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New developments for improved carbon flux estimation by remote sensing

Lars Eklundh

Dept. of Physical Geography and Ecosystem Science, Lund University



Contents

- Spatial heterogeneity and carbon modeling
- New vegetation index: Plant Phenology Index (PPI)
- Data modeling by spatio-temporal filtering

COST OPTIMISE Workshop, Madrid, 30 Mar.-1 Apr. 2015



Managing the gap between flux towers and global modeling

Flux tower infrastructure
Fluxnet, ICOS, etc.

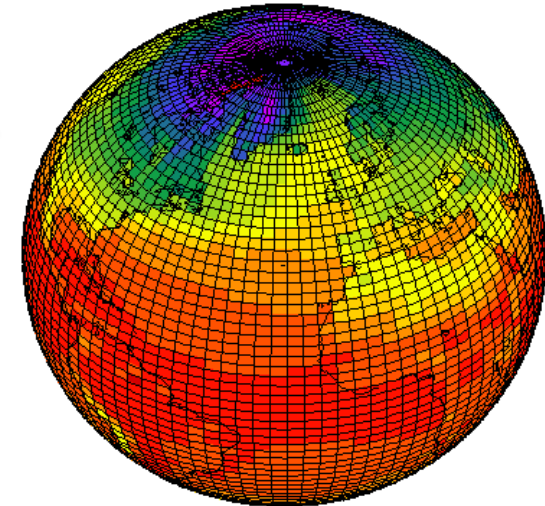


Norunda ICOS flux tower
Source: www.necc.nu



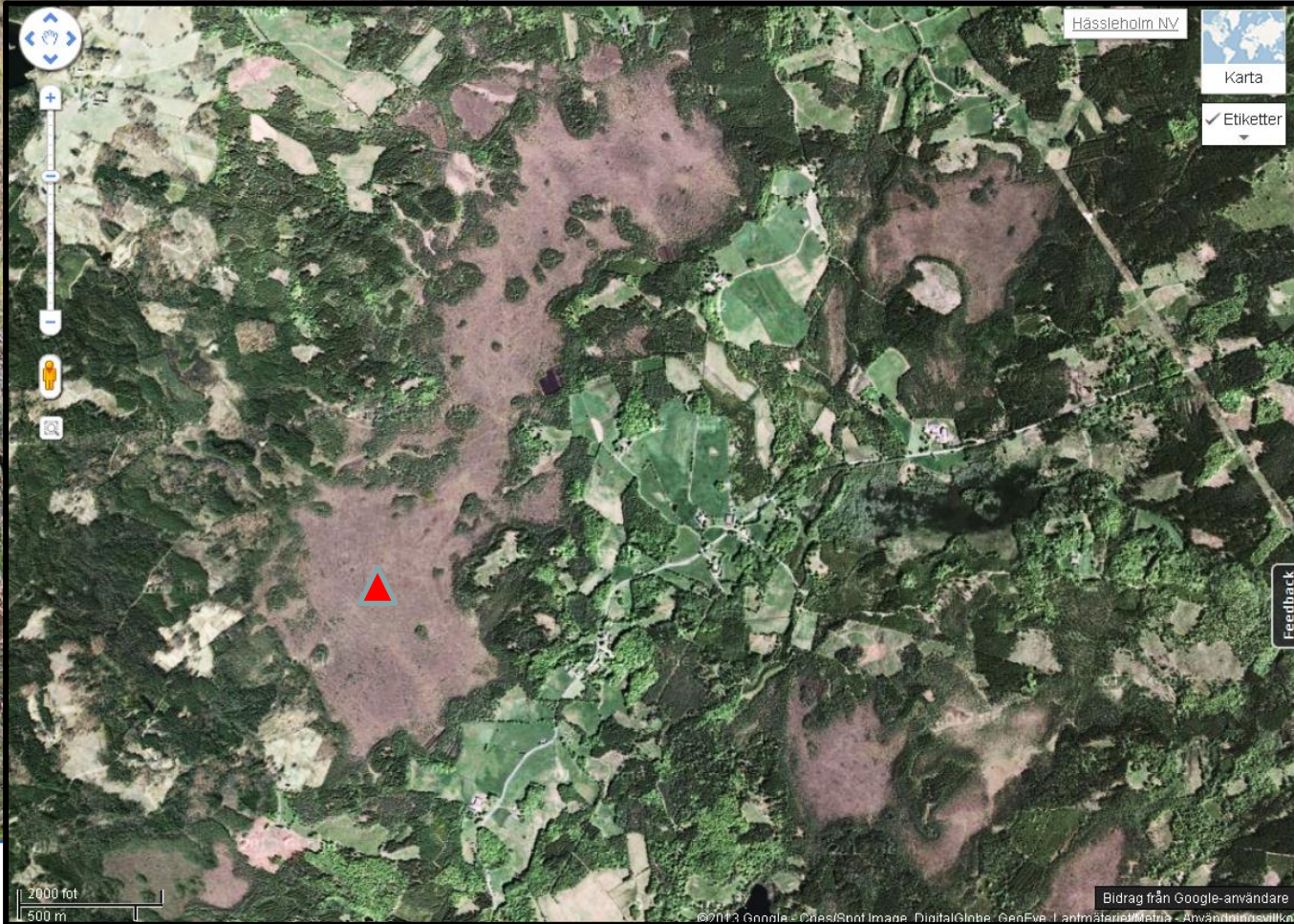
SPOT satellite image 2010-06-28. Source: Lantmäteriet / Saccess

Carbon Cycle Data
Assimilation Models
(CCDAS)



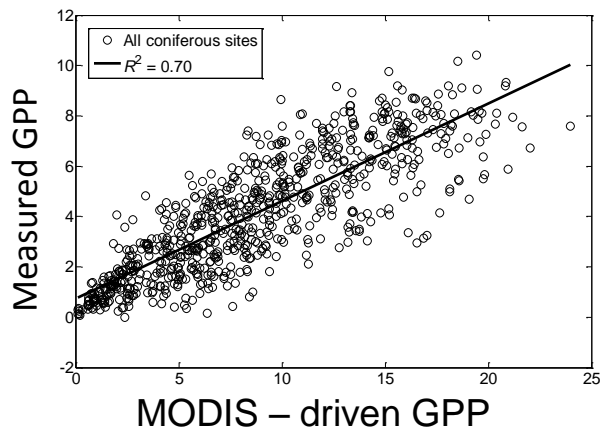
Source: <http://www.geos.ed.ac.uk>

The image is a composite of three panels. The top panel is a photograph of a grassy hillside with a single tree. The middle panel is a photograph of a grassy hillside with a single tree. The bottom panel is a satellite map of the same area, showing a red triangle marker on a cleared area.



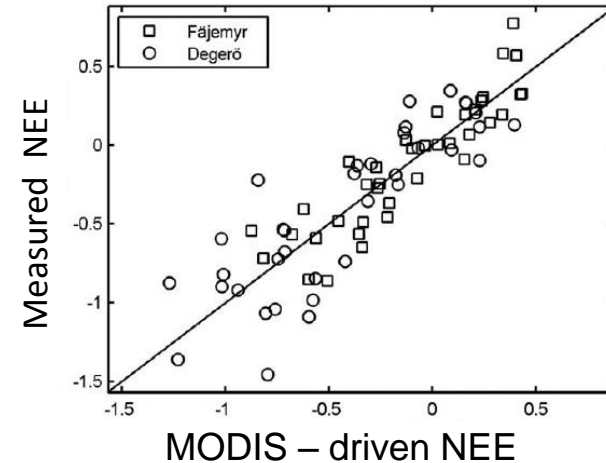
Temporal satellite data can mimick carbon variations

GPP of boreal forest stands



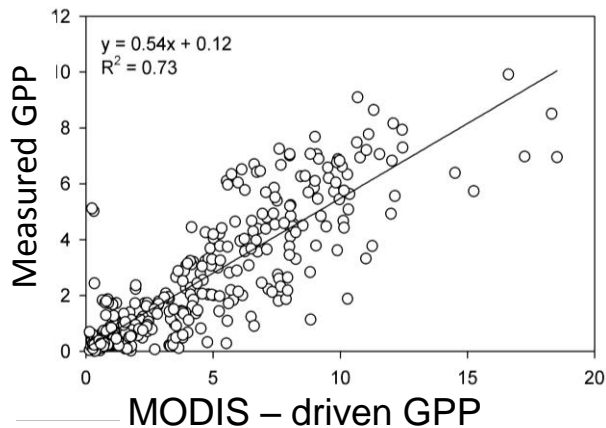
Schubert et al. , 2012, *Rem Sens Env.*, 126.

NEE of two peatlands



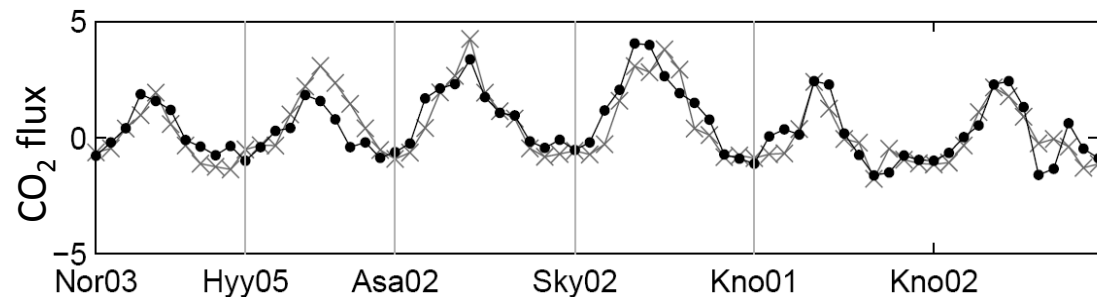
Schubert et al. , 2010, *Rem. Sens. Env.*, 114.

GPP of African drylands



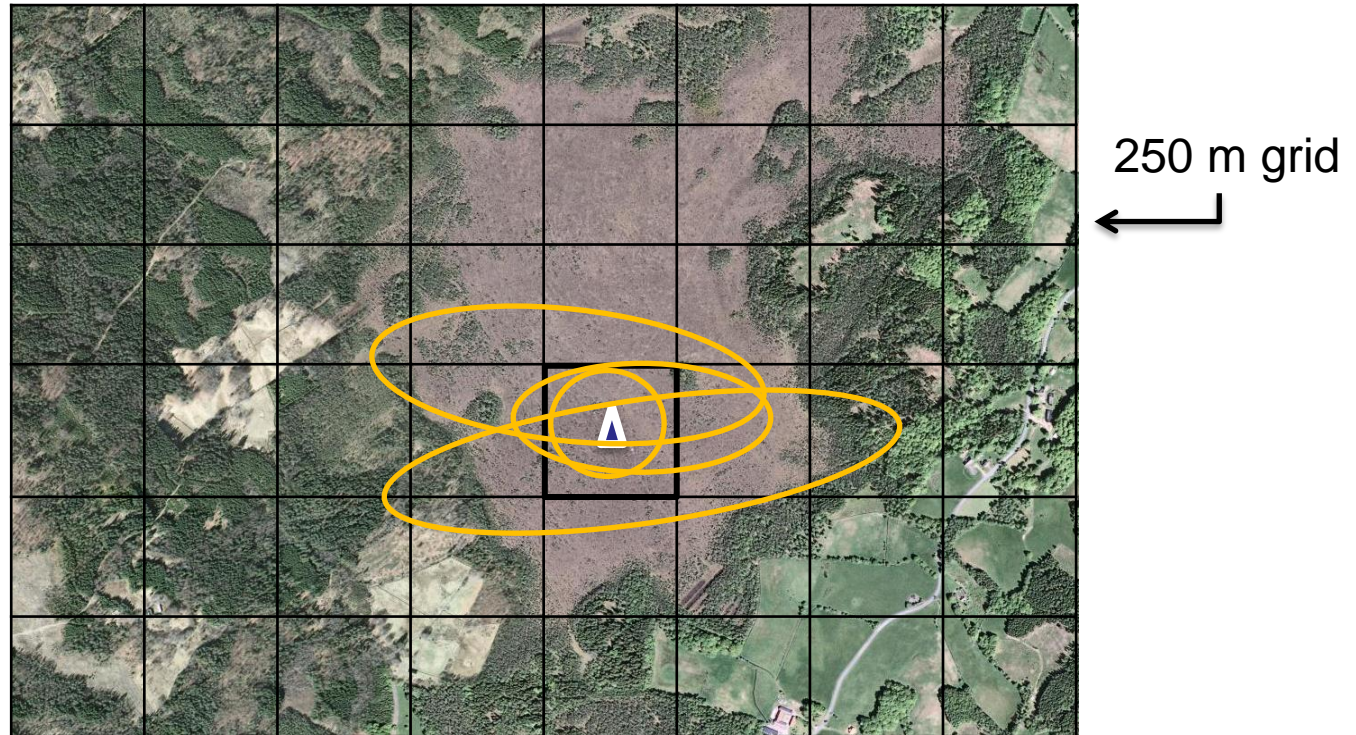
Sjöström et al., 2009, *Biogeosciences*, 6.
Sjöström et al., 2011, *Rem Sens Env*, 115.

