

23 October 2018
Copernicus Global Land User Conference
Toulouse (France)

Copernicus Global Land
Providing bio-geophysical products of global land surface



Testing Copernicus Products to estimate forest carbon

SESSION: GLOBAL
TOWARDS HIGH
RESOLUTION



Pilar Durante
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Introduction

Results

**Materials and
methods**

**Discussion and
conclusions**



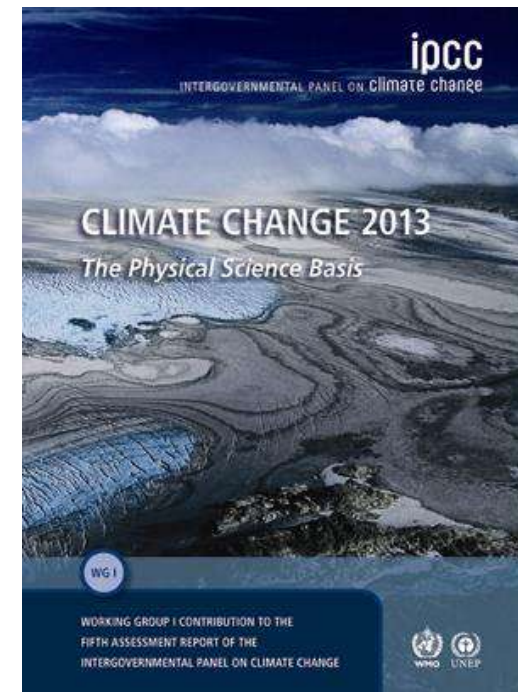
Future step

INTRODUCTION



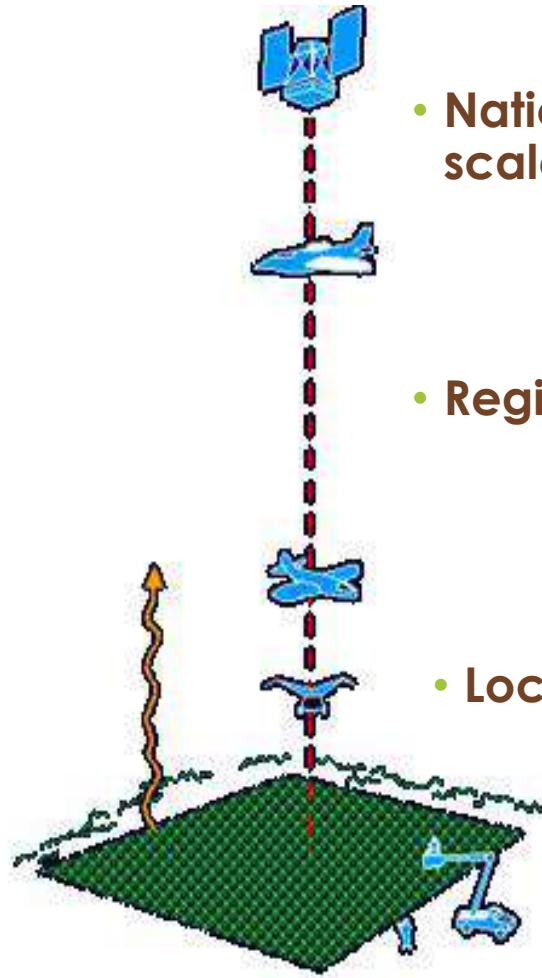
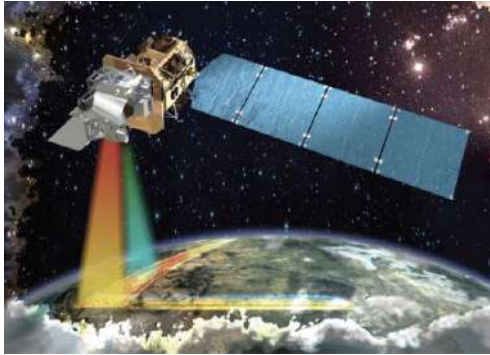
Introduction

- ✓ Threatened by global change
- ✓ 2 ° C Global Temperature Target
- ✓ Potential carbon sinks: Forest at national scale



Introduction

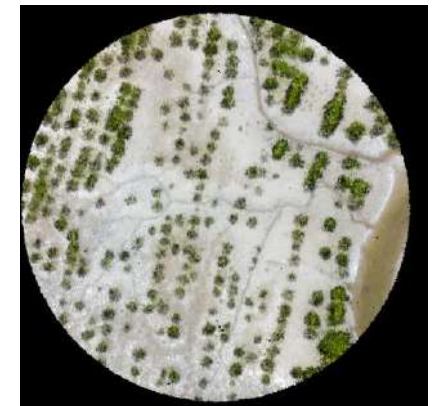
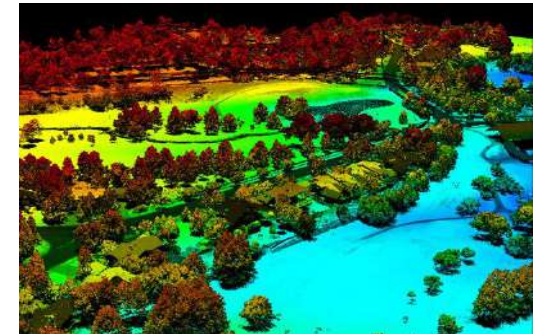
The issue of scale in remote sensing



- National/Global scale

- Regional scale

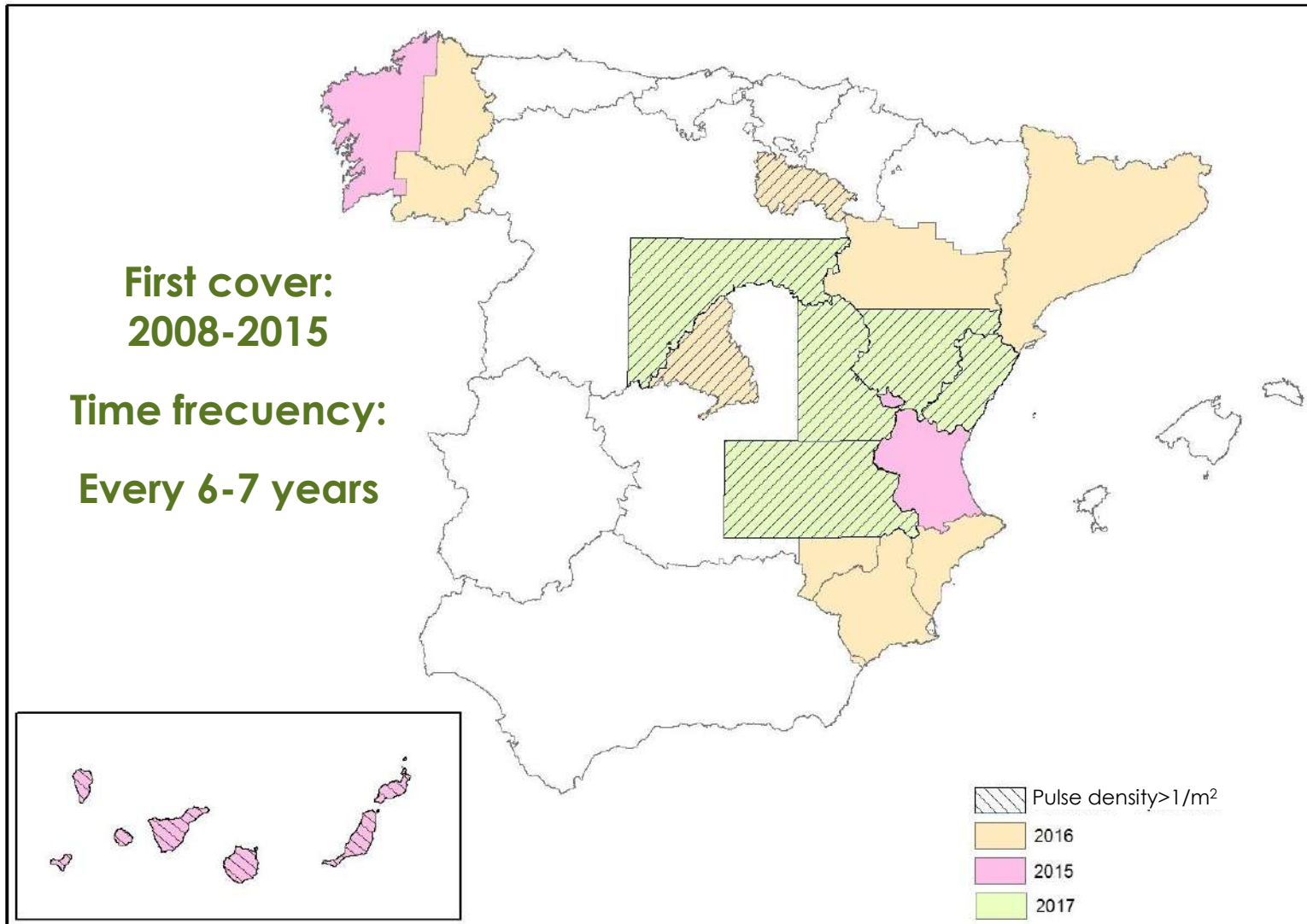
- Local scale



Introduction



LiDAR Images availability : Second cover of National Plan for Aviation Ortho (PNOA) 2016

