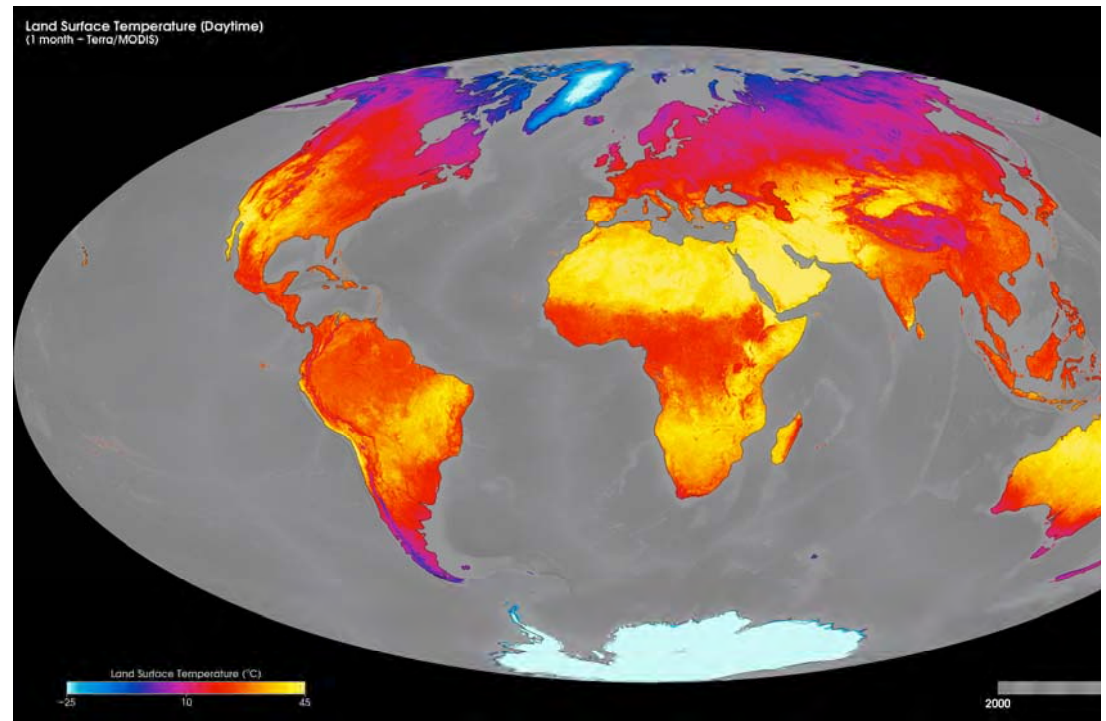
- How much shortwave radiation is reflected by the surface
- Obtained from multi-angular observations over a moving window
- Algorithms provide black-sky (directional) albedo and vs white-sky (diffuse) albedo
- Shortwave broadband or provided per spectral band (BRDF correction)



<http://modis.gsfc.nasa.gov>

# Land surface temperature (LST)

- Radiant temperature in Kelvin
- Variable resuming the consequence of the energy balance
- Linked to LW by Stephen-Boltzmann law
$$LW = \epsilon \sigma T^4$$
- Obtained from multi-angular observations of TIR reflectance
- Ill-posed problem inversion



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# Evapotranspiration or Latent Heat

Penman-Monteith equation

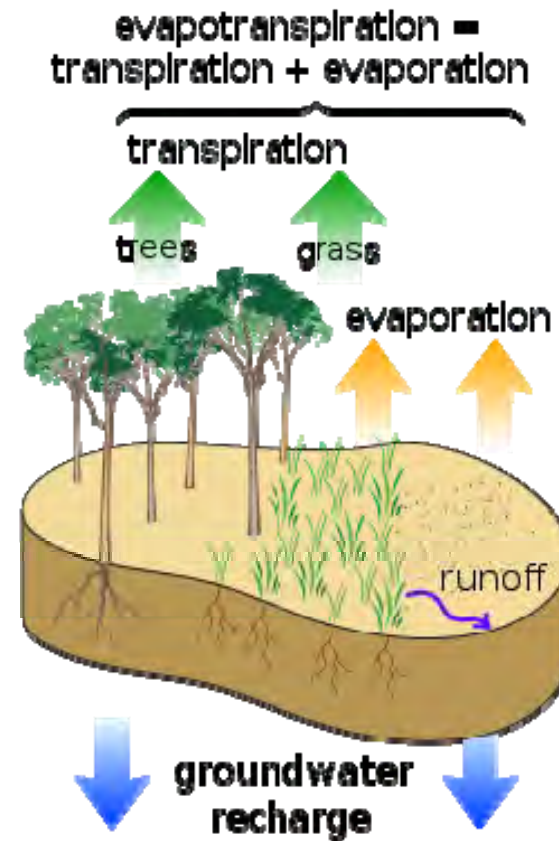
$$ET_o = \frac{0.408\Delta(R_n - G) + \gamma \frac{900}{T + 273} u_2 (e_s - e_a)}{\Delta + \gamma(1 + 0.34u_2)}$$