Seasonal forest NEE



Olofsson et al., 2008, *Biogeosciences*, 5.

Coarse resolution satellite geometric uncertainty

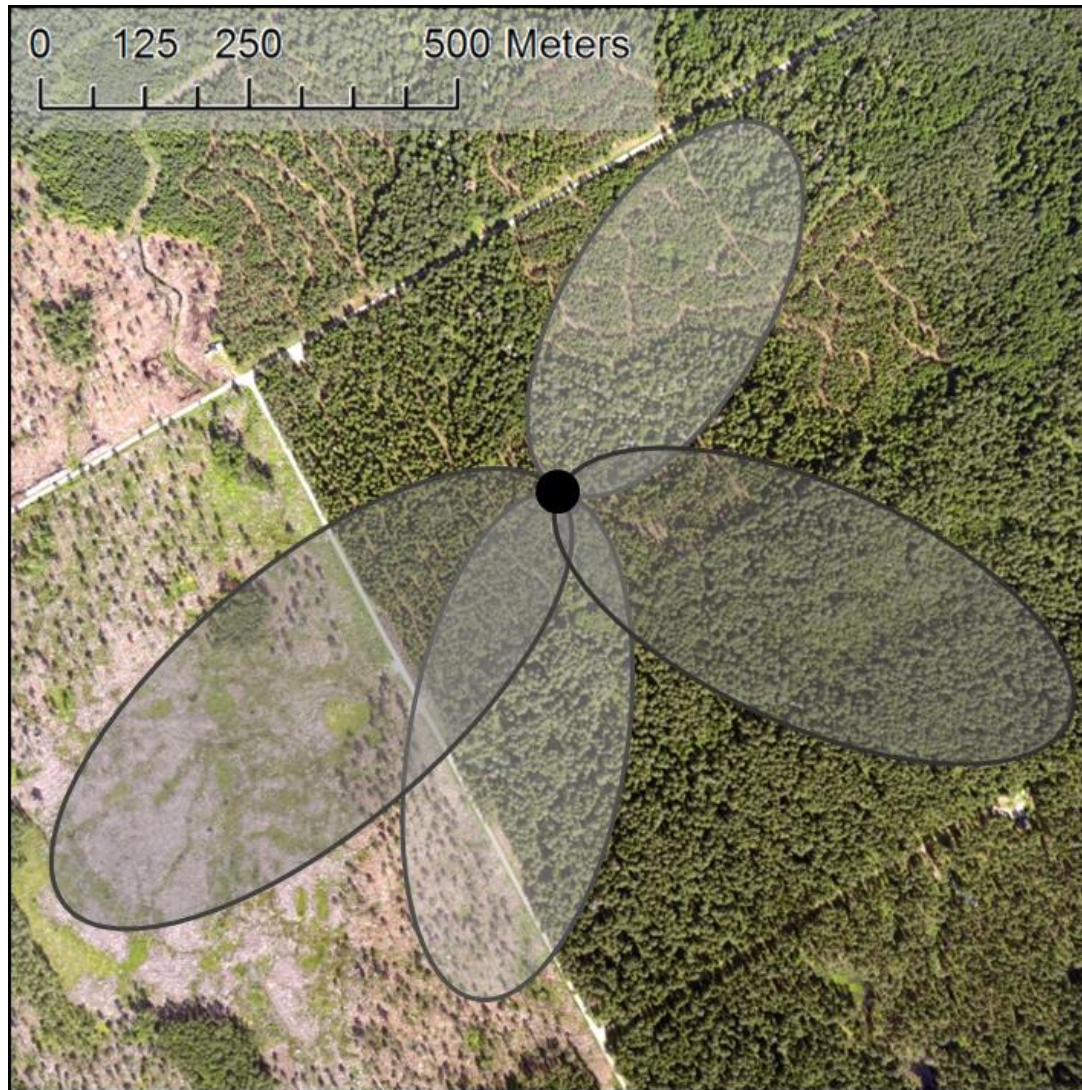


(Note: simulated data...)

Map: hitta.se

- Clouds
- Most important other problems:
 - Atmospheric disturbances
 - Angular effects
 - Uncertainty of target processes

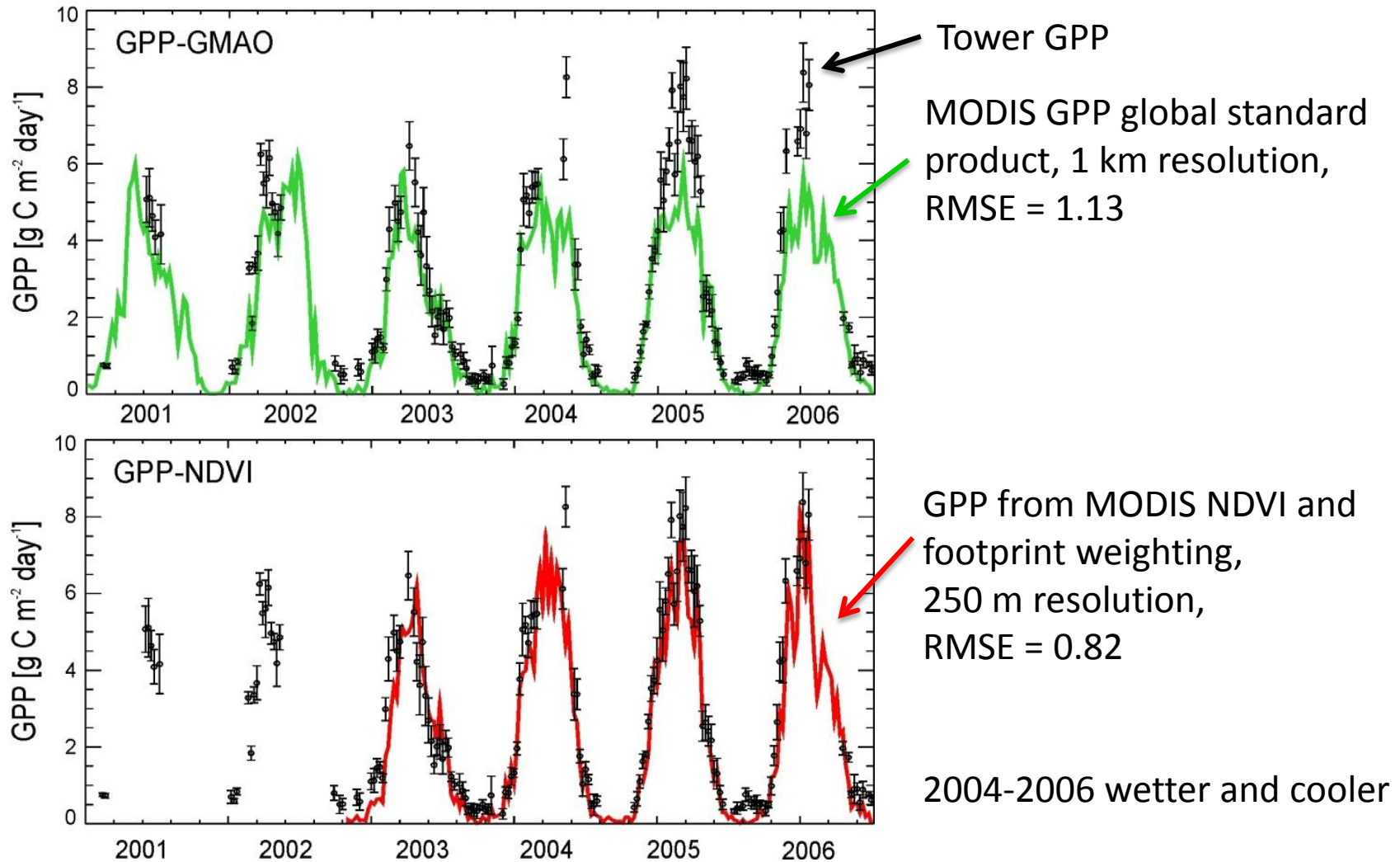
Flux footprint variations



Half-hour flux footprints from the ICOS Norunda mast (black dot)

Credit: Natascha Kljun

Matching tall tower GPP with footprint-derived NDVI



Gelybo, Barcza, Kern, Kljun, 2013, AFM 174-175, 43-53



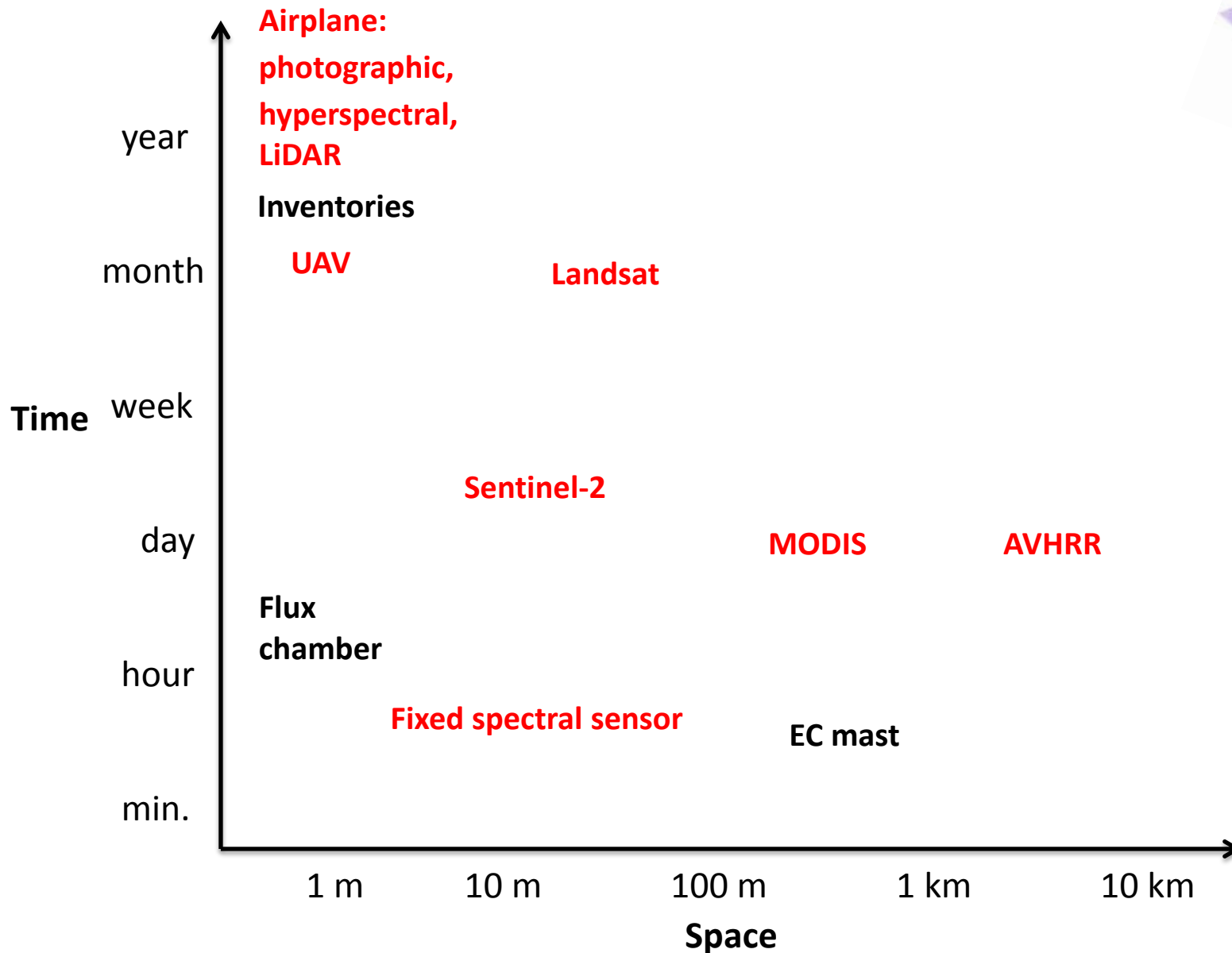
Some vegetation parameters for upscaling

