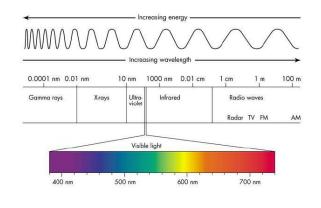


Science@SAFE 2014

# Forest degradation from above forest canopies

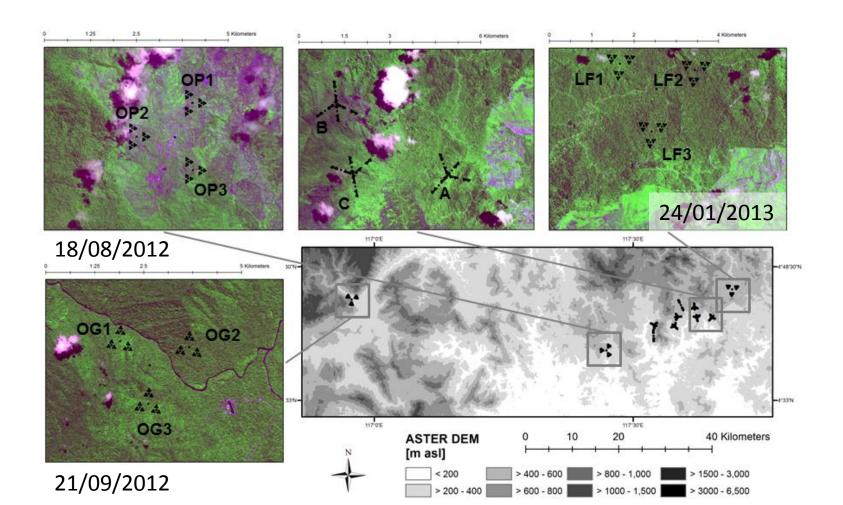
Marion Pfeifer<sup>1</sup>, Robert Ewers<sup>1</sup>



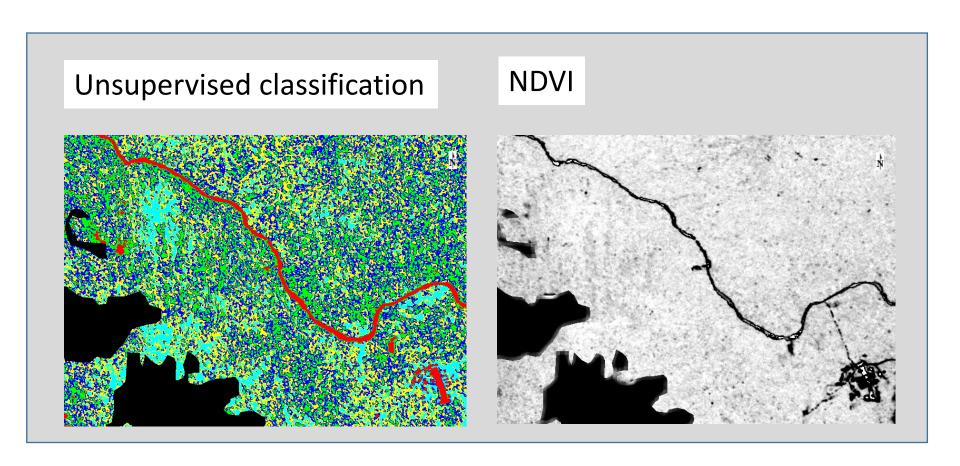




# RapidEye at SAFE provided by ESA and BlackEye

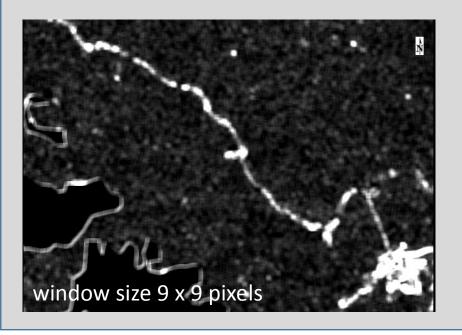


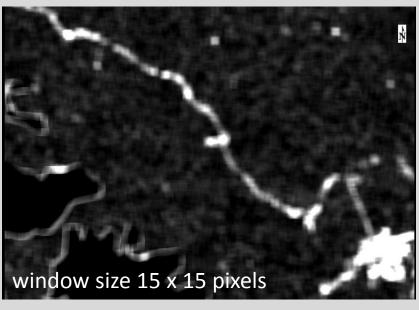
Linking plot attributes to spectral data measured from above canopies