Priestly-Taylor equation

$$ET_p = \alpha \frac{\Delta}{\Delta + \gamma} (R_n - G) \quad [1]$$

Some existing products:

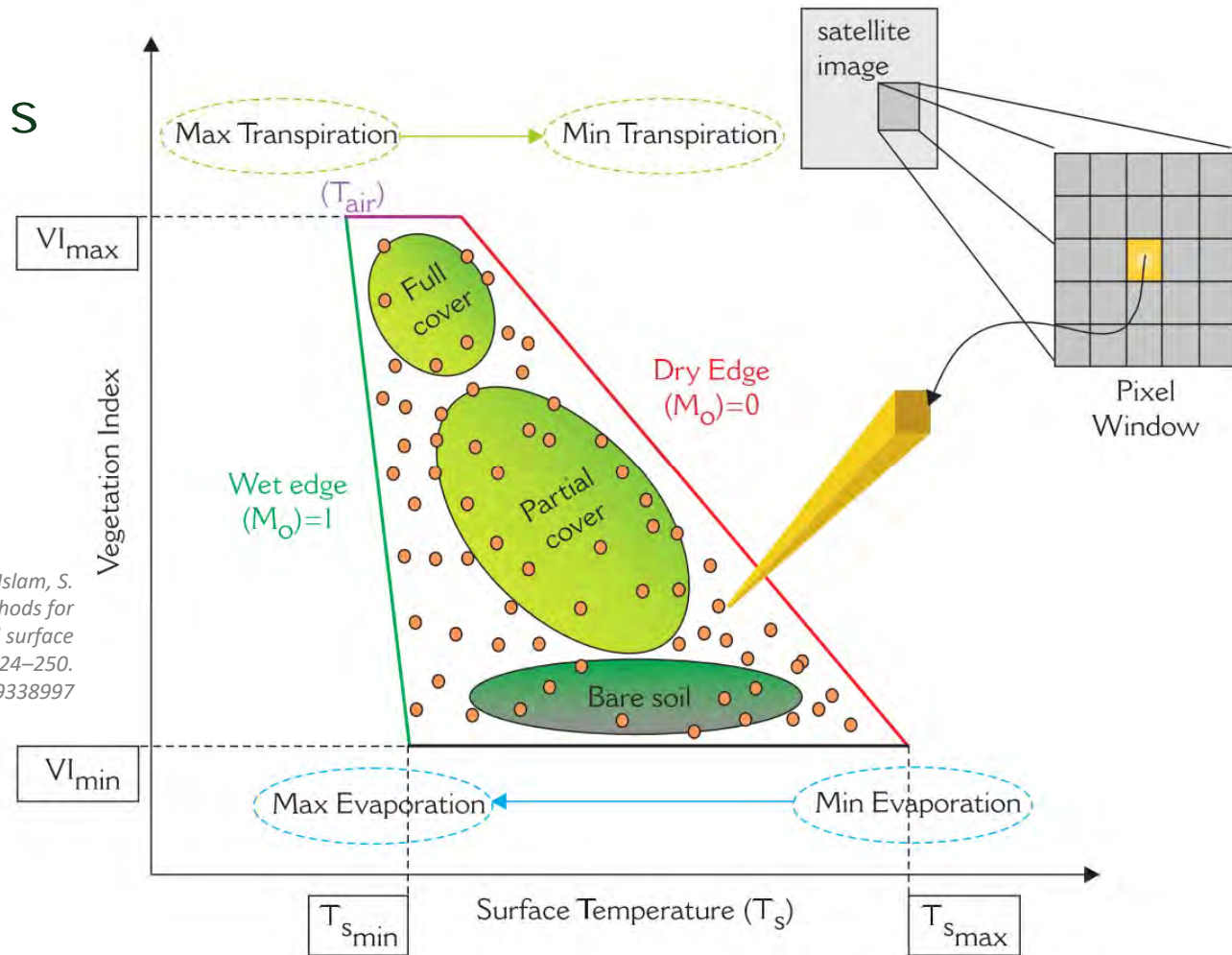
MOD16, 



"Surface water cycle" by Mwtowes - Own work. Licensed under CC BY 3.0 via Wikimedia Commons - [https://commons.wikimedia.org/wiki/File:Surface\\_water\\_cycle.svg#/media/File:Surface\\_water\\_cycle.svg](https://commons.wikimedia.org/wiki/File:Surface_water_cycle.svg#/media/File:Surface_water_cycle.svg)