- Phenology, interannual foliage variations
- fAPAR (radiation absorption)
- Leaf area index
- Canopy light response
- Canopy height
- Vegetation structure
- Chlorophyll concentration
- Canopy moisture
- Canopy surface temperature
- Biomass, forest volume
- Light use efficiency
- Soil moisture
- Chlorophyll fluorescence
- Vegetation nutrient level
- Species

Challenge

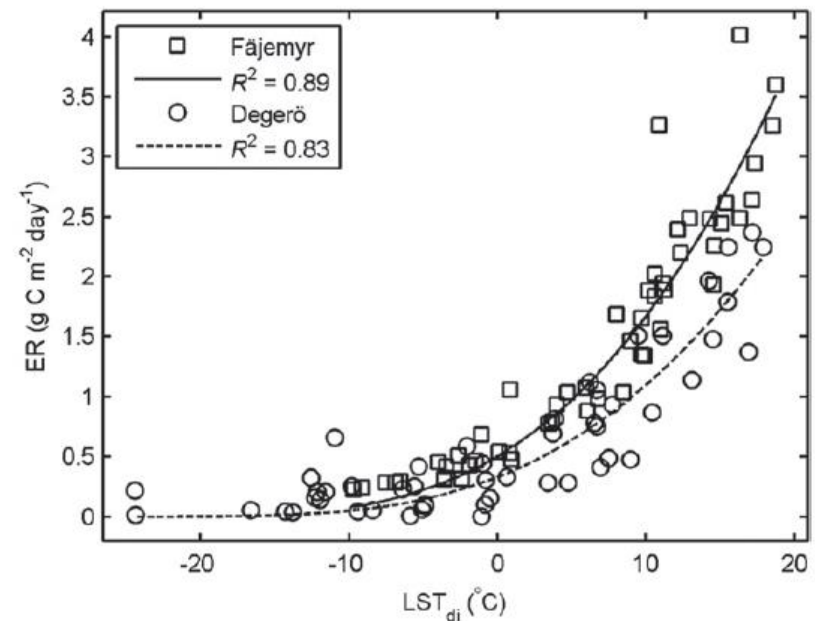


Time-space graph of measurements



Some necessary advances for improved process understanding and upscaling

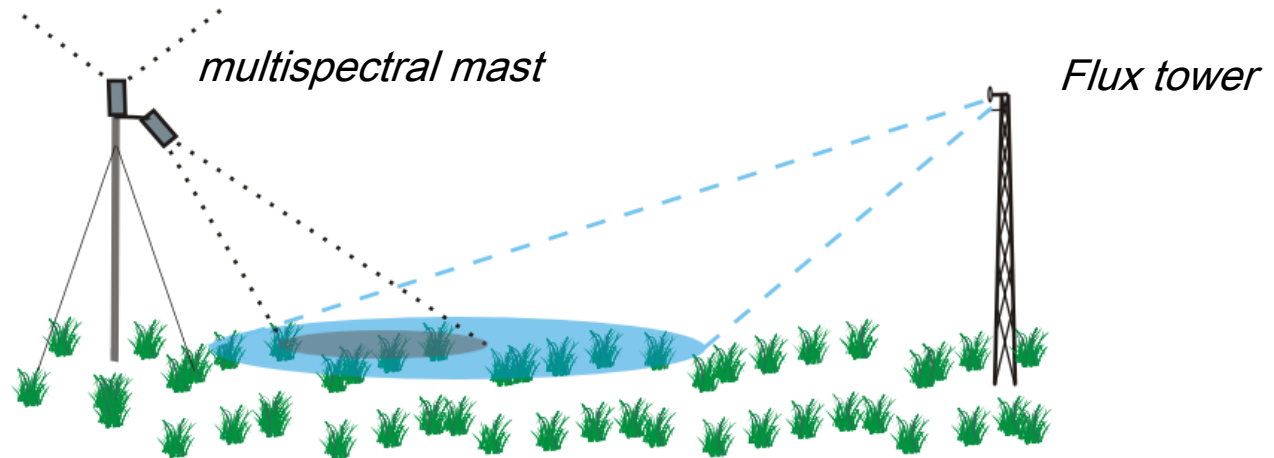
- Improved sampling at different spatial scales
 - Fixed sensors
 - UAV
 - New satellite data: Sentinel-2
- Better understanding of spectral data and vegetation indices
 - PRI: photochemical reflectance index vs. light use efficiency
 - SWIR radiation: canopy moisture
 - Land surface temperature
 - PPI – plant phenology index
- Better data processing methods for handling noise and missing data
 - Bayesian methods for data integration
 - Spatio-temporal data modeling



Ecosystem respiration estimated from MODIS land surface temperature. Schubert et al. (2010) Rem. Sens. Env, 114.

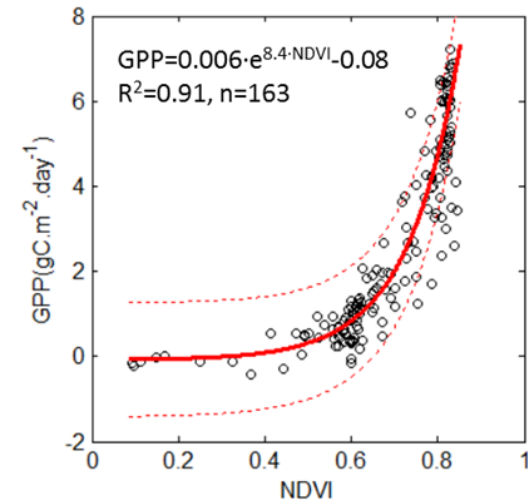
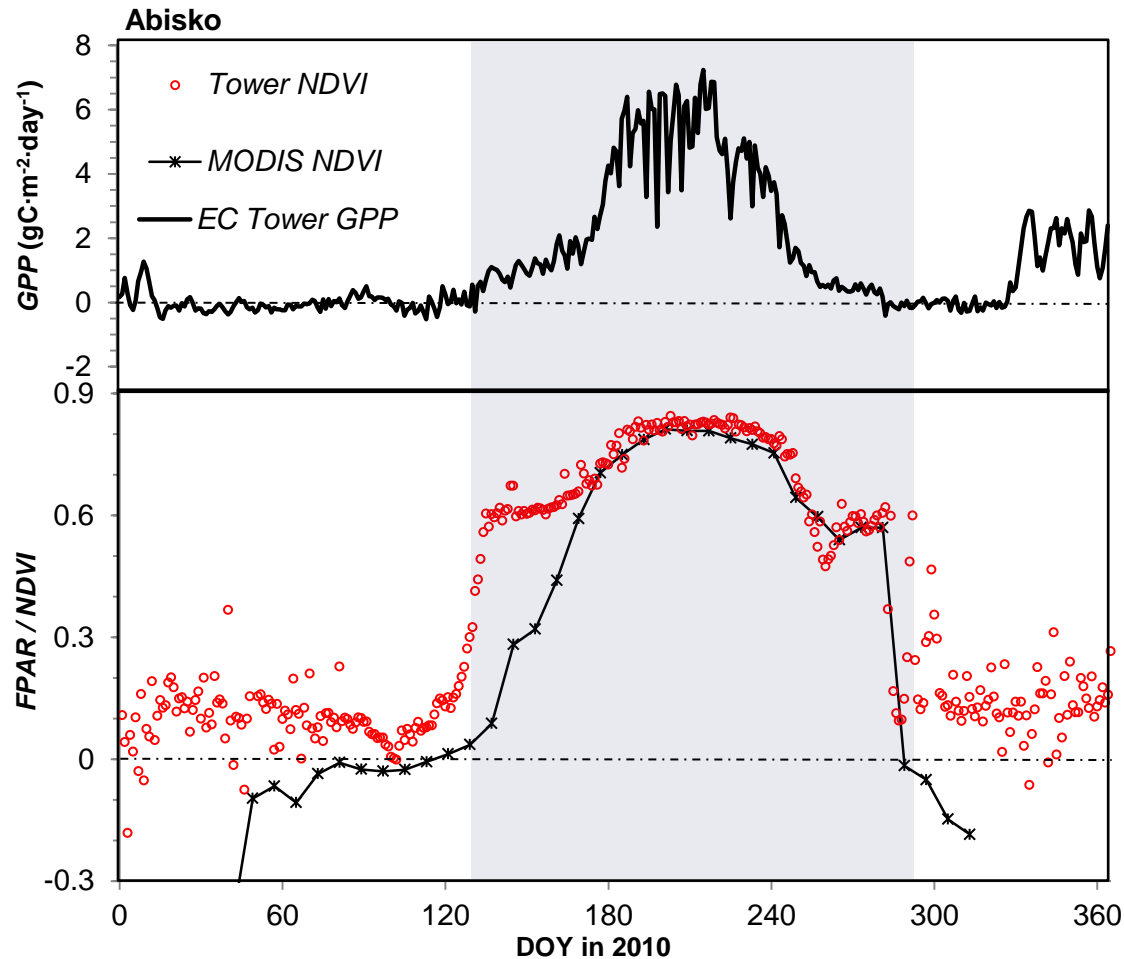