# Linking plot attributes to texture data derived from spectral data measured from above canopies

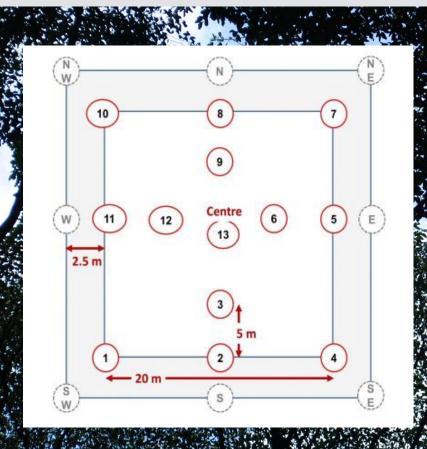
Dissimilarity band 3, offset =1, varying window sizes





## **Mapping forest attributes**

#### **Earth Observation At SAFE**



N = 193 (AGB), N = 203 (LAI)

# Leaf Area Index, fCover

Using hemispherical photography

# Aboveground biomass, AGB<sub>live</sub>

AGB\_Chave\_wet: global

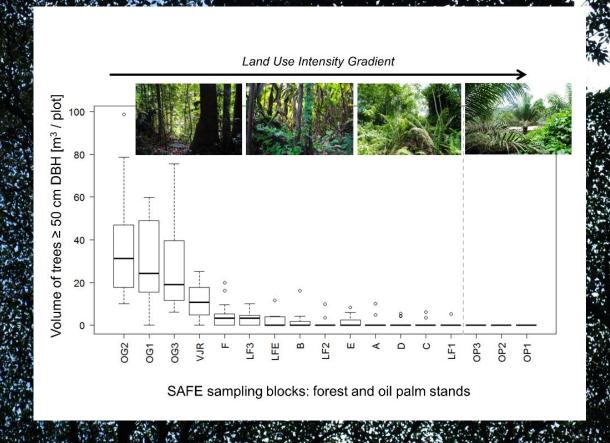
AGB\_Chave\_moist: global

AGB\_Niiyama\_2010: Malaysia

AGB\_Kenzo\_2009: Sabah

AGB\_Basuki\_2009: East Kalimantan

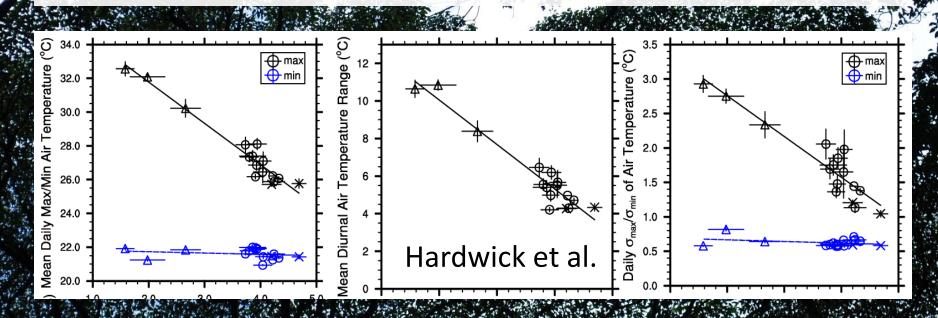
 Biomass (AGB<sub>live</sub>) declines significantly with increasing land degradation at SAFE