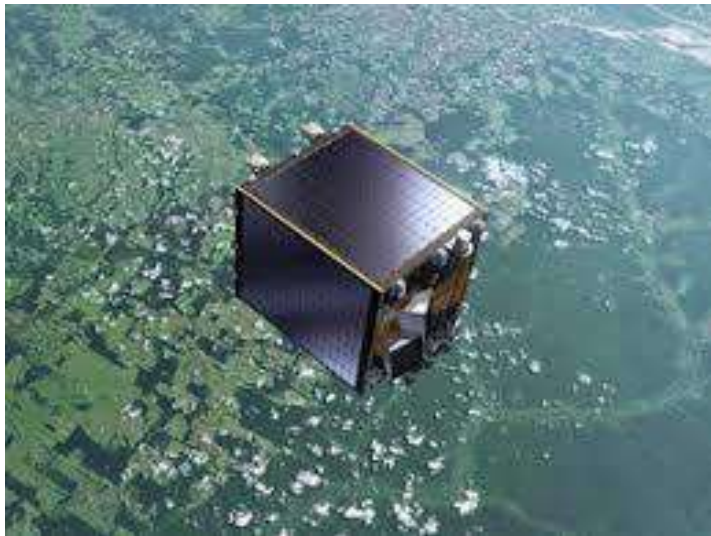


- Pulse density:
0.5 pulse/m²
- Mean vertical accuracy:
20 cm

Introduction



Copernicus Global Land Service: Providing bio-geophysical products of global land surface in near-real time. Vegetation products of 300 m pixel resolution each 10-day period



PROBA-V ESA Mission (2014)

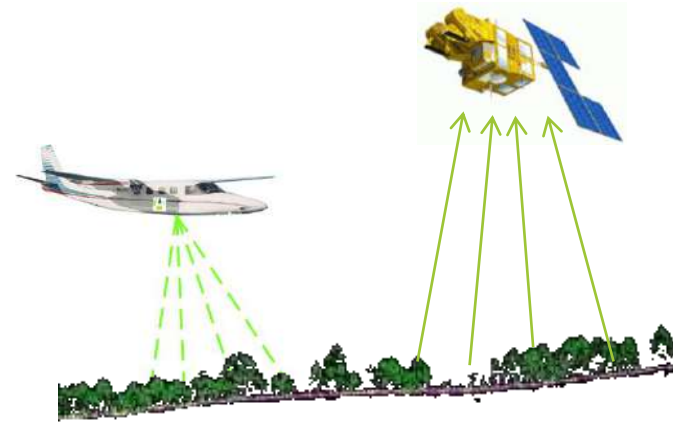


Introduction

What if we combine them?

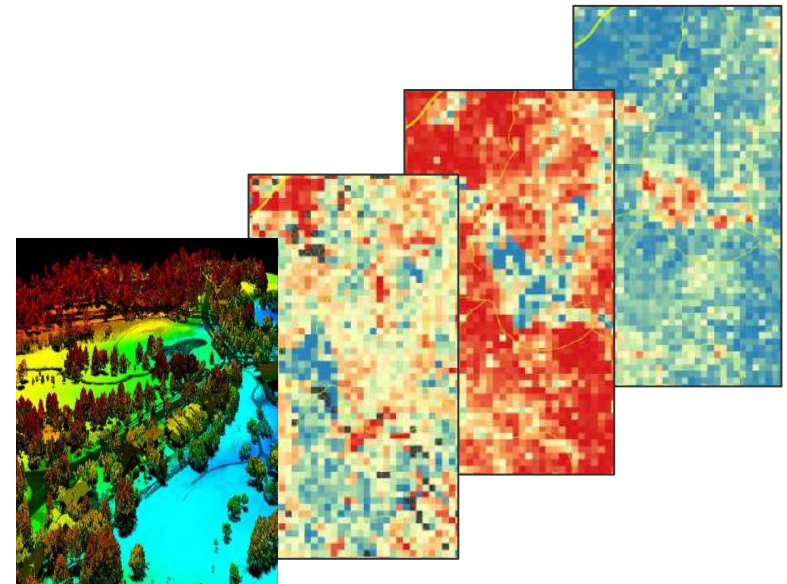


Combining LiDAR data with optical remote sensing images



High-accuracy LiDAR data at local scale

High temporal resolution at global scale



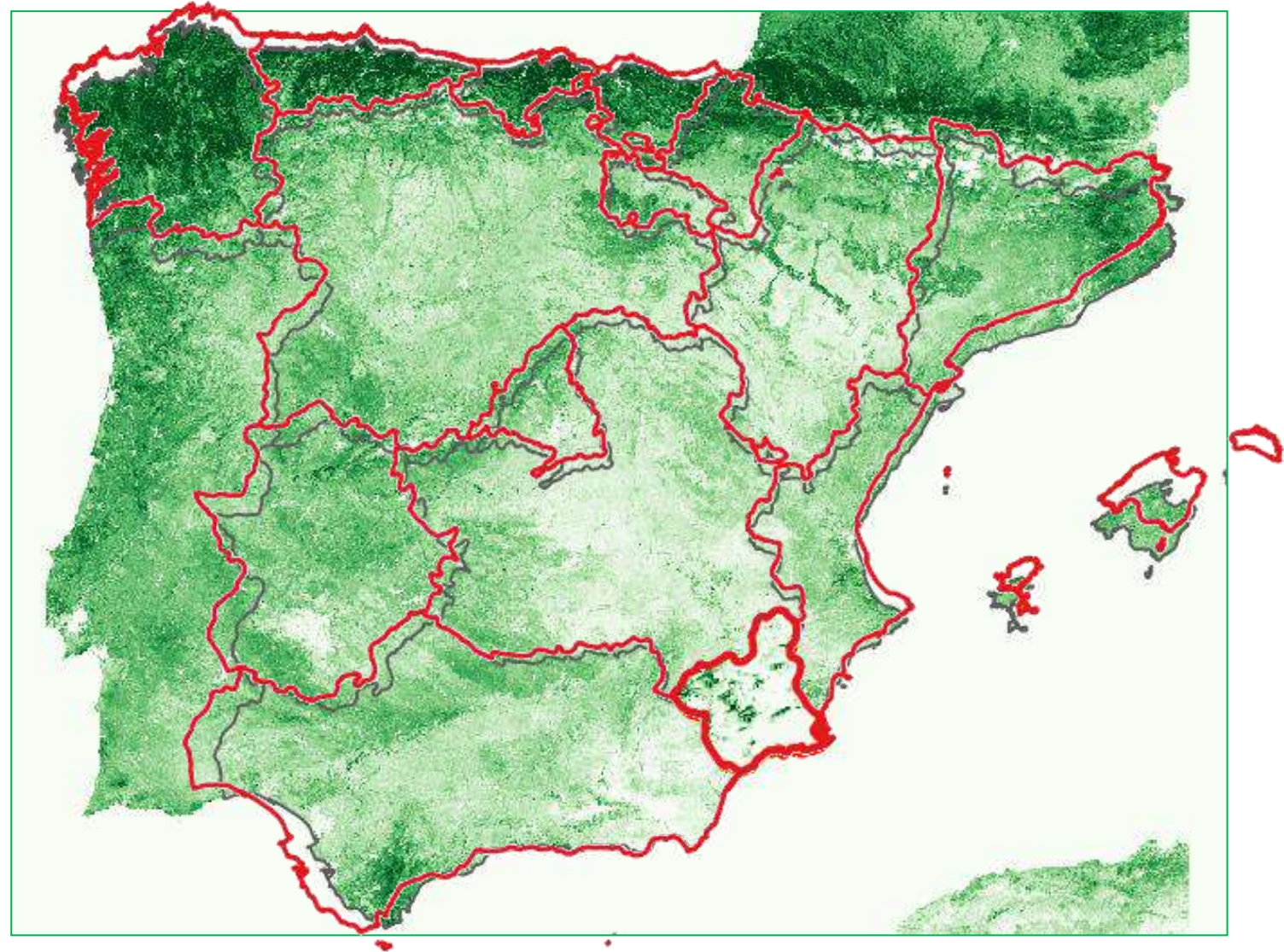
A close-up photograph of a tree trunk cross-section, showing concentric growth rings in shades of brown and tan. The texture is rough and natural. The text 'Materials and methods' is overlaid in a bright yellow-green color.