Point measurements for high-frequency and long-term data sampling

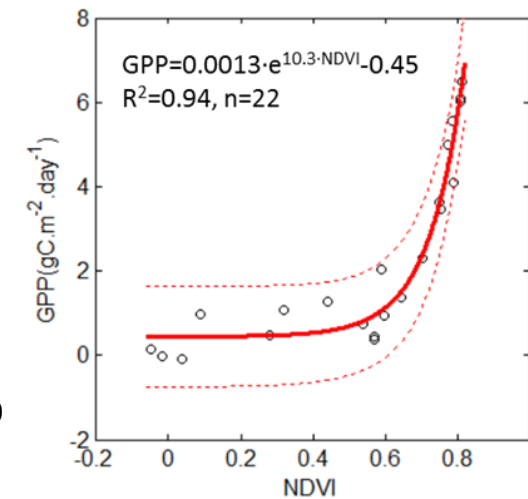


Seasonal GPP and NDVI from mast

Abisko birch forest 2010



Ground
NDVI

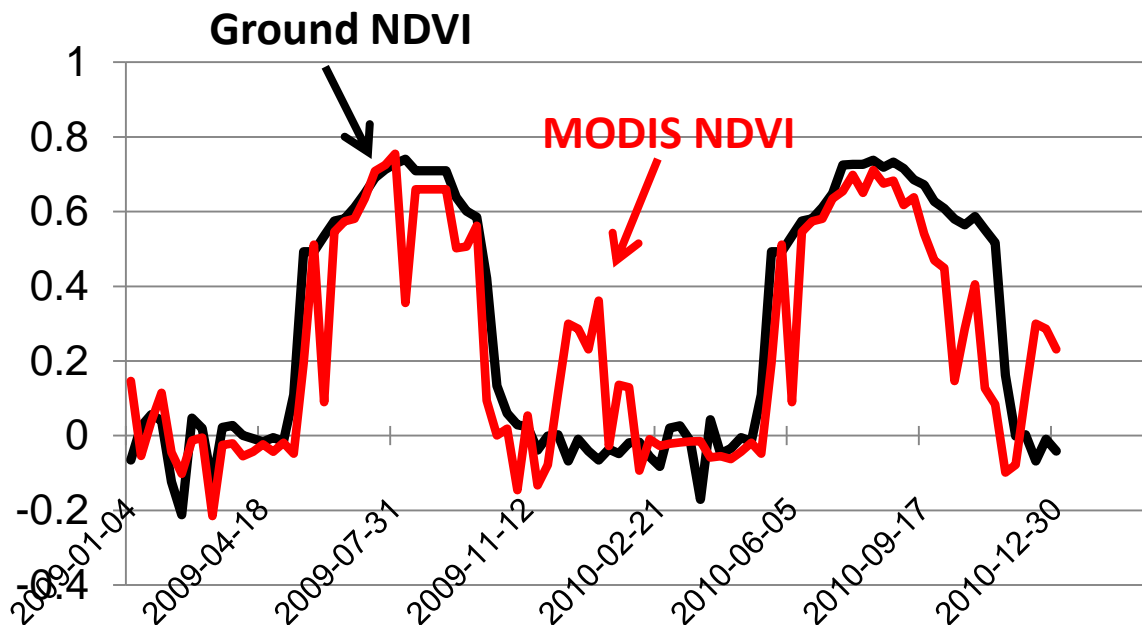


MODIS
MOD43A2
NBAR
NDVI

Modified from Eklundh et al. (2011)

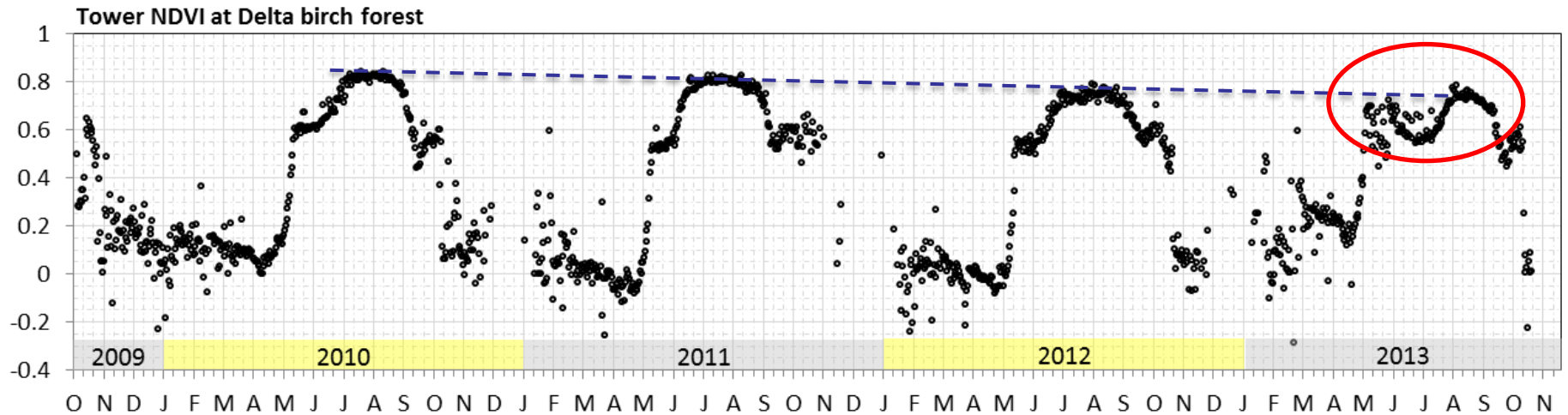


Reference data for validating satellite data



Method	Raw MODIS	Ass. Gauss	Sav. Gol.	Doub. log.
Default TIMESAT	0.1724	0.1399	0.1377	0.1437
Spike removal	0.1724	0.1285	0.1232	0.1289
Envelope iterations	0.1724	0.1148	0.1059	0.1165

Slow and sudden environmental impacts



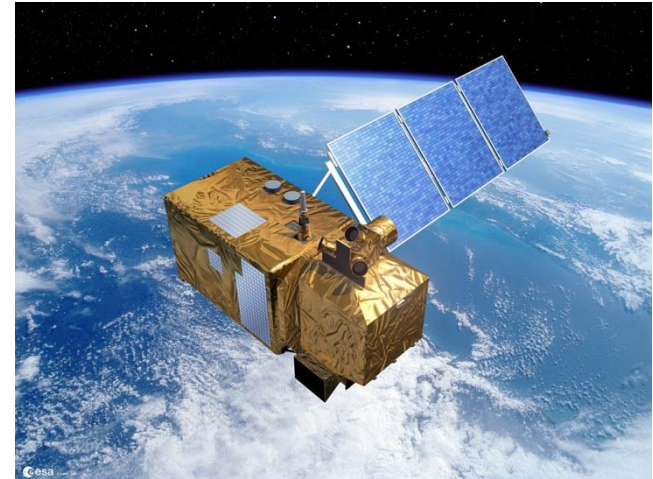
1. Tendency for NDVI to decrease 2010 – 2013
2. 2013 season strongly affected by insect attack



New satellites

Sentinel-2

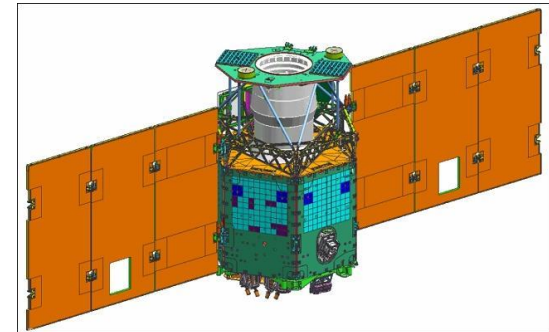
- Launch June 2015
- 13 spectral bands (Vis-NIR-TIR)
- 10-60 m ground resolution
- 3-5 days temporal resolution



<http://spacefellowship.com>

Venµs

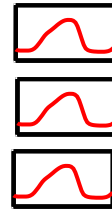
- Launch 2016
- 12 spectral bands (Vis-NIR)
- 10 m ground resolution
- 2 days temporal resolution
- Experimental – only 100 selected sites globally
- Lifetime 2.5 years



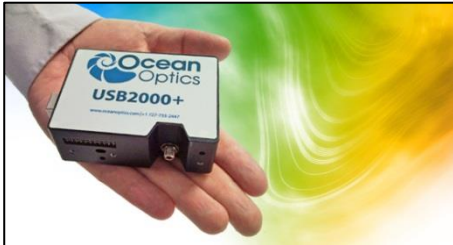
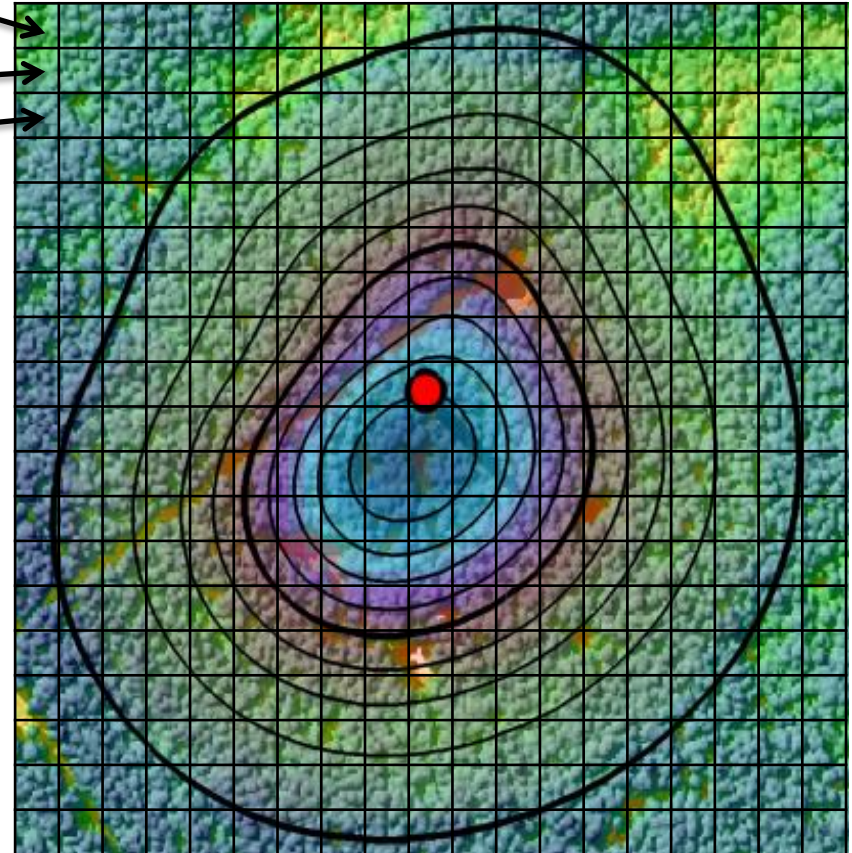
<https://directory.eoportal.org/>

Scaling within flux footprints: high-resolution dynamic data for weighting pixel contribution

Use fixed sensors, UAV, and new satellite data for creating better representation of the flux footprint variation



...



Source:
<http://www.pro-lite.co.uk/>



Source: <http://multirotor.co.za/>

PPI – Plant Phenology Index

Remote Sensing of Environment 152 (2014) 512–525



Contents lists available at ScienceDirect

Remote Sensing of Environment

journal homepage: www.elsevier.com/locate/rse



A physically based vegetation index for improved monitoring of plant phenology



Hongxiao Jin*, Lars Eklundh

Department of Physical Geography and Ecosystem Science, Lund University, Sölvegatan 12, SE-223 62 Lund, Sweden