# Estimating ET from LST + VIs



Petropoulos, G., Carlson, T. N., Wooster, M. J., & Islam, S. (2009). A review of  $T_s$ /VI remote sensing based methods for the retrieval of land surface energy fluxes and soil surface moisture. *Progress in Physical Geography*, 33(2), 224–250. doi:10.1177/0309133309338997

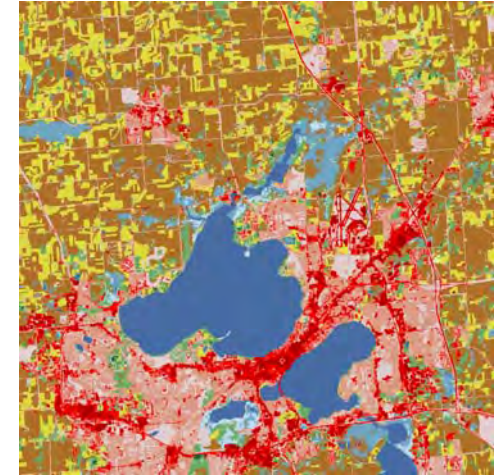
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# Description of the terrestrial surface

**Land Cover** = physical material at the surface of the earth  
(grass, asphalt, trees)

**Land Use** = description of how people *utilize* the land  
(wheat field, short-rotation coppice, ...)



**Plant Functional Types** = group of plants based on common features:  
structural (grasses/shrubs/trees),  
physiological (broadleaf/needleleaf)  
phenological (deciduous/evergreen)

**Optically distinguishable functional types** = based on detectable traits

*Ustin, S. L., & Gamon, J. A. (2010). Remote sensing of plant functional types. The New Phytologist, 186(4), 795–816.  
<https://doi.org/10.1111/j.1469-8137.2010.03284.x>*

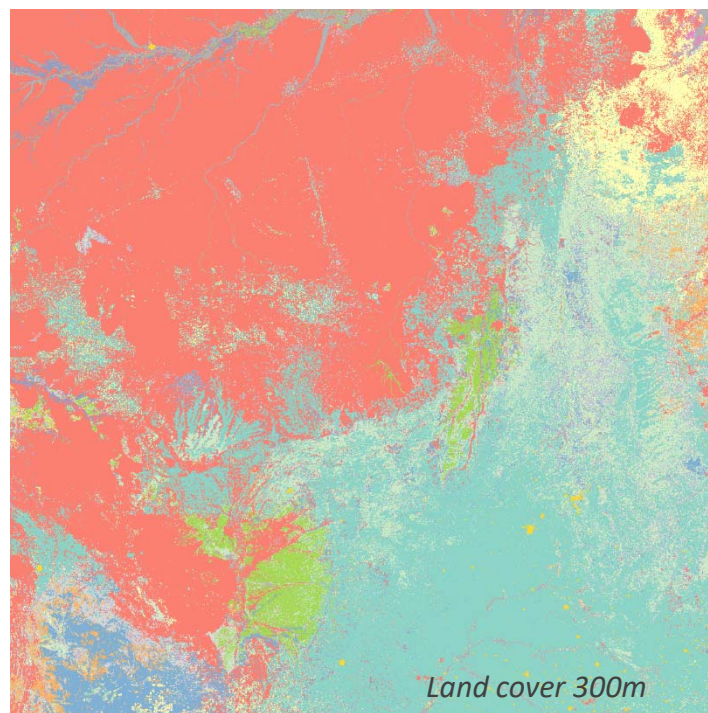
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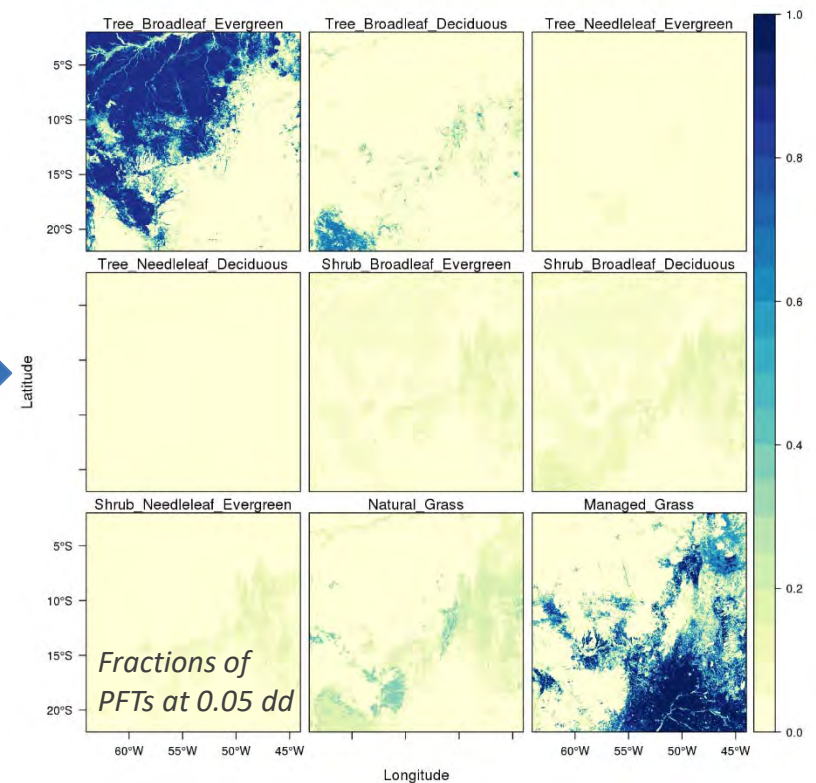