Materials and methods



Materials and methods

Study area: Region of Murcia



Materials and methods

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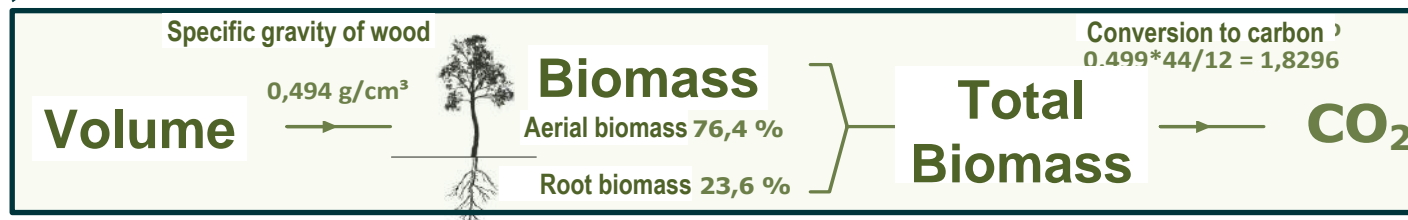
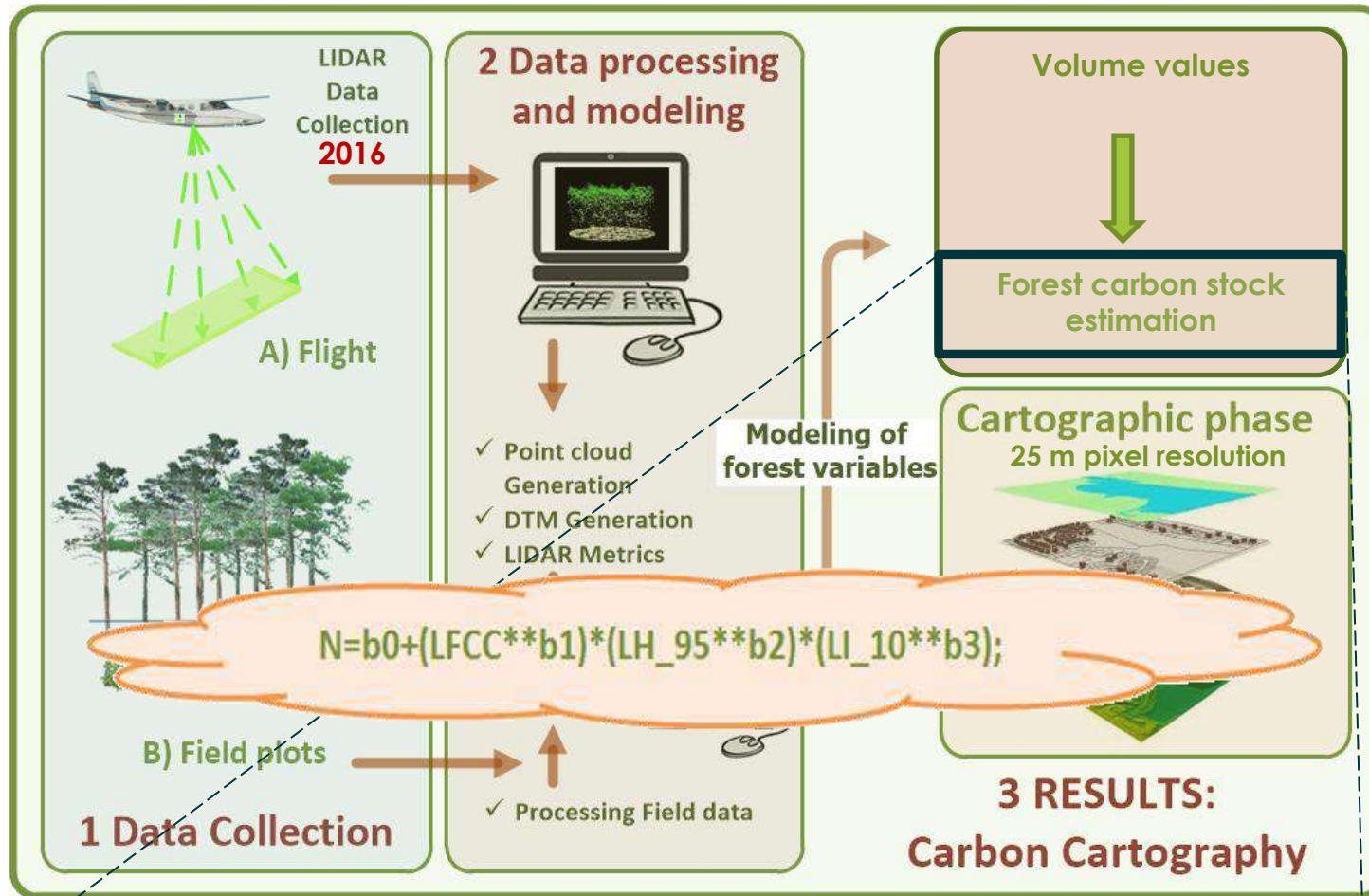


Materials and methods



✓ LiDAR data (PNOA2016) → Aboveground forest carbon map

SOURCE DATA



Materials and methods

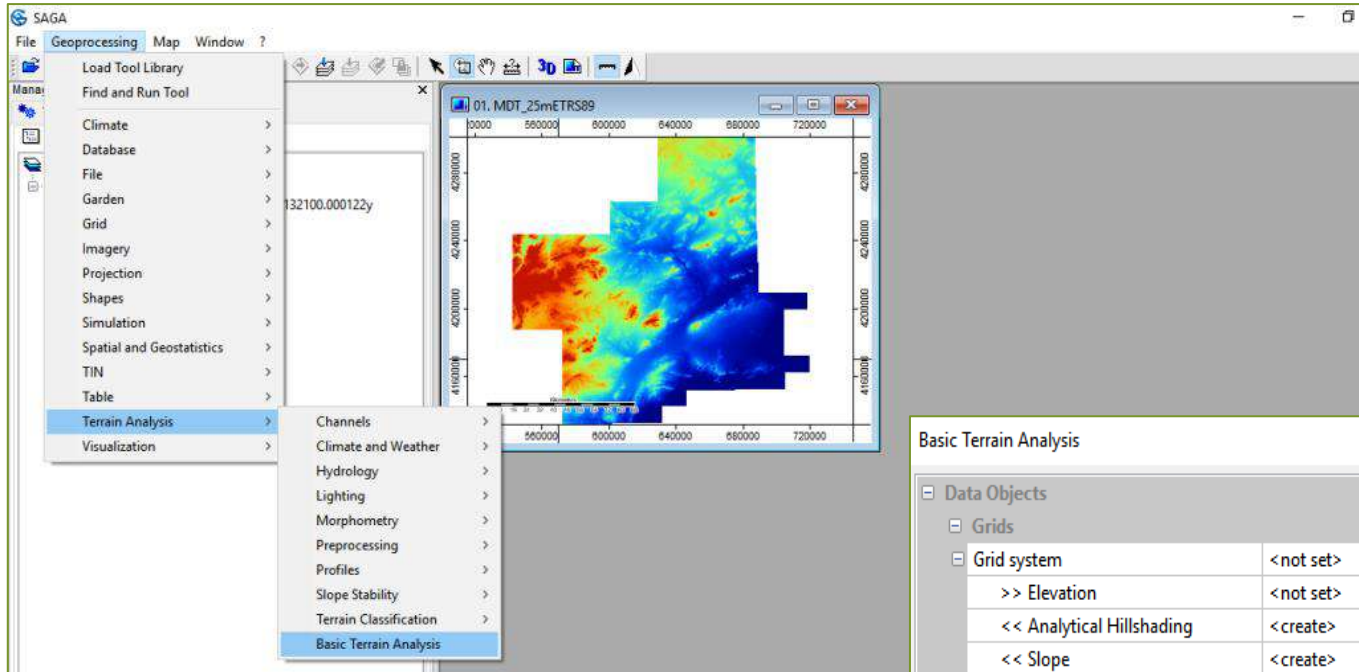


- ✓ LiDAR data (PNOA2016)  Aboveground forest carbon map
- ✓ DEM 25-m pixel resolution from «Spanish National Geographic Institute» (CNIG):
 - 14 terrain variables (SAGA GIS software)

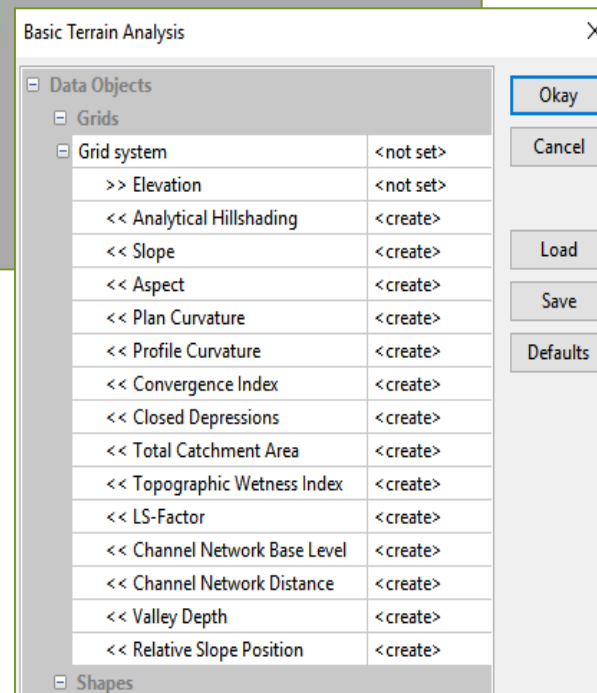
SOURCE DATA

Materials and methods

Source data




DEM-SAGA GIS →
14 topographic variables



Materials and methods



- ✓ LiDAR data (PNOA2016)  Aboveground forest carbon map
- ✓ DEM 25-m pixel resolution from «Spanish National Geographic Institute» (CNIG):
 - 14 terrain variables (SAGA GIS software)
- ✓ Forest Map of Murcia (1:25000) (MAPAMA): GIS Cover linked to a detailed forest database.
- ✓ Copernicus Global Land Service. Vegetation products 300 m:
 - NDVI , LAI, FAPAR, FCOVER