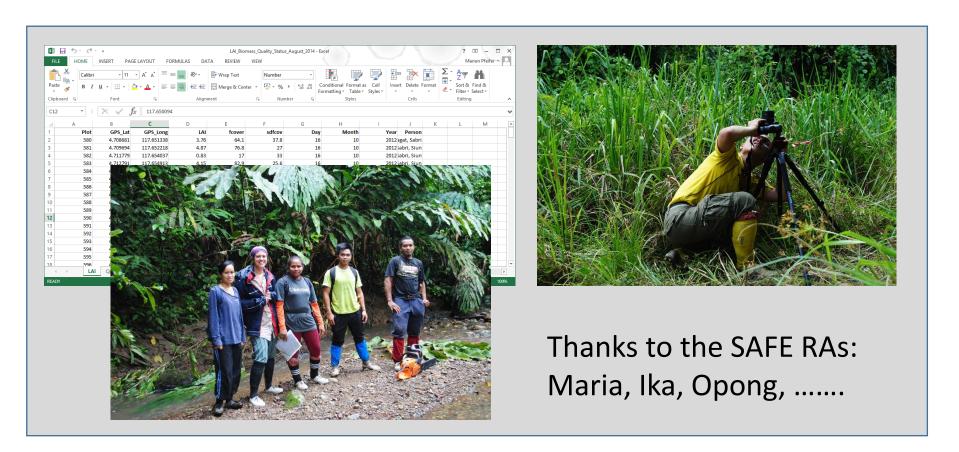
# Imperial College London

#### **Earth Observation At SAFE**

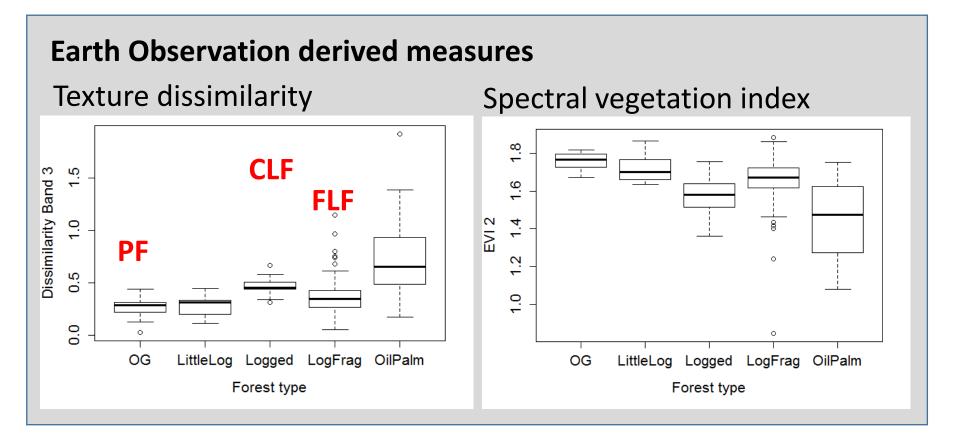
- Biomass (AGB<sub>live</sub>) declines significantly with increasing land degradation at SAFE
- LAI varies within and among forest stands and is clearly linked to microclimate



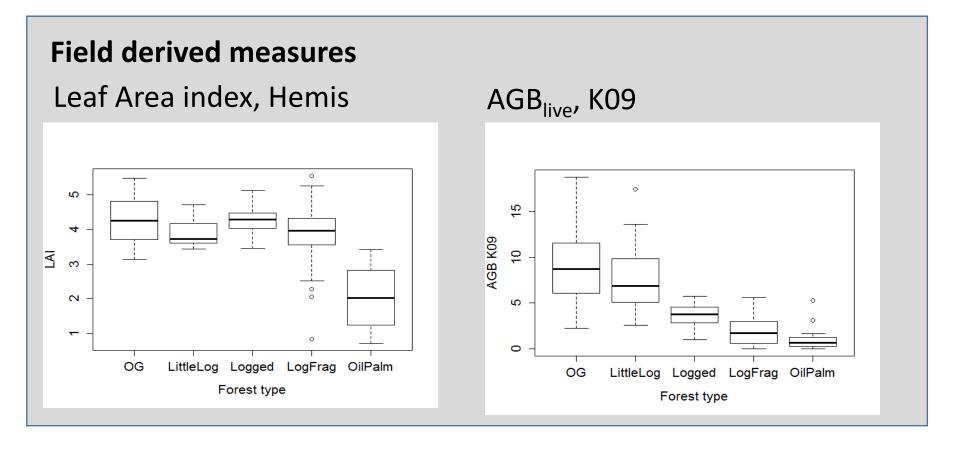
# Data distributed for SAFE analyses: LAI\_Biomass\_Quality\_Status\_August\_2014.xls