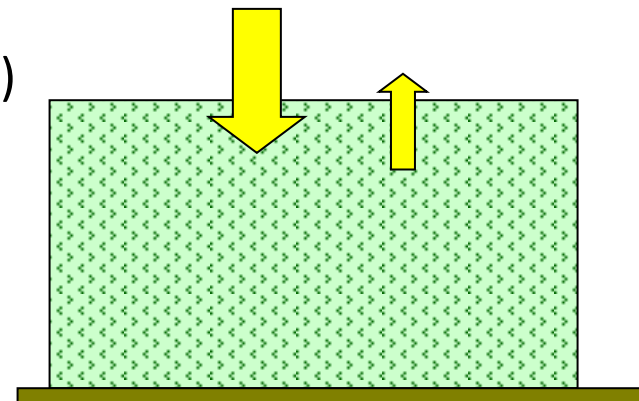
- Uses red and NIR reflectance
- Diffusive reflectance theory by Hapke (1993)

$$\rho_C \approx \rho_V + (\rho_S - \rho_V) \cdot e^{-k \cdot L},$$

Canopy
reflectance

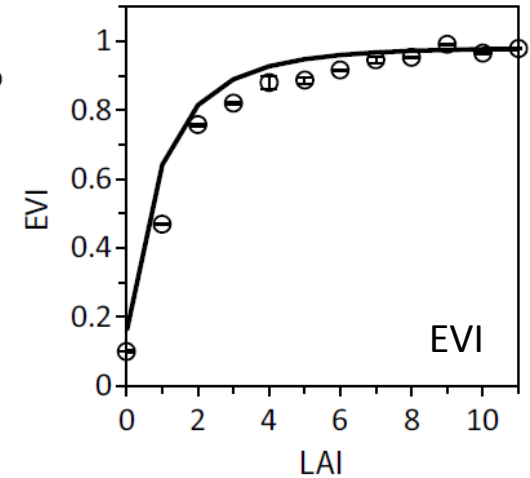
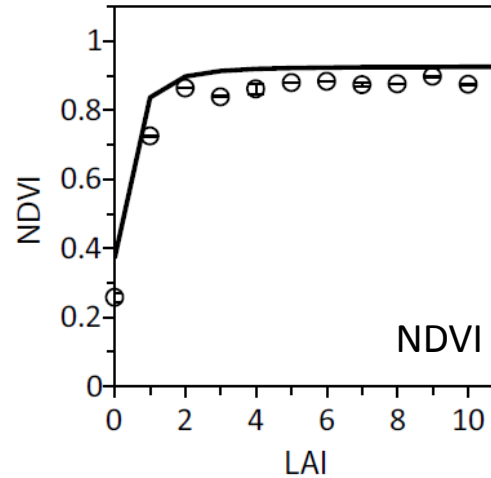
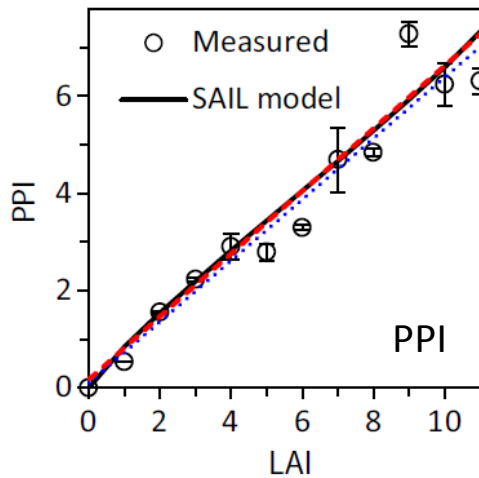
Soil vs.
veg. refl.

LAI

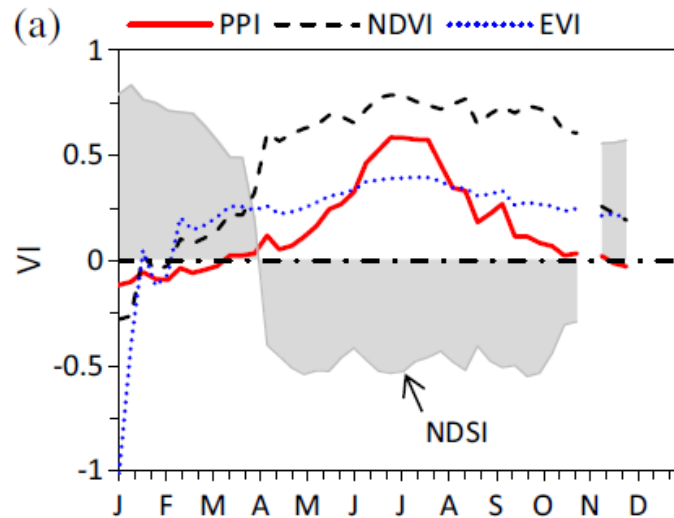


PPI - properties

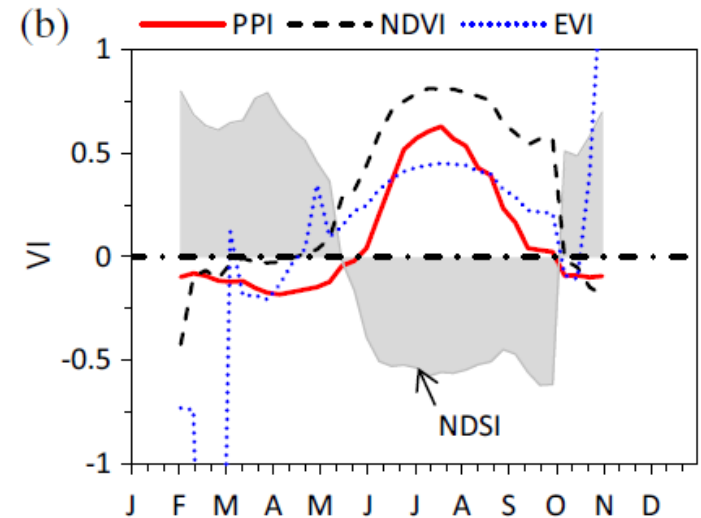
Linear
with green
LAI



Hyytiälä pine forest



Abisko birch forest

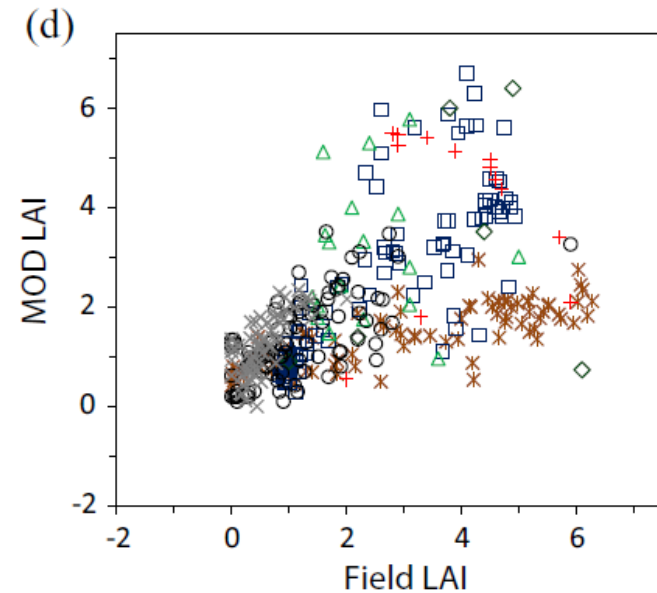
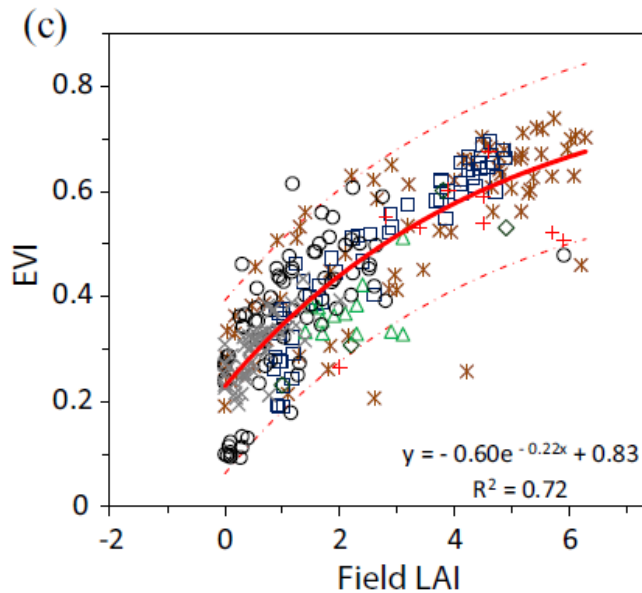
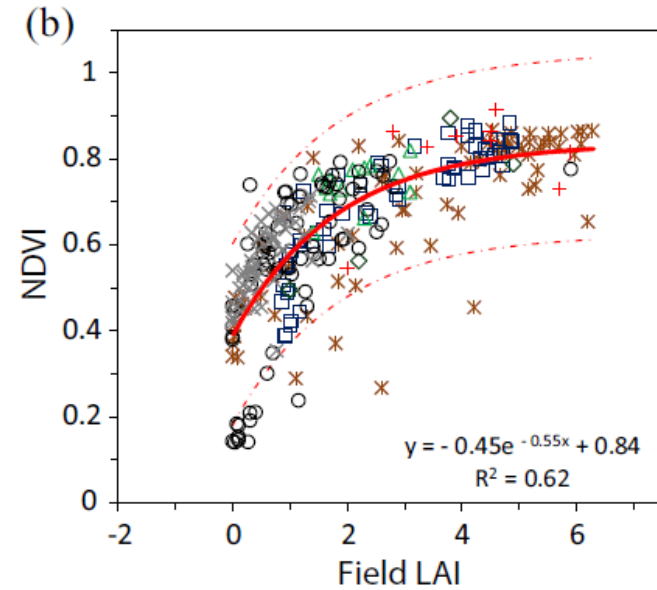
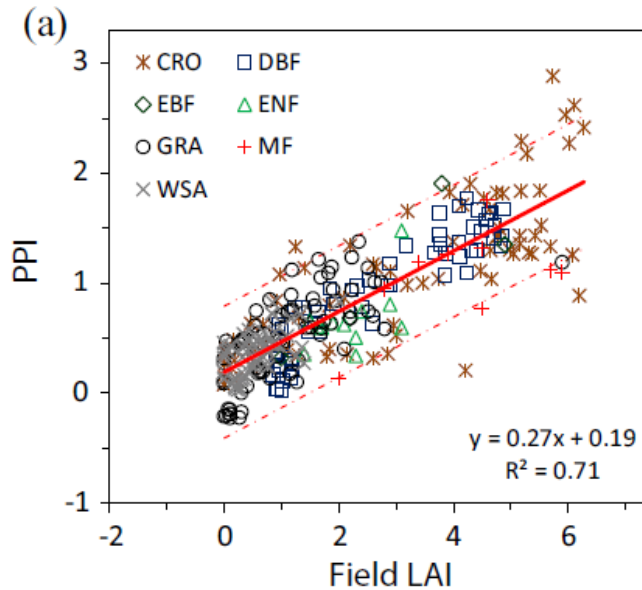


Insensitive
to snow



Validation against field LAI data

LAI data from
46 global
validation
sites
(BELMANIP,
Fluxnet, etc.)