SOURCE DATA

Materials and methods

Source data

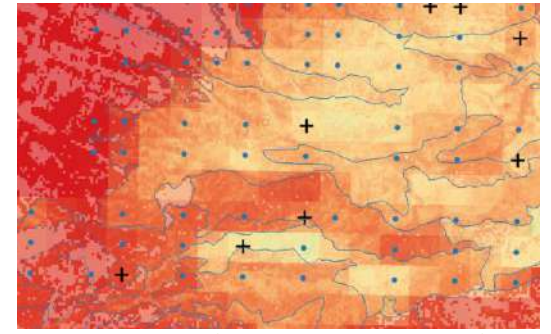
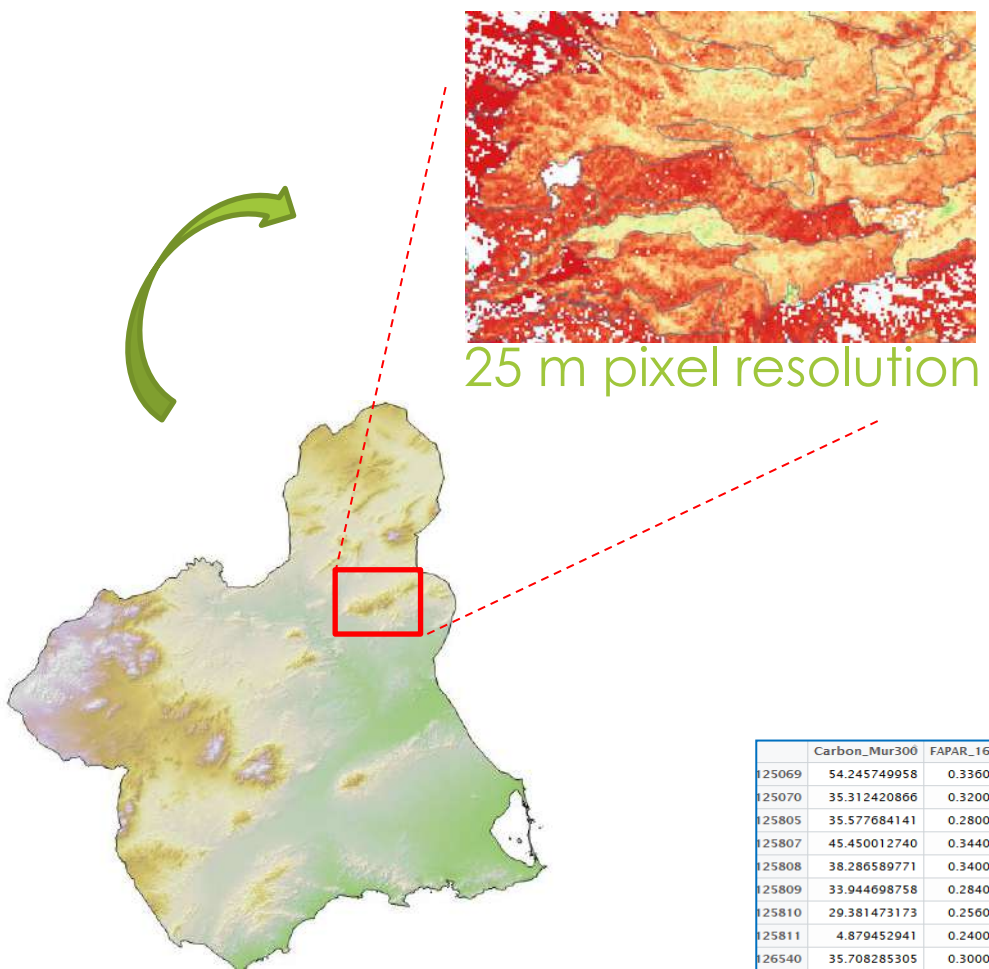


The screenshot displays the Copernicus Global Land Service website interface. At the top, there are navigation links for 'About' and 'Contact us'. The main header features the 'Copernicus Europe's eyes on Earth' logo and a navigation bar with 'Home', 'Products', 'News', 'Product Access', 'Viewing', and 'Library'. Below this, there are two prominent buttons: 'Burnt Area' and 'Land Cover'. The central part of the page shows a map of Europe with various cities labeled, such as Copenhagen, Warsaw, and London. To the right of the map, there is a 'Legend' section and a 'Catalogue search' box. Below the search box, there is a table for 'Vegetation Indicators - NDVI 300m V1' with columns for 'Date', 'Start date', 'End date', and 'ROI'. The table shows data for May and October 2016.

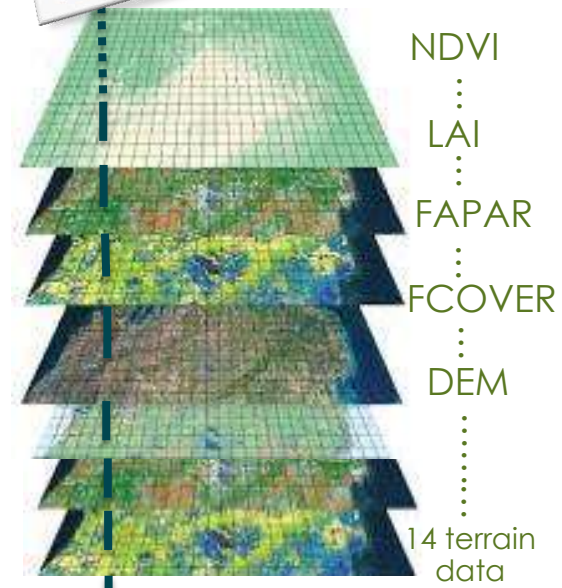
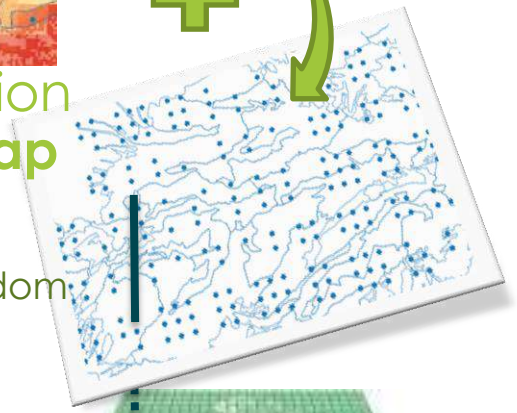
Date	Start date	End date	ROI
01/01/2014	01/01/2014	04/06/2018	38.76
04/06/2018	04/06/2018	04/06/2018	-2.4174
04/06/2018	04/06/2018	04/06/2018	37.35

May and October 2016

Methodological scheme



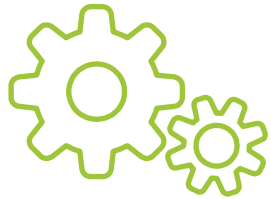
Train data set
Test data set (3,000 random points)



	Carbon_Mur300	FAPAR_160510	FAPAR_160520	FAPAR_160531	FAPAR_161010	FAPAR_161020
125069	54.245749958	0.33600003	0.33200002	0.31600001	0.32400000	0.33200002
125070	35.312420866	0.32000002	0.31200001	0.30400002	0.31600001	0.33600003
125805	35.577684141	0.28000000	0.28400001	0.27200001	0.31200001	0.34000000
125807	45.450012740	0.34400001	0.33200002	0.30800003	0.32400000	0.34000000
125808	38.286589771	0.34000000	0.33600003	0.32400000	0.34000000	0.36400002
125809	33.944698758	0.28400001	0.29200003	0.30000001	0.31600001	0.32000002
125810	29.381473173	0.25600001	0.25600001	0.23600002	0.17200001	0.18000001
125811	4.879452941	0.24000001	0.22400001	0.19200000	0.08800001	0.08800001
126540	35.708285305	0.30000001	0.28000000	0.24800001	0.28000000	0.29200003
126541	33.383498357	0.30400002	0.28800002	0.27600002	0.30000001	0.32000002
126542	30.456027013	0.29200003	0.28000000	0.27200001	0.30000001	0.30800003
126543	41.383917295	0.33200002	0.31200001	0.30400002	0.33200002	0.35600001

Methodological scheme

Carbon spatial predictive model



➤ **Three stages:**

- ✓ 1. Predictor variable selection
- ✓ 2. Spatial interpolation (QRF)
- ✓ 3. Estimation of uncertainty