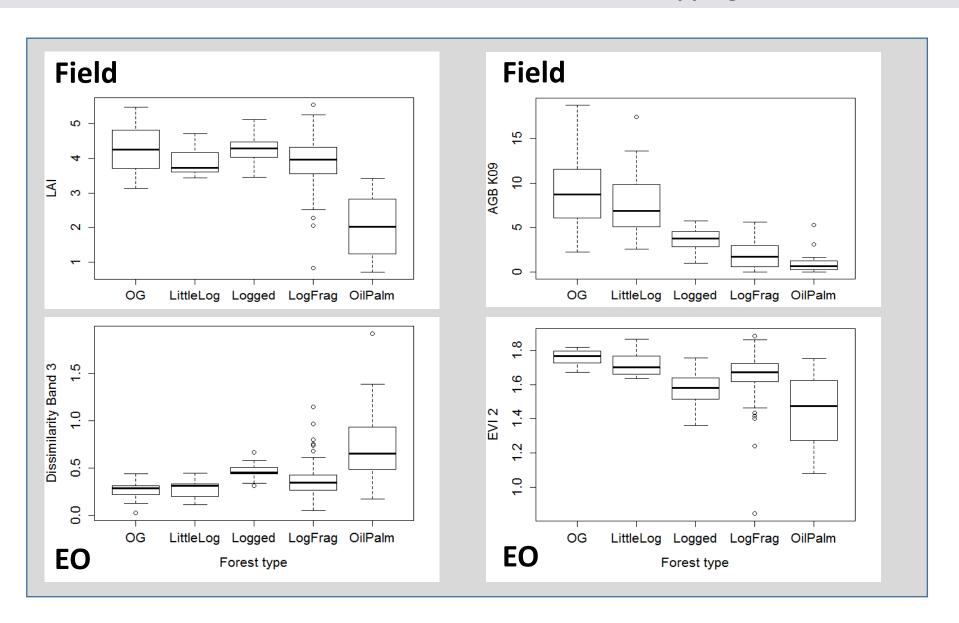
Grouping forest plots into four classes: pristine (OG), slightly logged (LittleLog), logged (Logged), heavily logged (LogFrag); and grouping plots in oil palm stands (OilPalm)



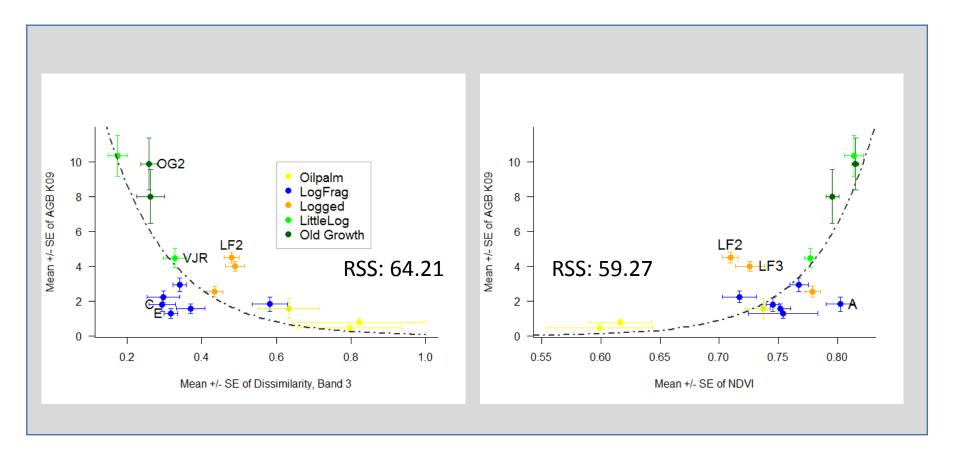
LAI, N = 191 (LFE plots are in the clouds), AGB, N = 185