# From land cover to plant functional types (PFTs)



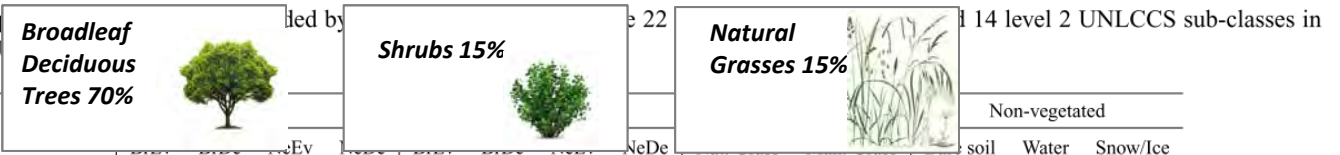
ESA CCI



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**Table 2.** Default land cover to plant functional type (PFT) mapping. Values in *italics*. The units are % coverage of each PFT per land cover class.



LCCS	UNLCCS Land Cover Class Description	Trees 70%						Non-vegetated			
Class				Ev			NeDe		soil	Water	Snow/Ice
10	Cropland, rainfed										
11	Herbaceous cover										
12	Tree or shrub cover										
20	Cropland, irrigated or post-flooding										
<div>Tree cover, broadleaf, deciduous, closed to open (&gt;15%)</div>		( $< 50\%$ )	5		5	5	15	100			
		( $< 50\%$ )	5		7.5	10	7.5	100			
		( $< 50\%$ )	90		5	5	25	50			
		( $< 50\%$ )	70		15	15	35	100			
		( $< 50\%$ )	70		15	15	35	100			
70	Tree cover, needleleaf, evergreen, closed to open ( $> 15\%$ )			70	5	5	5	15			10
71	Tree cover, needleleaf, evergreen, closed ( $> 40\%$ )			70	5	5	5	15			
72	Tree cover, needleleaf, evergreen, open (15–40%)			30	5	5	5	30			30
80	Tree cover, needleleaf, deciduous, closed to open ( $> 15\%$ )			70	5	5	5	15			
81	Tree cover, needleleaf, deciduous, closed ( $> 40\%$ )			70	5	5	5	15			
82	Tree cover, needleleaf, deciduous, open (15–40%)			30	5	5	5	30			30
90	Tree cover, mixed leaf type (broadleaf and needleleaf)		30	20	10	5	5	5	15		10
100	Mosaic tree and shrub ( $> 50\%$ )/herbaceous cover ( $< 50\%$ )	10	20	5	5	10	5	40			
110	Mosaic herbaceous cover ( $> 50\%$ )/tree and shrub ( $< 50\%$ )	5	10	5	5	10	5	60			
120	Shrubland				20	20	20	20			20
121	Shrubland evergreen				30	30	30	20			20
122	Shrubland deciduous							20			20
130											
140											
150											
152											
153											
160	Tree cover, flooded, fresh or brackish water	30	30					20			20
170	Tree cover, flooded, saline water	60			20						20
180	Shrub/herbaceous cover, flooded, fresh/saline/brackish water		5	10		10	5	40			30
190	Urban areas		2.5	2.5				15		75	5
200	Bare areas										100
201	Consolidated bare areas										100
202	Dispersed bare areas										100