➤ **1. Variables selection**

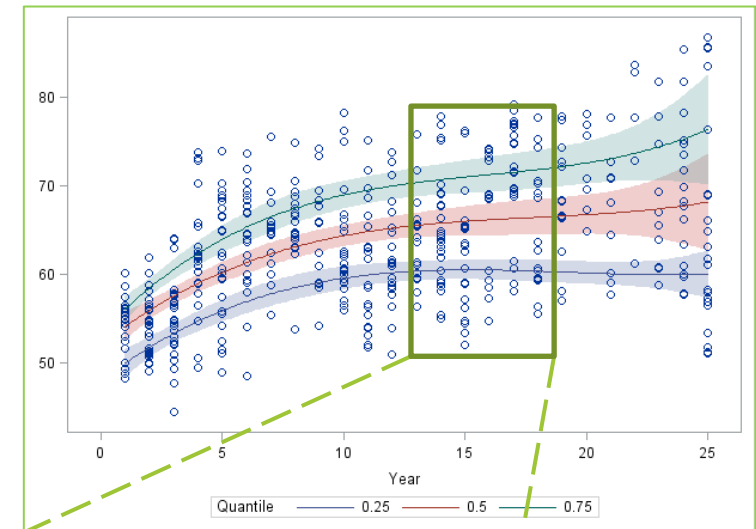
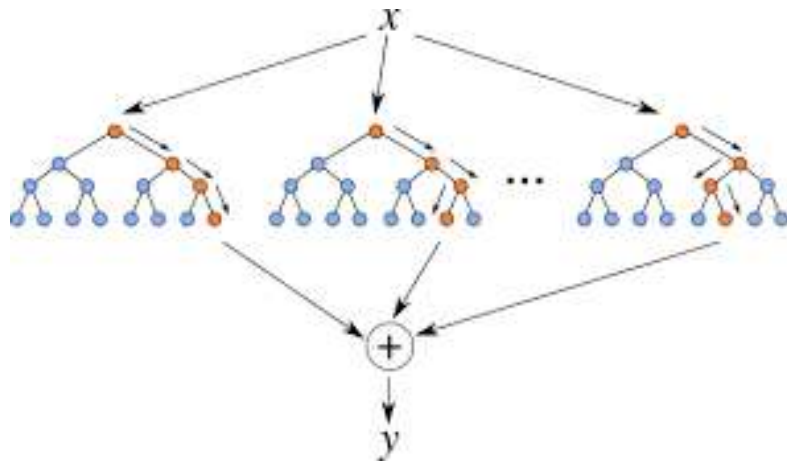
- ✓ Linear models (stepwise)
- ✓ Variance inflation factor (VIF)
- ✓ Variable Selection Using Random Forests (VSURF -R package)

Methodological scheme

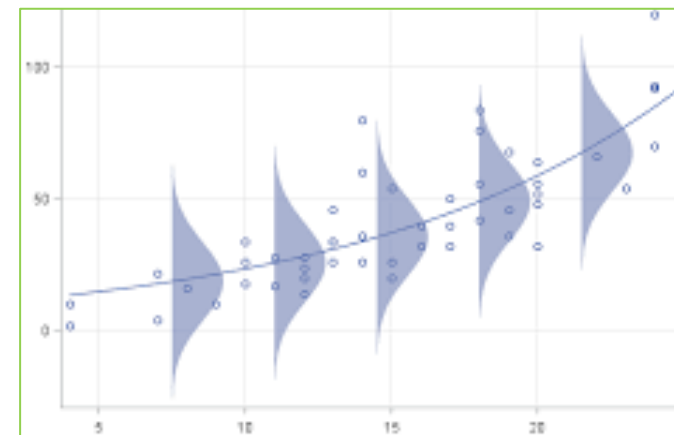
Carbon spatial predictive model




2. Spatial interpolation: Quantile Regression Forest (Meinshausen 2006)



3. Estimation of uncertainty



A close-up photograph of a tree trunk cross-section, showing concentric growth rings in shades of brown and tan. The texture is rough and natural. The word "Results" is overlaid in a bright green, bold, sans-serif font in the center.

Results



1. Selected variables:

- Vegetation index: NDVI (2016/10/11 and 2016/05/21) and LAI (2016/10/10);
- Topographic variables: DEM, relative slope position and channel network base level.

2. Spatial interpolation. Predictive model:

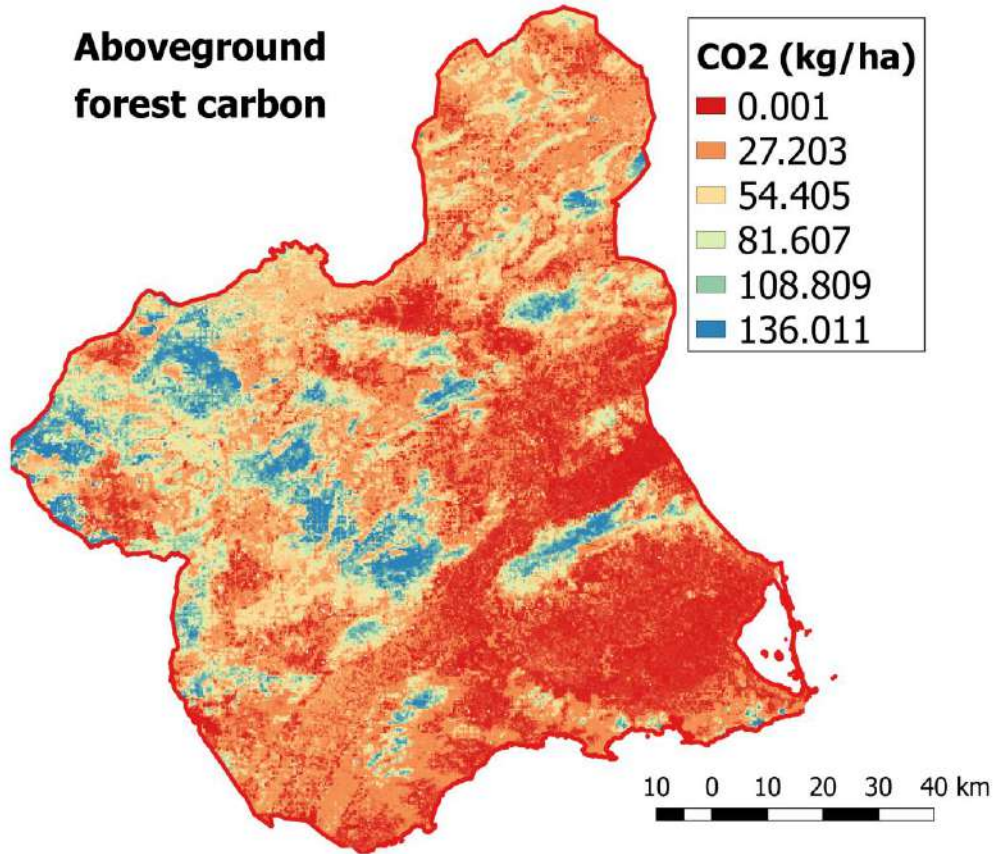
RMSE	R squared	Accuracy	RMSE val
24.45	0.69	0.70	25.38