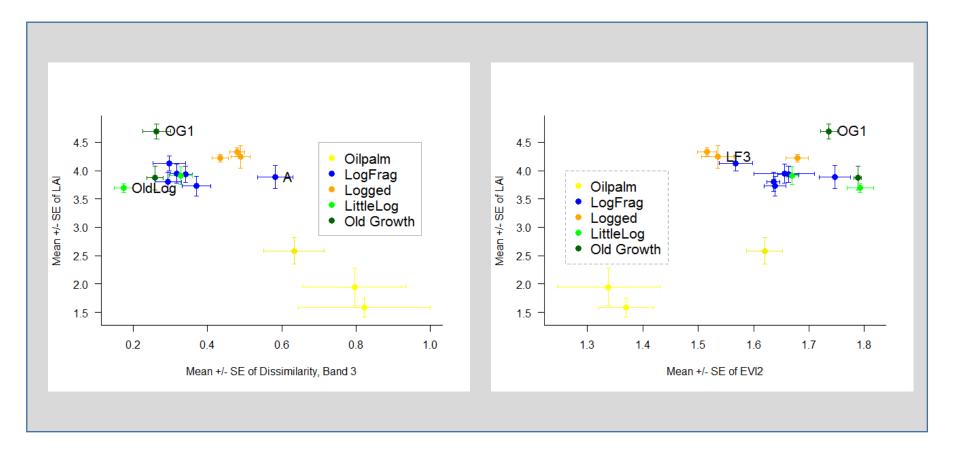
# **Mapping forest attributes**



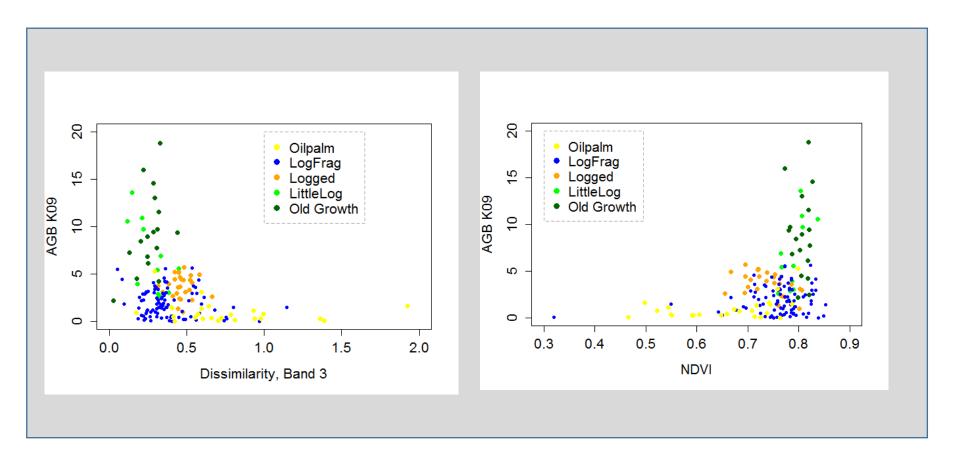
Maybe, at stand level, there is some relationship between satellite derived data and field derived AGB that can be exploited...... But!



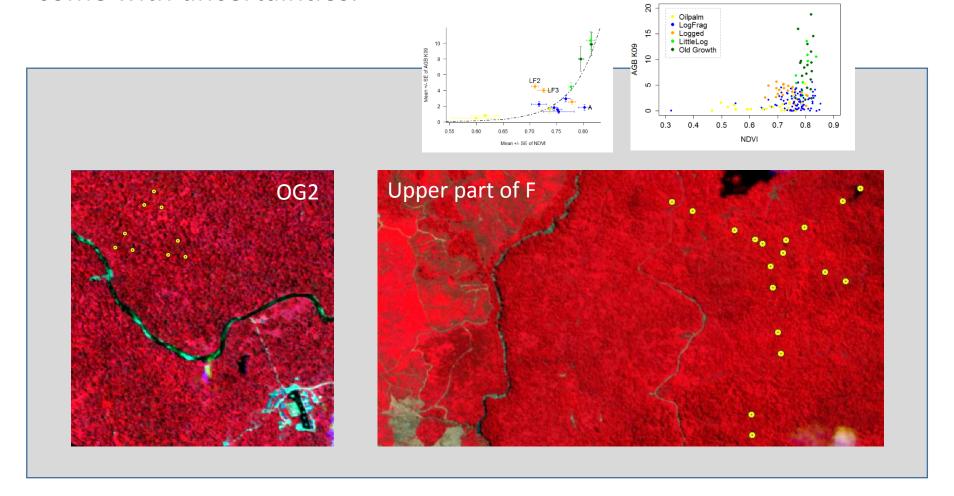
It doesn't look really good for Leaf Area Index, which is more directly linked to impacts on microclimate



And the predictive quality at plot level is questionable.....