But this transformation should not be necessary if we have 'optical functional types' or if we produce continuous maps of vegetation cover directly

# Climate modelling: some clarifications...

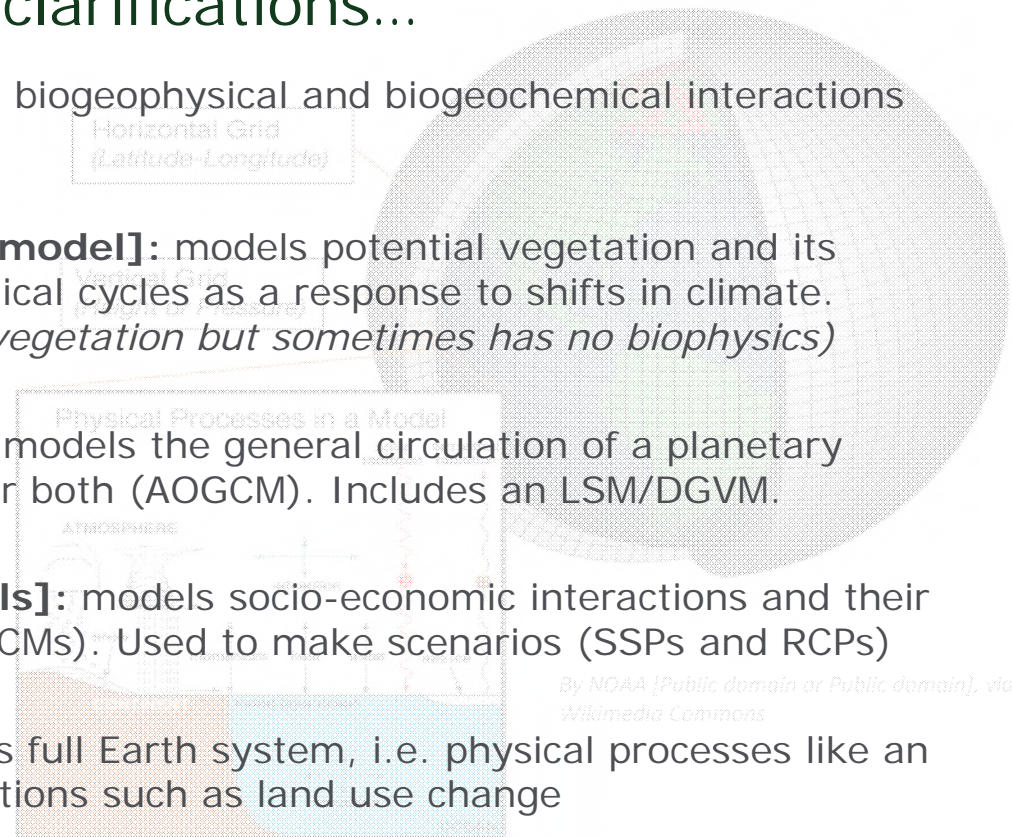
**LSMs [Land surface models]:** models biogeophysical and biogeochemical interactions between the land and the atmosphere.

**DGVMs [Dynamic global vegetation model]:** models potential vegetation and its associated biogeochemical and hydrological cycles as a response to shifts in climate. *(often equivalent to LSM with dynamic vegetation but sometimes has no biophysics)*

**GCMs [General Circulation Models]:** models the general circulation of a planetary atmosphere (AGCM) or ocean (OGCM) or both (AOGCM). Includes an LSM/DGVM.

**IAMs [Integrated Assessment Models]:** models socio-economic interactions and their responses to forced climate (eg. from GCMs). Used to make scenarios (SSPs and RCPs)

**ESMs [Earth System models]:** models full Earth system, i.e. physical processes like an AOGCM but also includes human interactions such as land use change



By NOAA [Public domain or Public domain], via  
Wikimedia Commons



# Role of remote sensing for climate science

- Full potential has been neglected to some extent by climate community
- no direct observation of carbon
- Requires archive... often too short...
- Cannot go in the future under difference scenarios
- Synoptic coverage → can bridge the gap between *in situ* and models



Thank you for your attention...

*Contact:*

gregory.duveiller@ec.europa.eu

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