Results

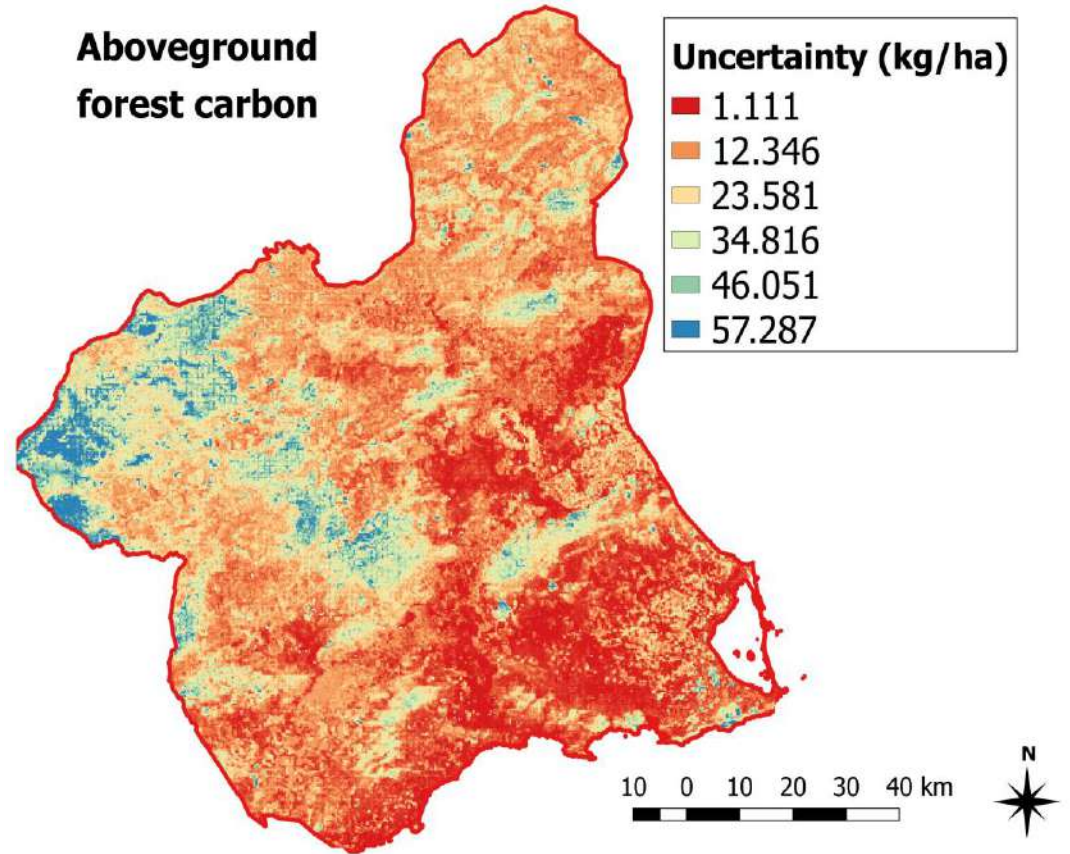
Spatial interpolation



Aboveground forest carbon



Aboveground forest carbon

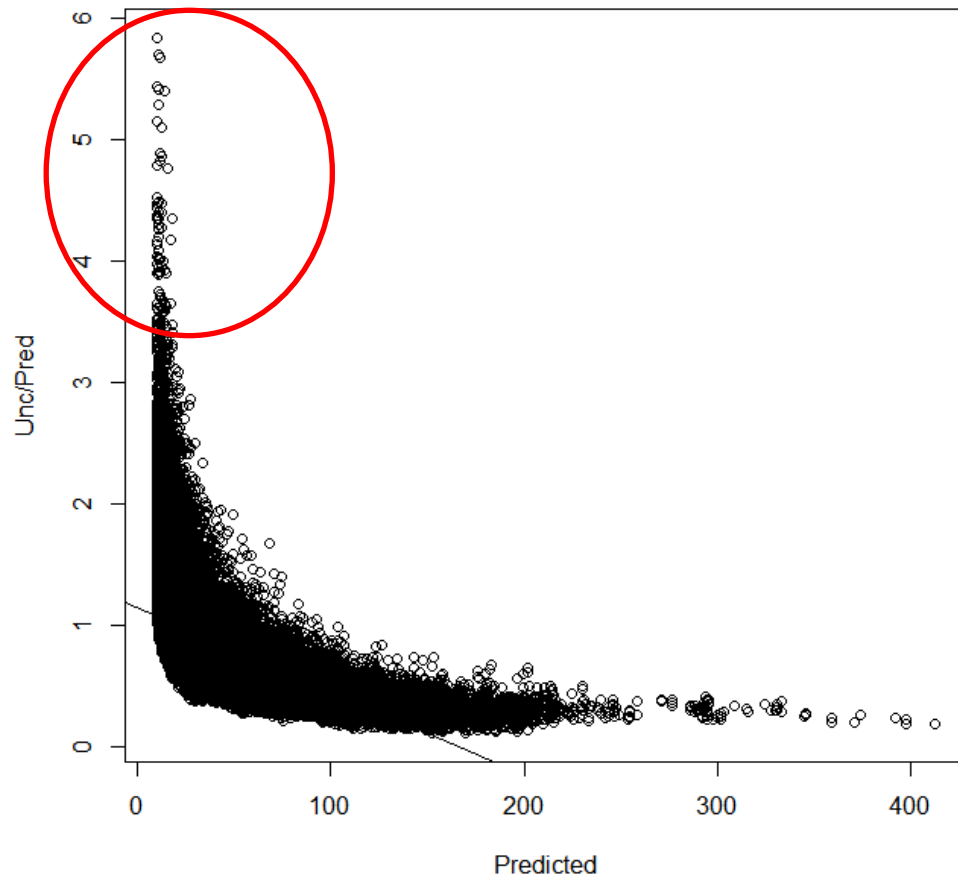


Results

Associated uncertainty



Carbon vegetation

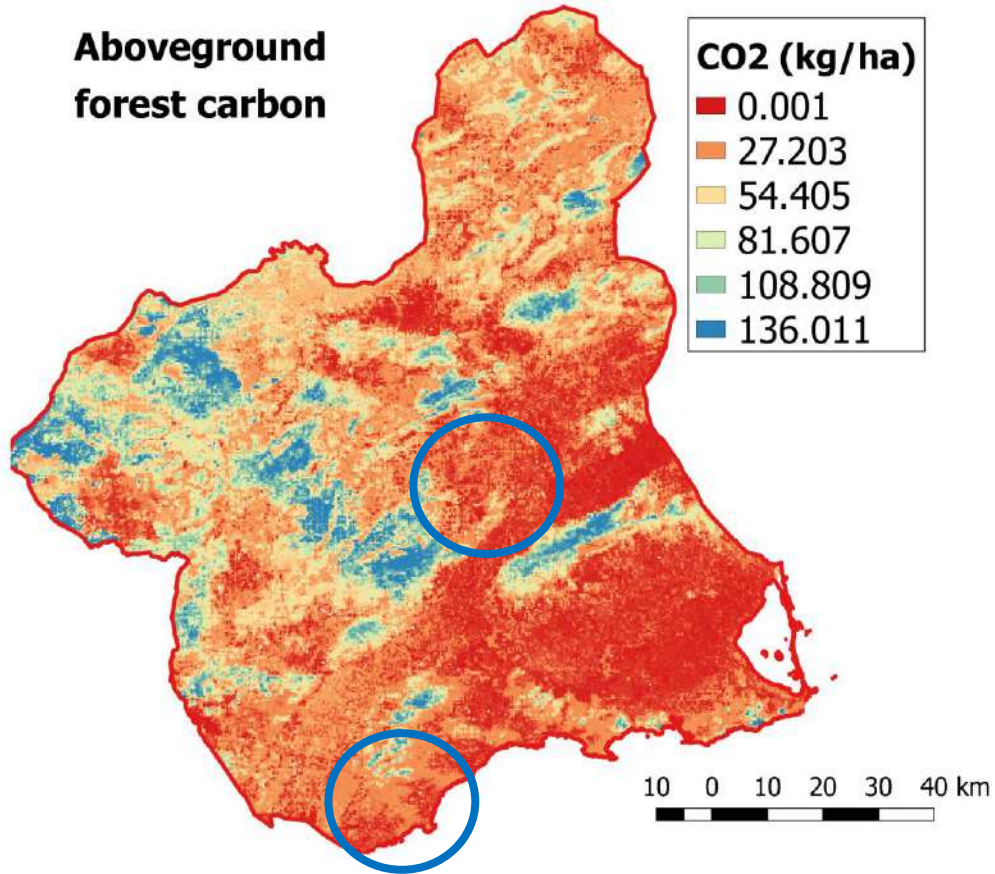


Results

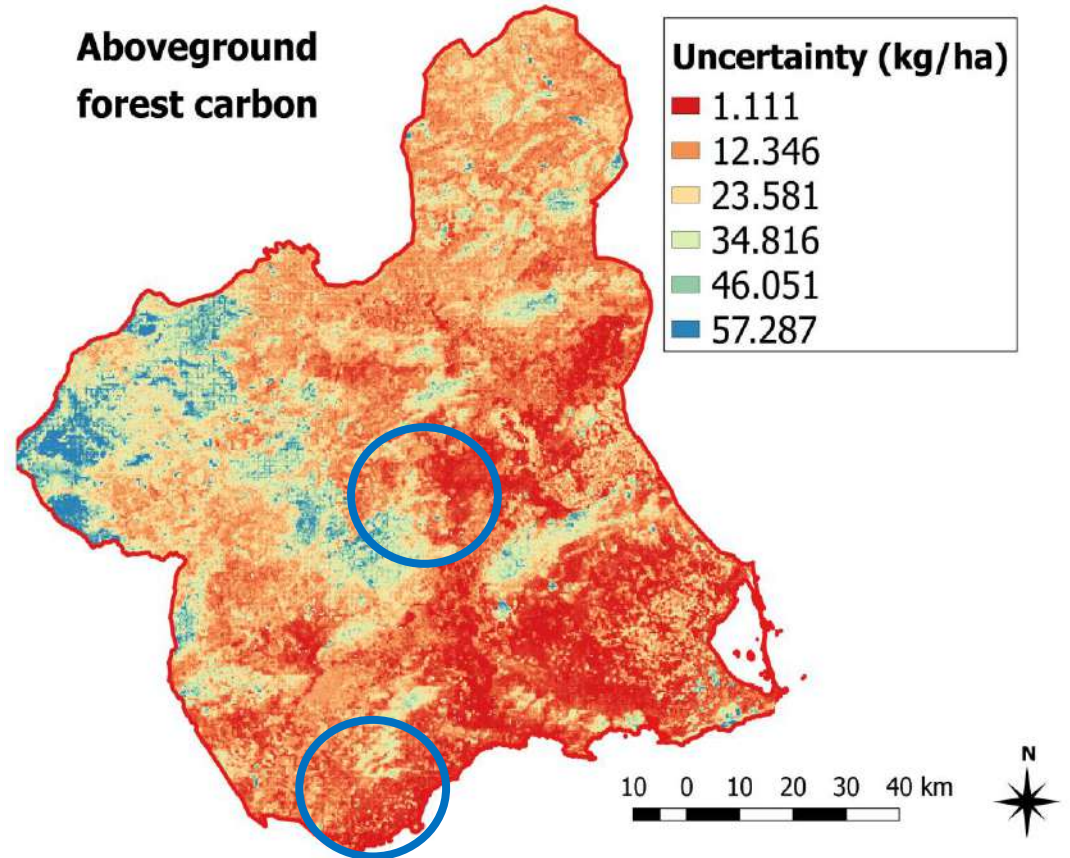
Spatial interpolation



Aboveground forest carbon



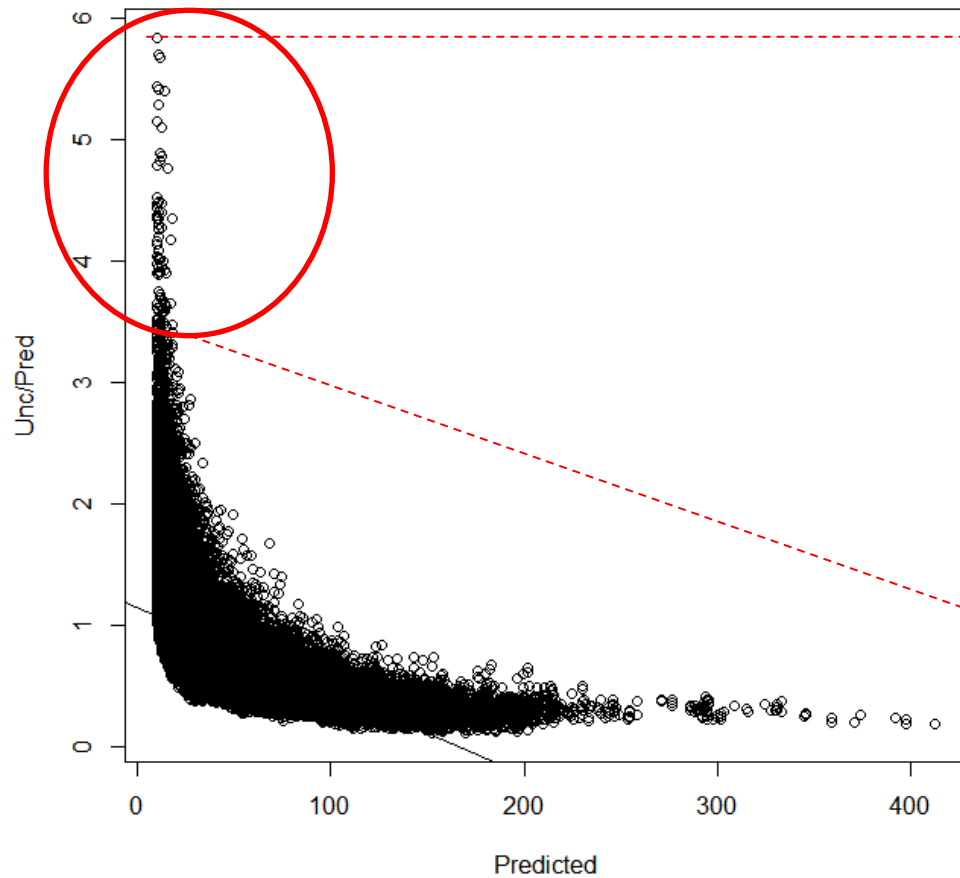
Aboveground forest carbon



Results

Associated uncertainty

Carbon vegetation



A close-up photograph of a tree trunk cross-section, showing concentric growth rings in shades of brown and tan. The texture is rough and natural.

Discussion and conclusions



Discussion and conclusions



- ✓ Close relationship between:

Aboveground forest carbon
(LiDAR)



Vegetation characterization
(Remote sensing)

NDVI

(Mean seasonal profile in
Mediterranean ecosystem:
max in spring and autumn)

LAI

(Thickness of the
vegetation cover.
Essential Climate Variable)



- ✓ Highest uncertainty associated with low predicted carbon values, it may be associated with the accuracy of the general allometric equation obtained from literature.
- ✓ Statistically-robust model; feasible and cost-effective approach for monitoring nationwide (moderate spatial resolution) and easily updated (high temporal availability)

A close-up photograph of a tree trunk cross-section, showing concentric growth rings in shades of brown and tan. The texture is rough and natural. The text "Future step" is overlaid in a bright green, sans-serif font.