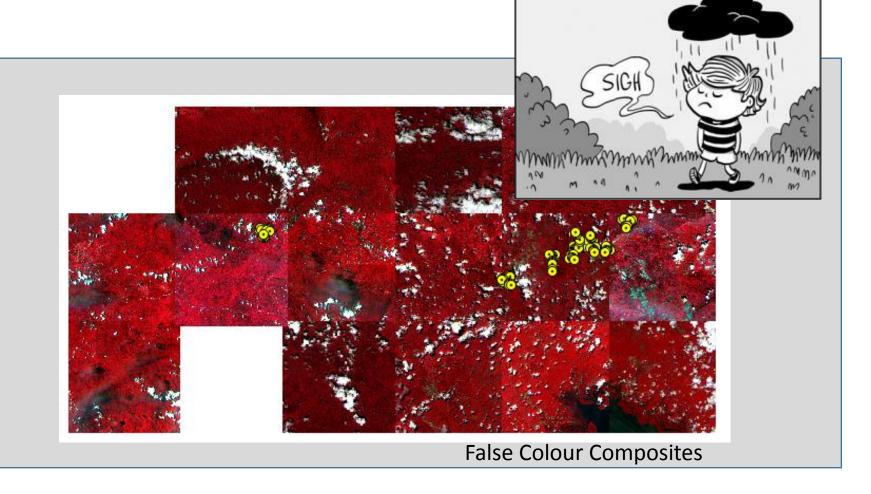


Predicting forest attributes for each pixel using these algorithms will come with uncertainties.

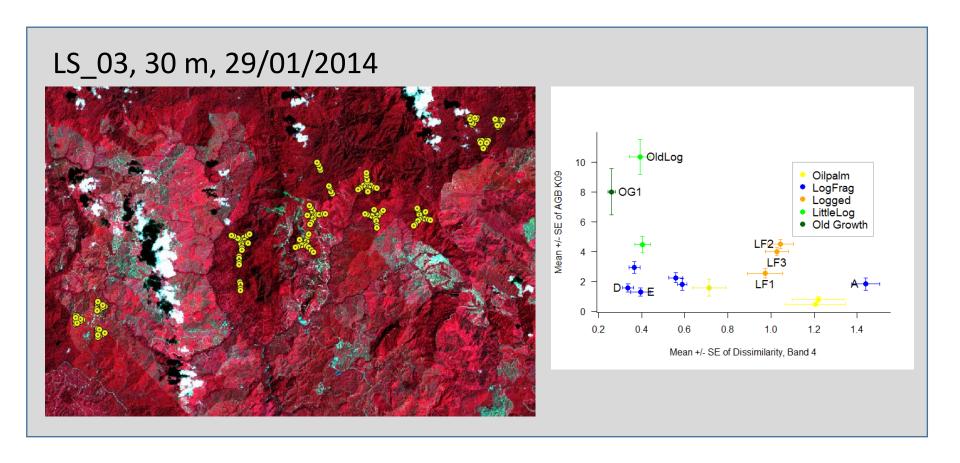


And to do so across the whole SAFE landscape requires to get

cloud-free images first.



Landsat images are good for cover change detection. But they seem to be less capable of depicting degradation.



But we do have high hopes for aerial images/aerial sensor data. Not that finding algorithms for these will make life any easier, for upscaling or back-casting. And there is plenty of room for using Sentinel data or other data sources.

### ?? Some time soon

# At the 'workshop' tomorrow:

- Which satellite data do we have in the SAFE database?
- Which sensor data are we hoping to get over the next years?
- What are these data what do they mean for my study?
- How can I make the image data work for me?
- And which maps are being produced to make life easy for me?