Shoot development and GPP in spruce and pine forests

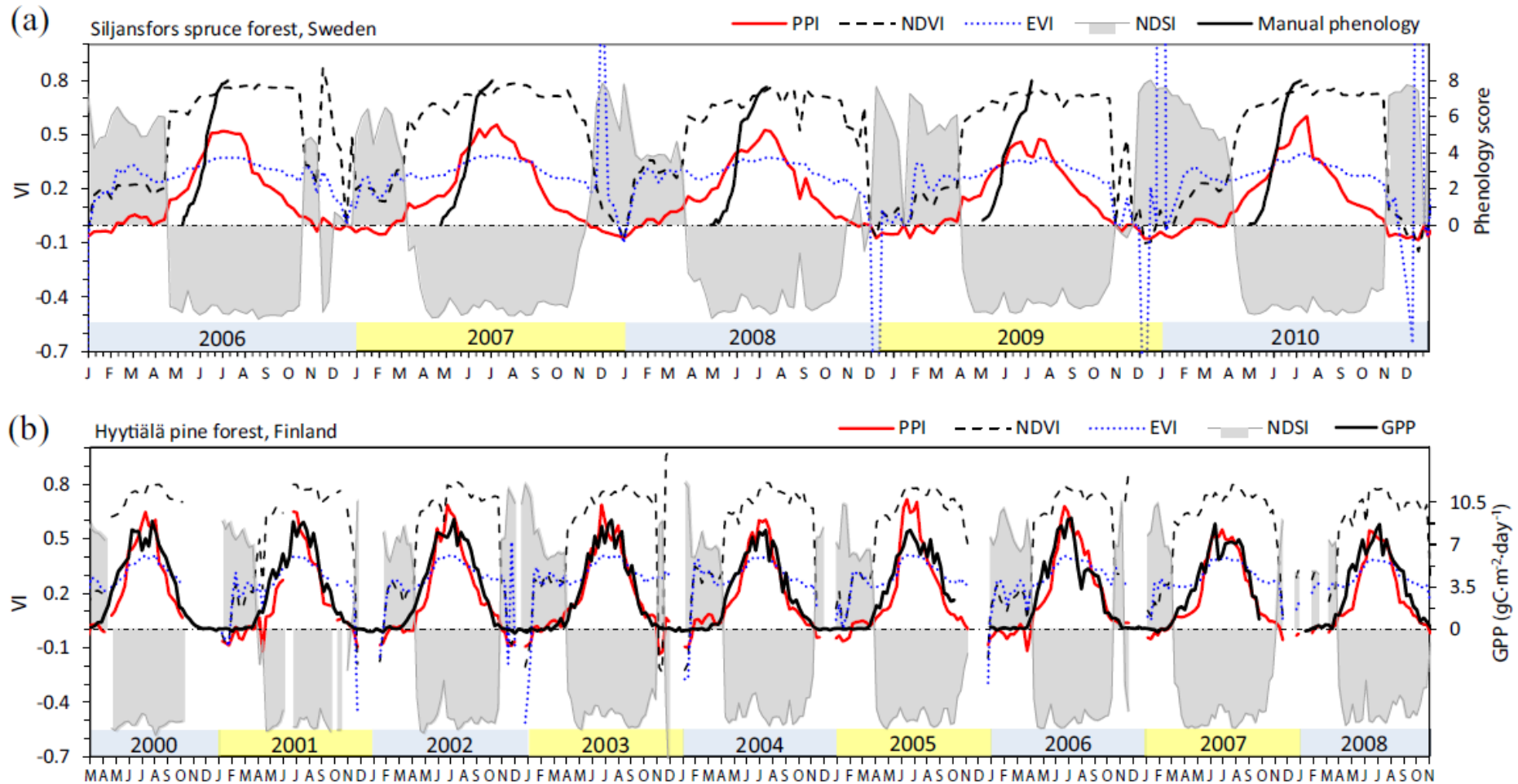
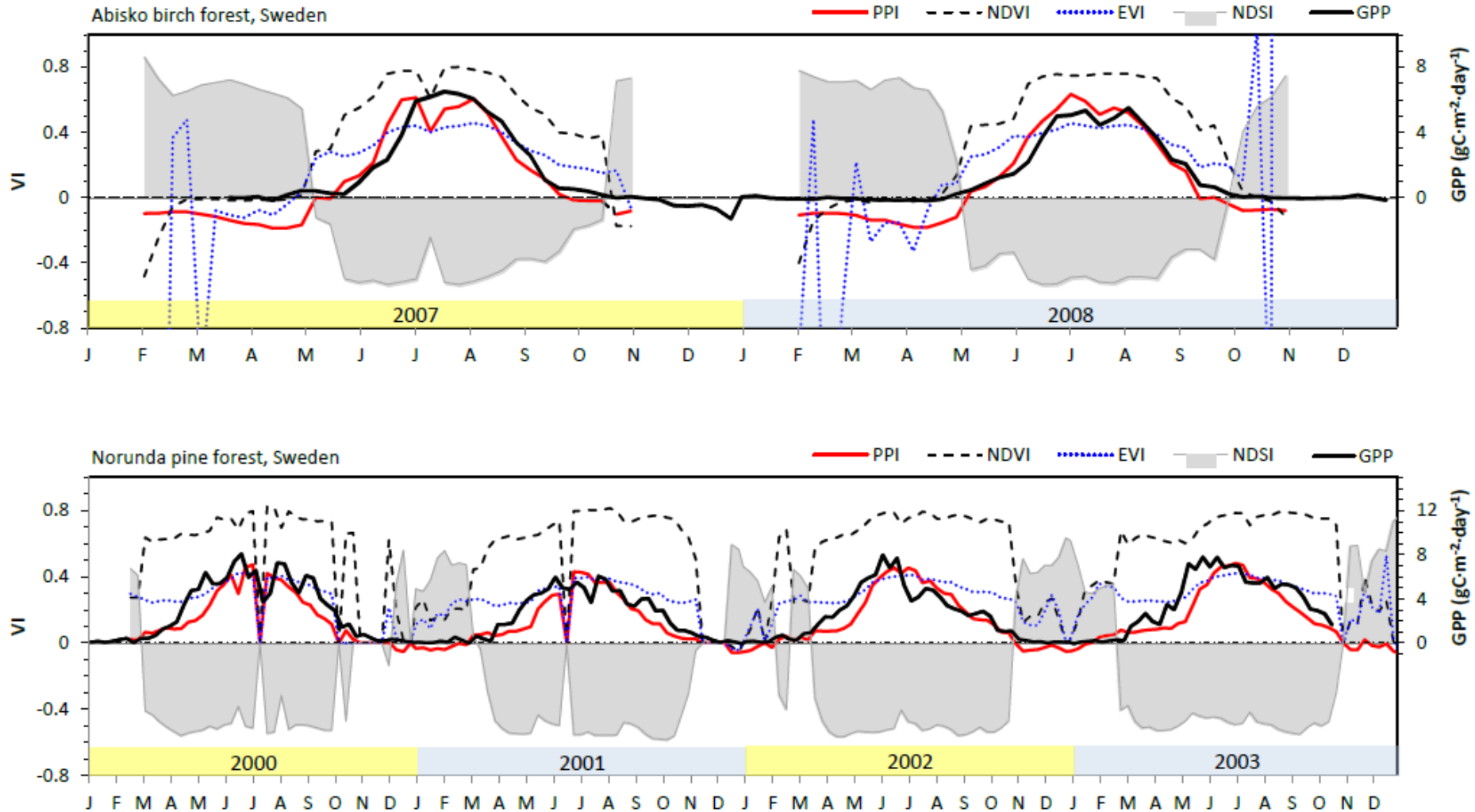


Fig. 11. Comparison of VI time series with ground observed vegetation dynamics: (a) manually observed spring phenology scores at Siljansfors spruce forest, middle Sweden; and (b) EC tower measured GPP at Hyytiälä pine forest, Finland. Shaded areas show snow activity indicated by the NDSI.

GPP: broadleaf and mixed coniferous sites



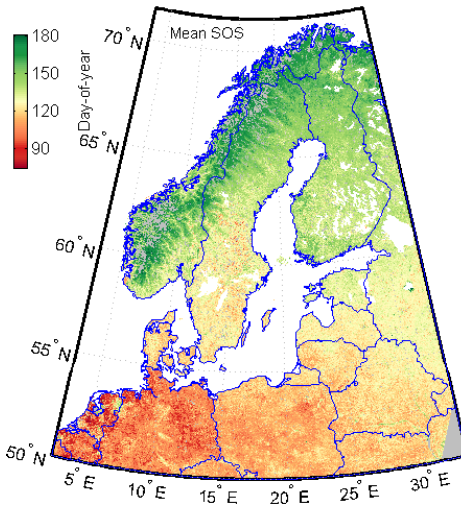
Correlations

Ground obs.	Index		
	PPI	NDVI	EVI
<i>Phenology score</i>			
Siljansfors spruce forest (2006–2010) (60.88°N, 14.40°E, 240 m a.s.l.)	0.93	0.81	0.94
Vindeln spruce forest (2006–2010) (64.23°N, 19.77°E, 180 m a.s.l.)	0.92	0.69	0.89
<i>GPP</i>			
Hyytiälä pine forest (2000–2008) (61.85°N, 24.29°E, 170 m a.s.l.)	0.93	0.73	0.61
Norunda pine forest (2000–2003) (60.09°N, 17.48°E, 70 m a.s.l.)	0.89	0.73	0.76
Abisko birch forest (2008–2009) (68.36°N, 18.80°E, 350 m a.s.l.)	0.93	0.79	0.77