Future step



Expectations



- ✓ Copernicus global products with high temporal and spatial resolution, applicable for global, national and regional scale for forest inventories, planning and management.

- ✓ According to our projects, implementation of products for quantification, assessment and monitoring at different scale approaches:
 - Afforestation /Deforestation
 - Fuel types
 - Burned areas
 - Changes detection in ecosystems

- ✓ Combination of different sensors:
 - Radar/optical
 - LiDAR/optical

Technology: Field instrument

Applications

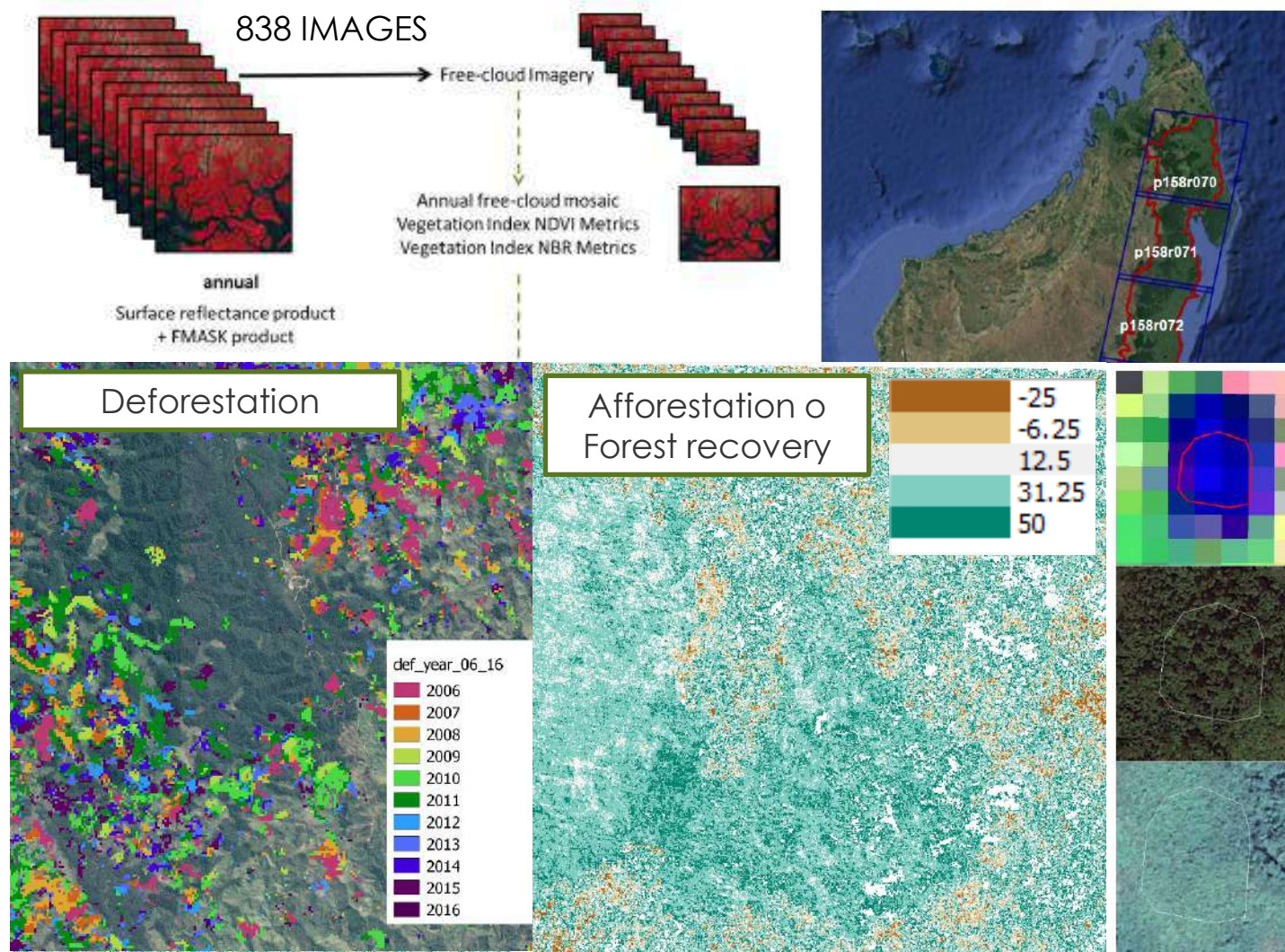
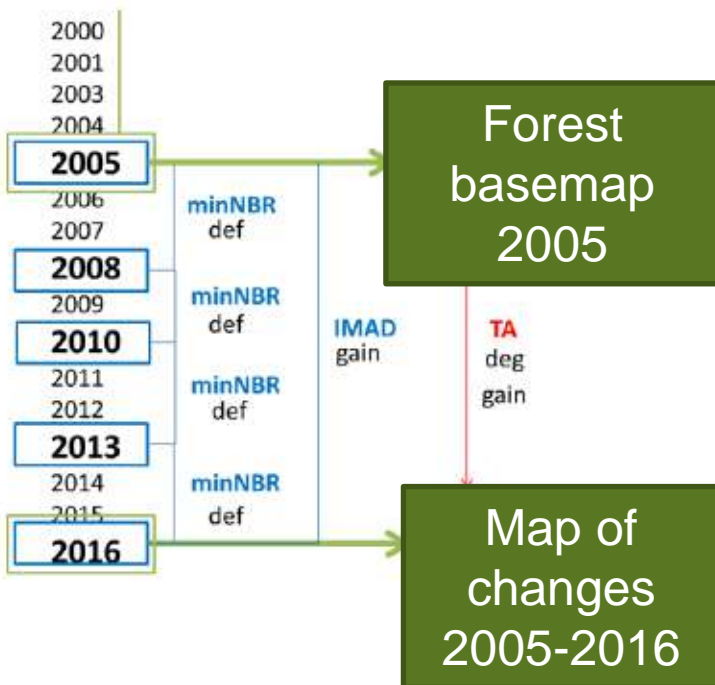
- ✓ Forest management:
 - Forest survey
 - Plots measurements
 - Timber volume assessment



REDD+ PROGRAM: Madagascar



Carbon losses by **degradation and deforestation** vs.
Carbon gains by **afforestation and forest recovery**





Remote Sensing of Environment 187 (2016) 267–280



Contents lists available at ScienceDirect

Remote Sensing of Environment

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



Generation of high-resolution fuel model maps from discrete airborne laser scanner and Landsat-8 OLI: A low-cost and highly updated methodology for large areas



Eva Marino ^{a,*}, Pedro Ranz ^a, José Luis Tomé ^a, Miguel Ángel Noriega ^a, Jessica Esteban ^a, Javier Madrigal ^{b,c}