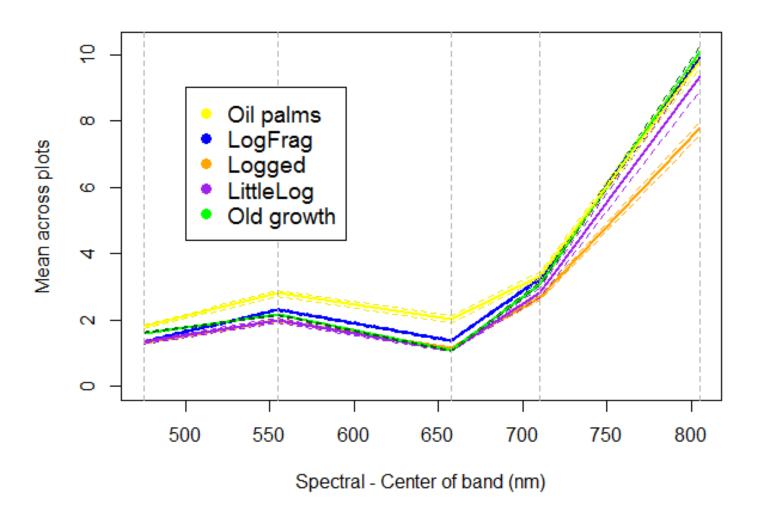
# Imperial College London

**Earth Observation At SAFE** 

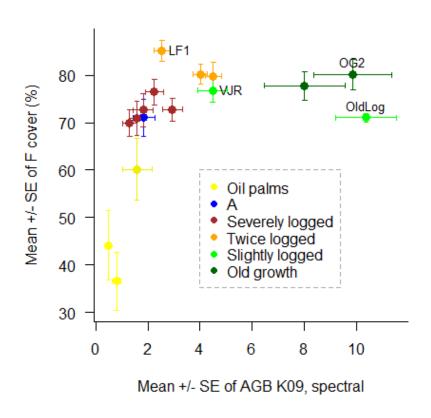
# Thank you

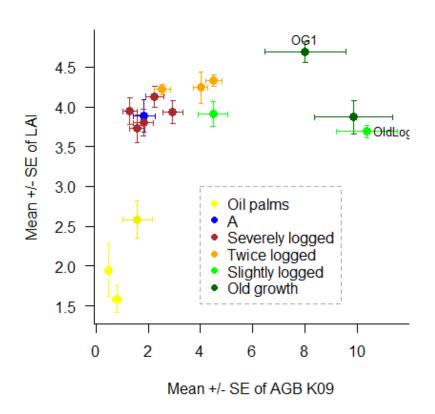


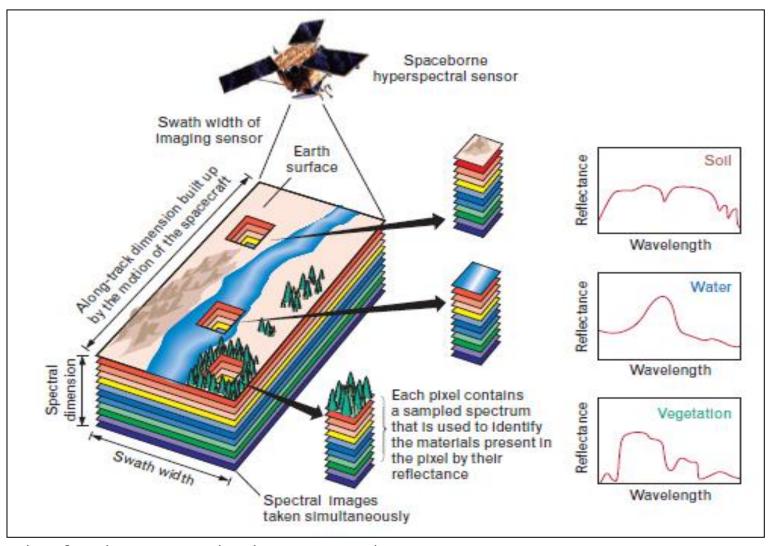
Funding ERC, Sime Darby Foundation, European Space Agency



## **Mapping forest attributes**







Shaw & Burke, 2003. Lincoln Laboratory Journal

# RapidEye

- five Earth Observation satellites, since 02/09
- Sensor: RapidEye Earth Imaging System (REIS)
- Image bands

Blue: 440 – 510 nm

Green: 520 - 590 nm

Red: 630 – 685 nm

Red Edge: 690 – 730 nm

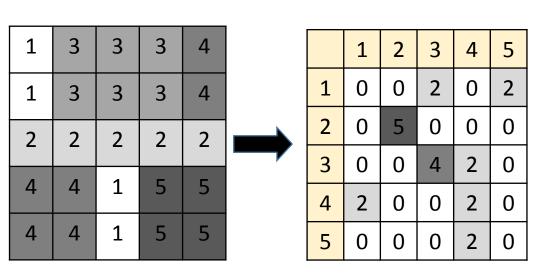
Near IR: 760 - 850 nm

- Resolution: 6.5 m (resampled to 5 m)
- Capability for daily revisit to any point on earth



# Texture analyses: e.g. grey level co-occurrence matrix (glcm)

- N of similar combinations of neighbours within window of specified size: to the right with offset = 1 (asymmetric)
- Option to rescale grey values before analyses



For each combination of neighbours: 1,1: Zero; 1,2; Zero; 1,3: Two, ....

Image with 5 grey values

Co-Occurrence matrix: 1 to right