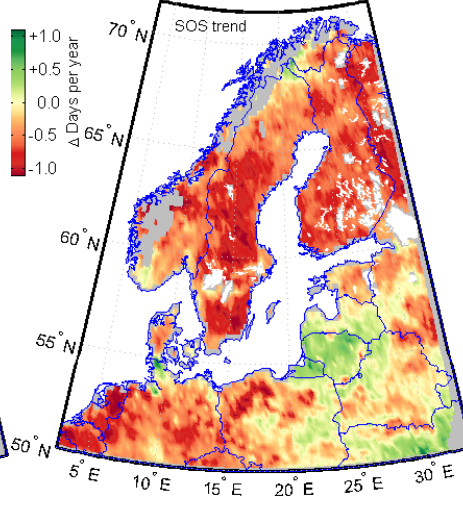


Phenology variations 15 years from MODIS PPI

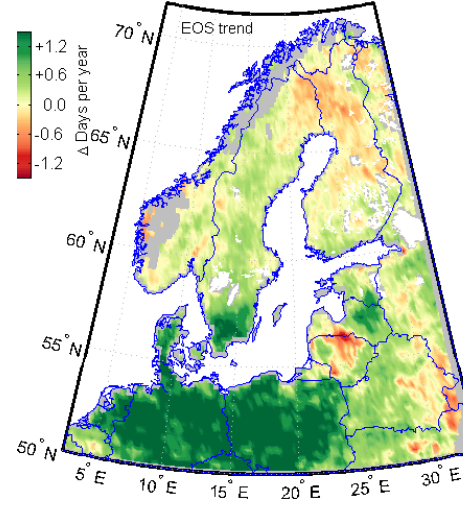
Mean SOS



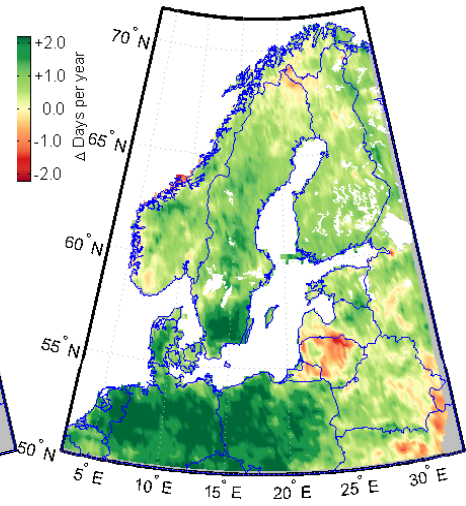
SOS trend



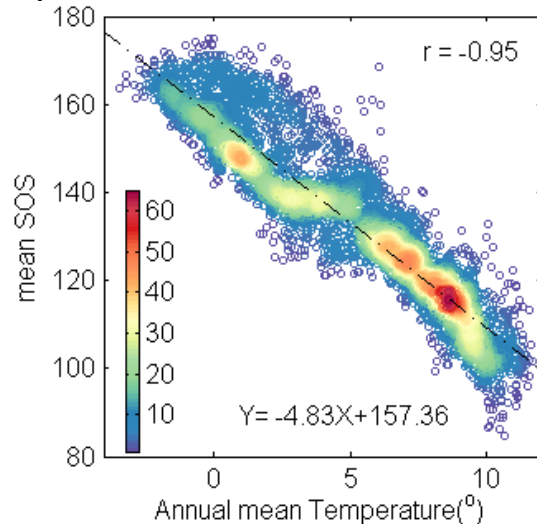
EOS trend



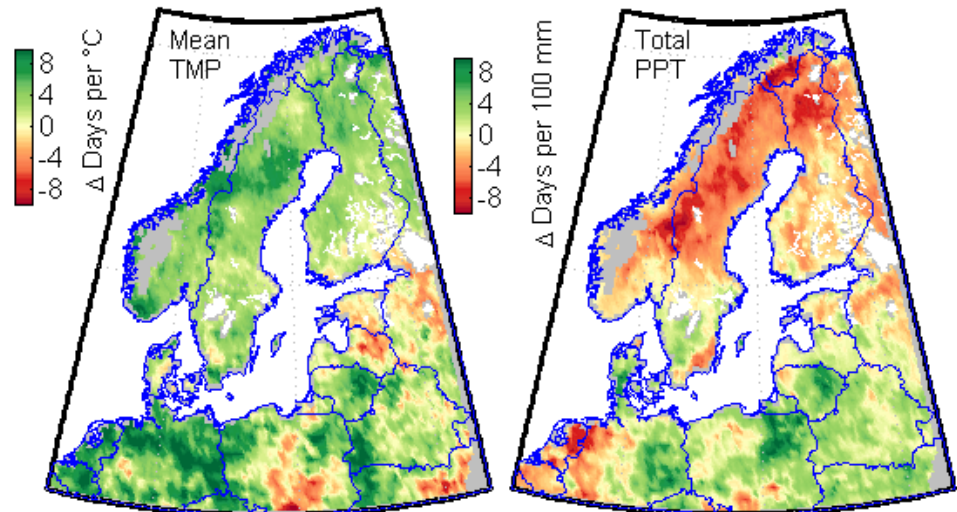
LOS trend



Spatial relation SOS – mean T

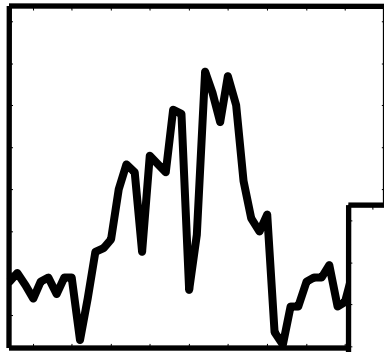


Sensitivity of LOS to temp and precip

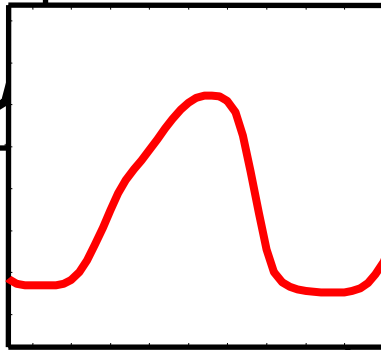


Hongxiao Jin et al. in prep.

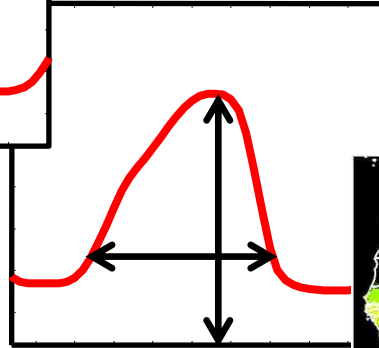
Modeling time-series data using TIMESAT



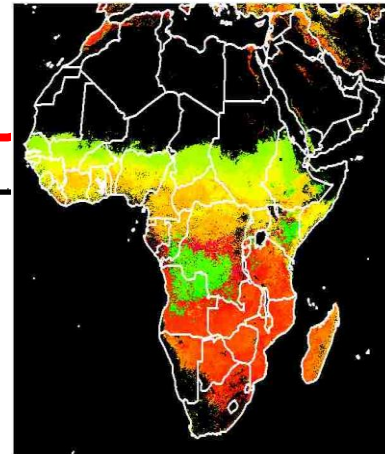
- Generate smooth model functions from noisy satellite sensor data



- Extract parameters that describe the seasons



- Map the seasonality parameters



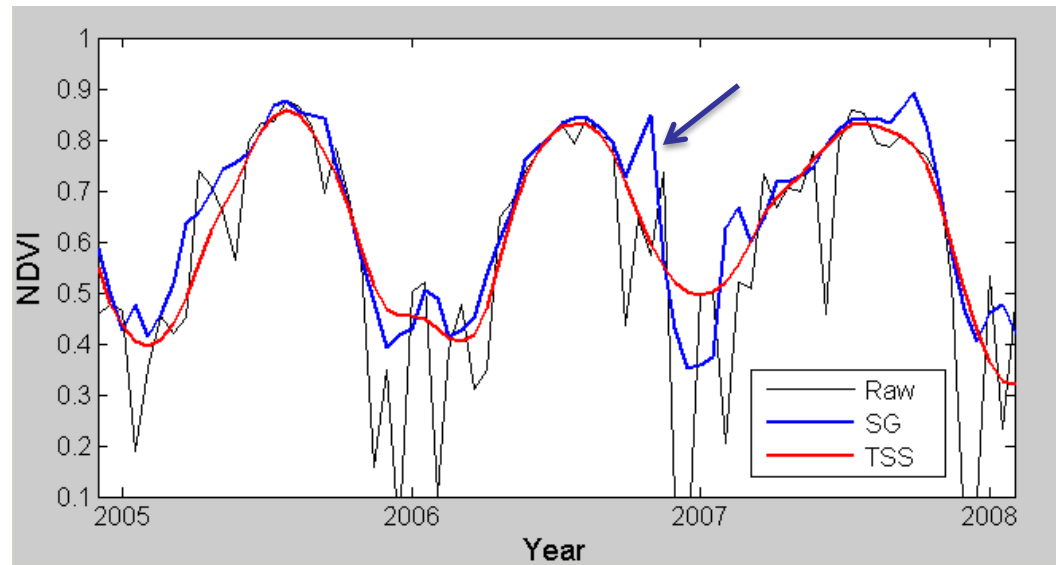
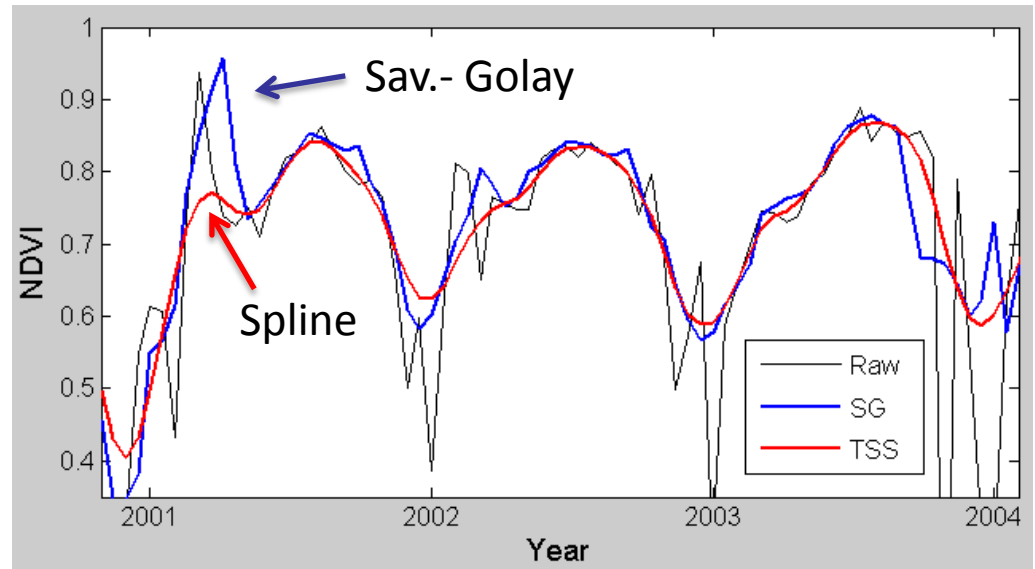
Jönsson & Eklundh 2002, *IEEE TGRS*, 40
Jönsson & Eklundh 2004, *Comp. Geosc.*, 30

<http://www.nateko.lu.se/TIMESAT>

Smoothing splines for local modeling

E.g. method by Garcia, 2010,
Comp. Stat. Data Anal,
54, 1167-1178

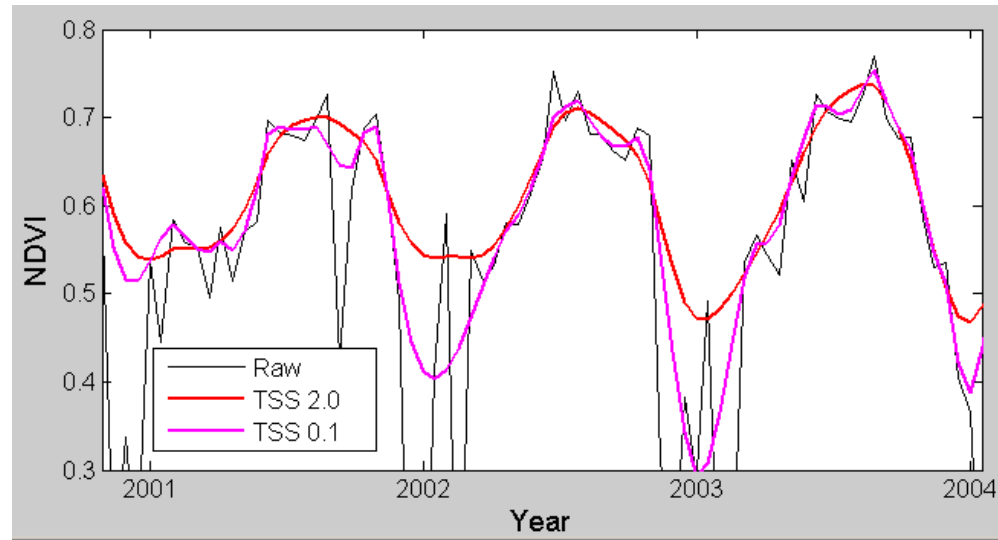
- Can work in one or several dimensions
- Is robust to outliers
- Allows for weighting the variables
- Can adapt to the upper data envelope
- Tolerates missing data