^a AGRESTA Sociedad Cooperativa, c/Duque de Fernán Nuñez 2, 28012 Madrid, Spain

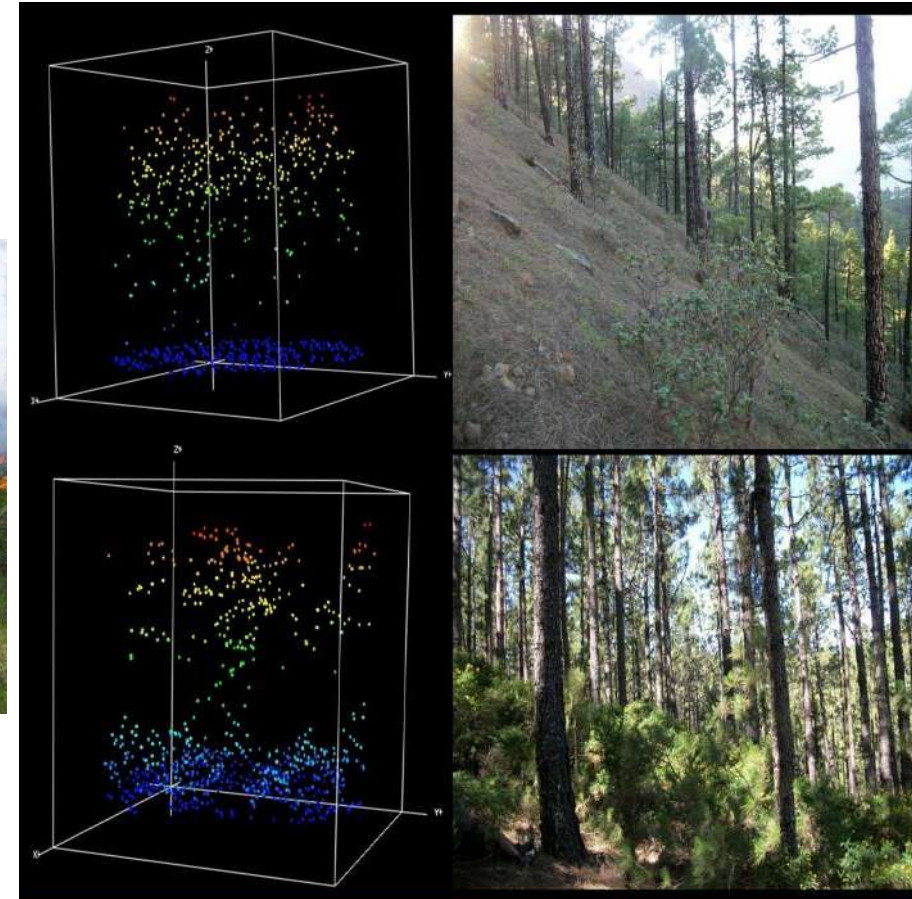
^b INIA, Forest Research Centre, Department of Silviculture and Forest Management, Forest Fire Laboratory, Crta. A Coruña Km 7.5, 28040 Madrid, Spain

^c Sustainable Forest Management Institute UVA-INIA, Crta. A Coruña Km 7.5, 28040 Madrid, Spain

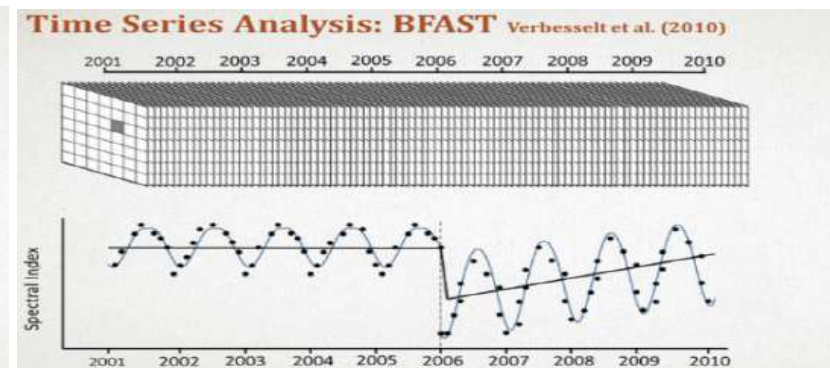
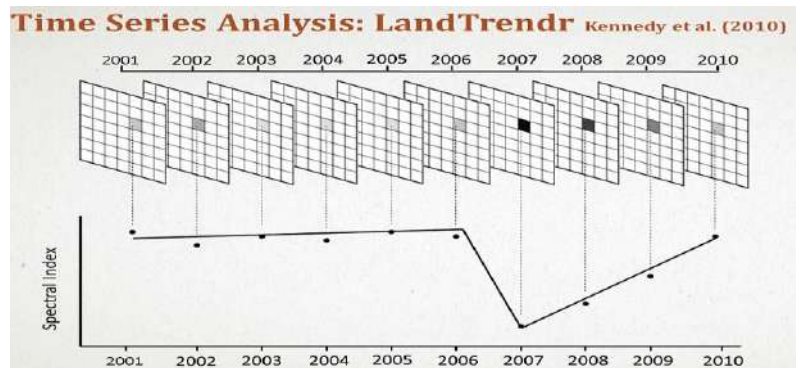
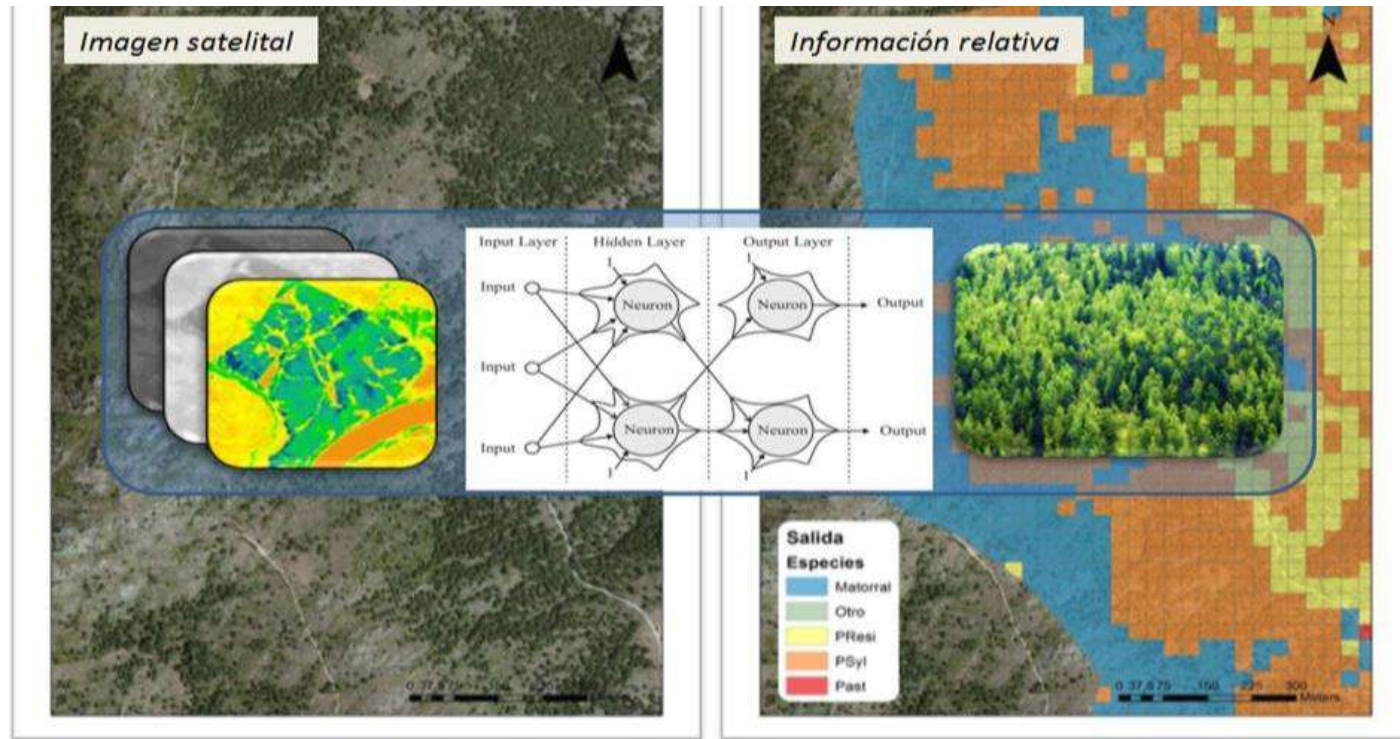
FIRE FOREST

GEPRIF PROYECT (RTA-INIA)

«Integrated evaluation
fire forest»



Artificial neural networks (ANNs)



CARBON PROJECT

- Project: SUPPORT FOR THE JUSTIFICATION AND QUANTIFICATION OF GHG EMISSIONS FROM FOREST DEGRADATION AND GHG EMISSIONS FROM ENHANCEMENTS OF CARBON STOCKS OF A PROPOSED EMISSION REDUCTION PROGRAM IN MADAGASCAR
- Country: **Madagascar**
- January 2017 – in Process
- Partner: ONG FANC (Manondroala), Pôle Carto
- Client: World Bank



Collaborating centers networking



UNIVERSIDAD DE CÓRDOBA





Thank you very much
for your attention

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