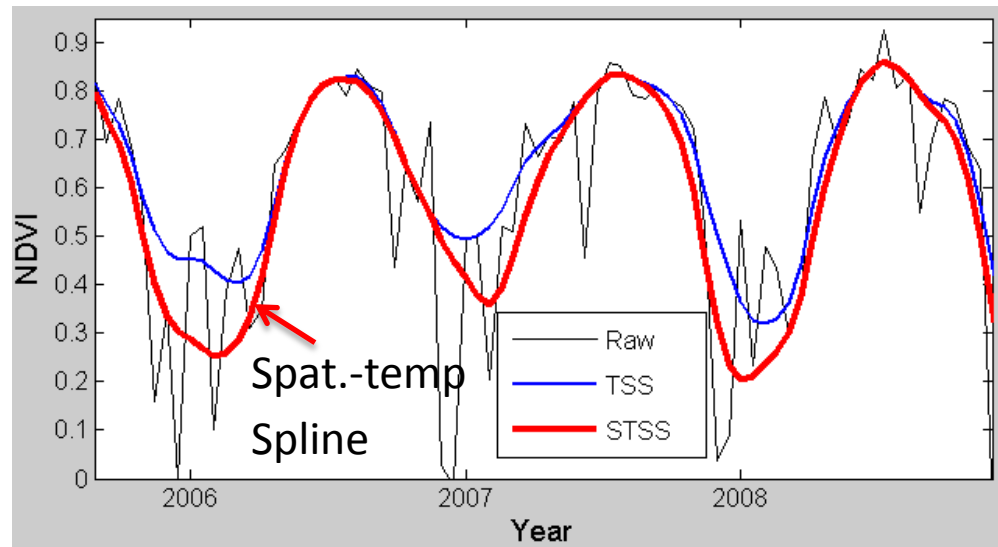
Eklundh et al. 2013, Proc. ISRSE

Modifying the spline smoothing

Modifying the
degree of
smoothing

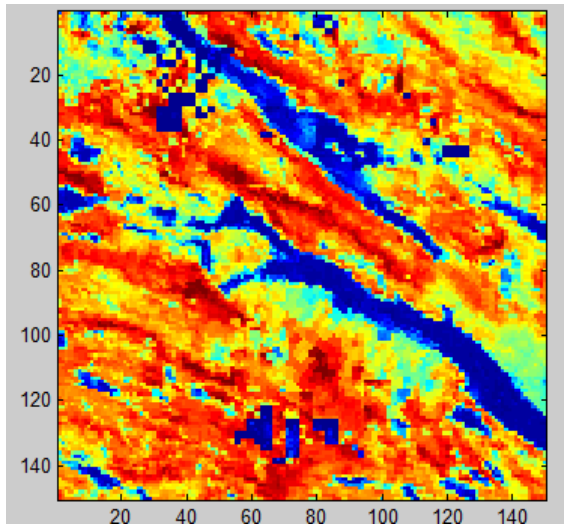


Smoothing over
the spatio-temporal
domain

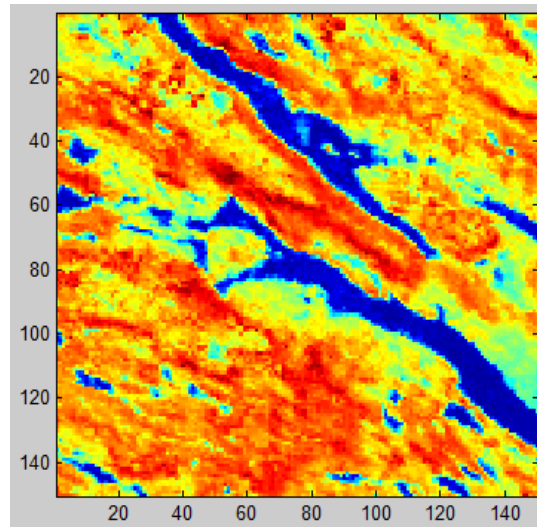


MODIS EVI images

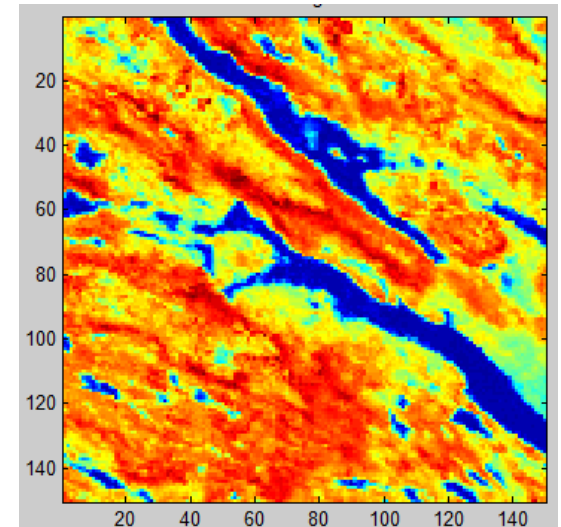
Raw data



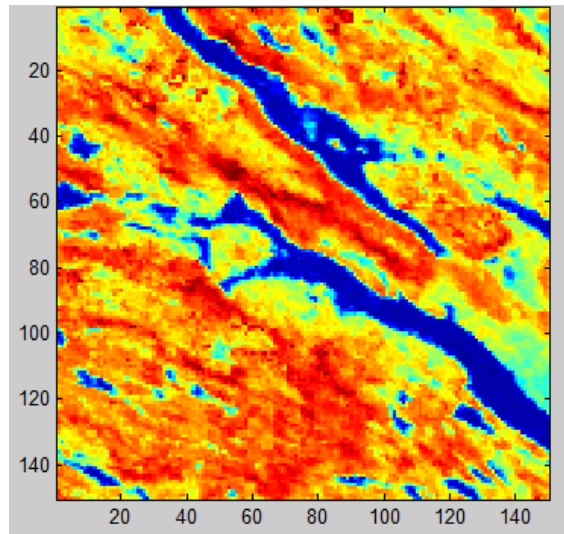
Savitzky-Golay



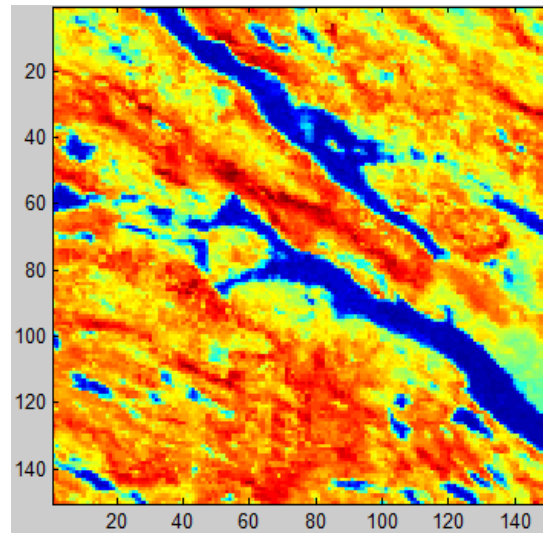
Asymmetric Gauss



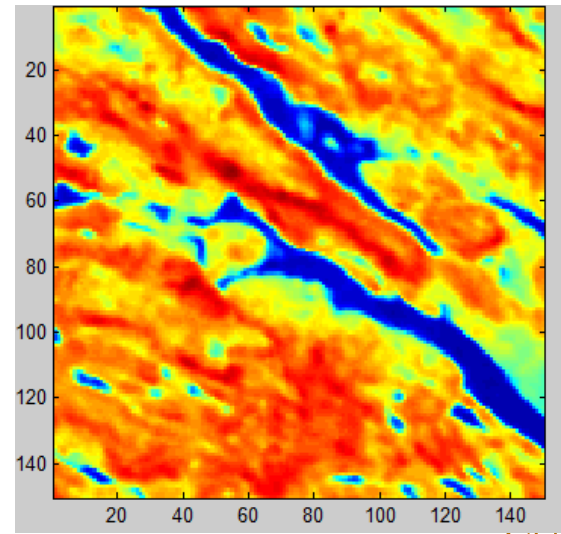
Double logistic



Temporal SSspline

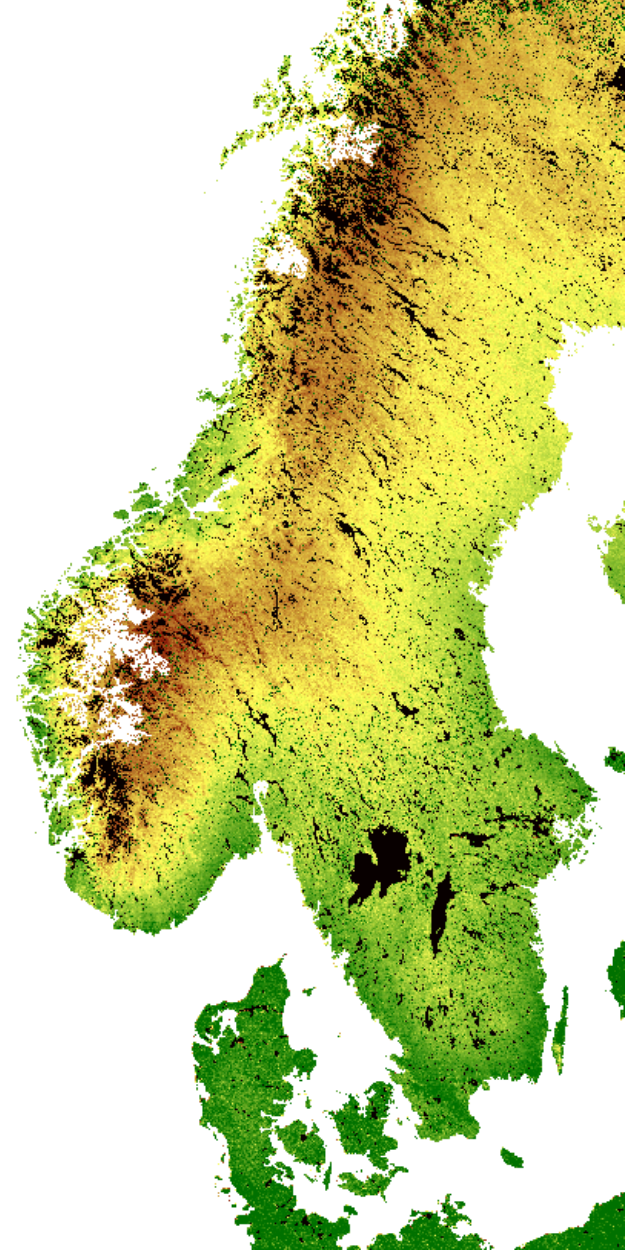


Spatio-temporal SS

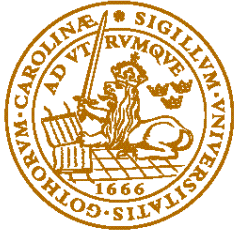


Summary

- New remote sensing techniques can fill the time-space data gaps and for improved understanding of general relationships
 - Fixed-sensor spectral measurements
 - UAVs
 - New satellites: Sentinel-2, Venμs
- Enable improved flux-footprint calculations
- PPI is a new vegetation index which responds to seasonal variations in LAI; shows excellent relationships with GPP and phenology
- New data modelling and integration methods, e.g. smoothing splines
- Success requires interdisciplinary collaboration between different science communities!



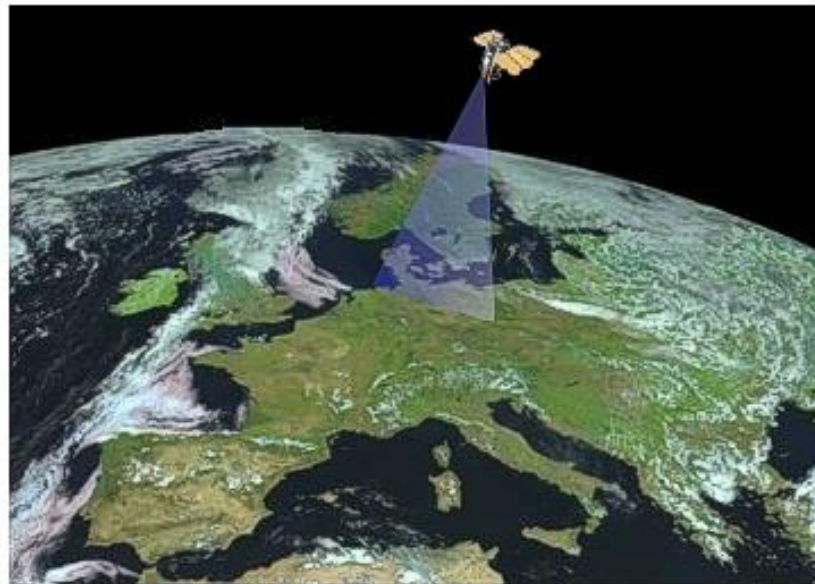
Thank you!



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E-mail: lars.eklundh@nateko.lu.se

Web: http://www.nateko.lu.se/earth_observation



PPI formulation

$$PPI = -K \times \ln \left(\frac{M - DVI}{M - DVI_S} \right)$$

DVI = difference VI

$$K = \frac{1}{4Q_E} \cdot \frac{1 + M}{1 - M} \cdot \frac{\varphi}{-\ln(1 - \varphi)}$$

M = max of DVI over a time period

K = extinction coefficient

$$Q_E = d_c + (1 - d_c) \cdot \frac{G}{\cos(\theta_i)}$$

Q_E = canopy light extinction efficiency

$$d_c = 0.0336 + 0.0477 / \cos(\theta_i)$$

d_c = diffuse fraction of solar radiation

$$G = \frac{\sqrt{\chi^2 \cos^2(\theta_i) + \sin^2(\theta_i)}}{\chi + 1.774(\chi + 1.182)^{-0.733}},$$

G = leaf angular distribution function

Jin, H. and Eklundh, L. (2014), A physically based vegetation index for improved monitoring of plant phenology. *Remote Sensing of Environment*, 152:512-525.

