

ESTIMATING THE SEVERITY OF DEFOLIATION CAUSED BY PINE PROCESSIONARY MOTH USING LANDSAT AND UAV IMAGERY

Kaori Otsu¹, Magda Pla², Lluís Brotons^{1,2,3}

¹ UAB, Centre for Ecological Research and Forestry Applications

² InForest JRU

³ Spanish National Research Council

JORNADA "OBSERVACIÓ DE LA TERRA I ESPAI FORESTAL, EINES DE DIAGNÒSTIC

ICGC, 8 NOVEMBER 2018



Background



Defoliator, pine processionary moth (PPM), since 1990s in the Mediterranean region



Expansion of host and insect distribution



Severely infested stands by PPM outbreaks



Increased demands for forest monitoring

PPM Monitoring



Generalitat
de Catalunya



❖ Annual field survey
and mapping in
Catalonia

- From 2010 by Rural Agents (Generalitat de Catalunya)

❖ Data entry

- Severity levels 1-4
- Tree species
- Elevation and orientation

❖ Outbreak in winter
over 2015-2016

Study Area (2015)



Study Area (2016)



❖ Severely affected areas with level 4

- 6800 ha near Solsona, Catalonia
- Elevation at 600-1100 m
- Mediterranean continental climate
- *Pinus nigra*, *P. sylvestris*

❖ Sketch mapping concerns

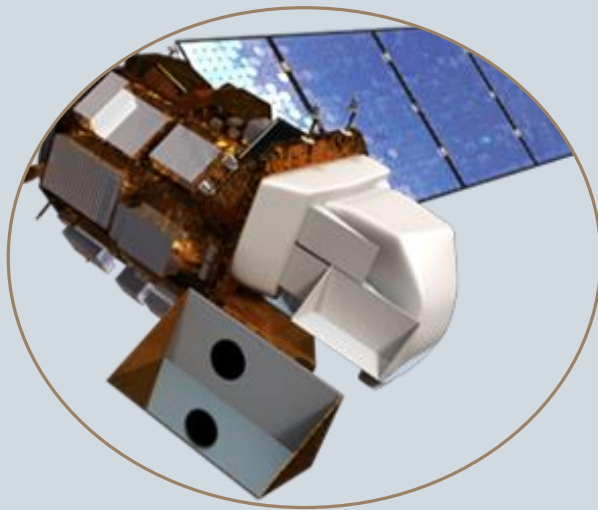
- Qualitative classification
- Coarse spatial resolution
- Inclusion of non-forest stands

Study Area (2017)

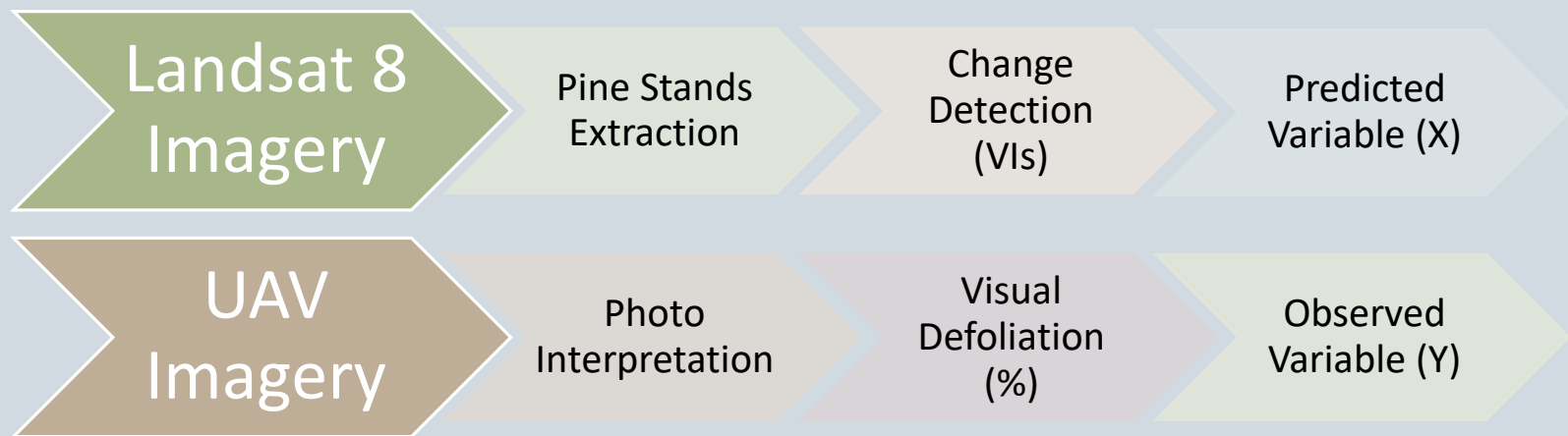


Objectives

- ❑ To quantify the severity of defoliation by the recent PPM outbreak with Landsat-based vegetation indices (VIs)
- ❑ To calibrate the VIs with defoliation degrees interpreted by unmanned aerial vehicle (UAV) imagery



Methodology Workflow



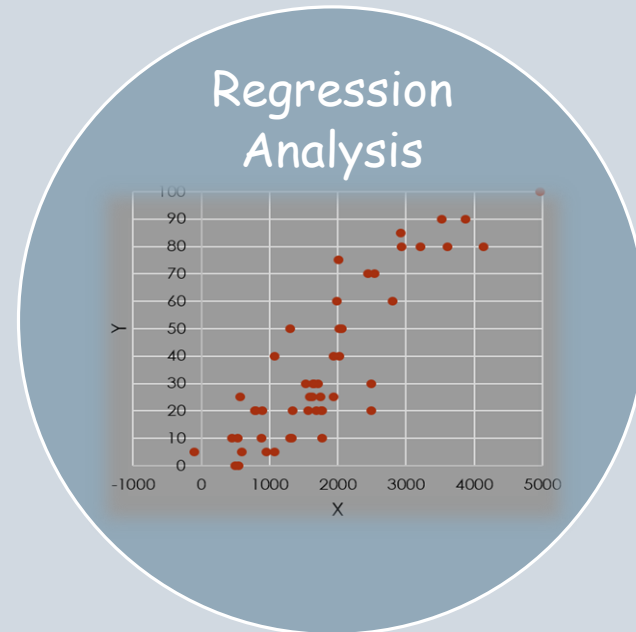
Methodology Workflow

X
Predicted
Vegetation
Indices



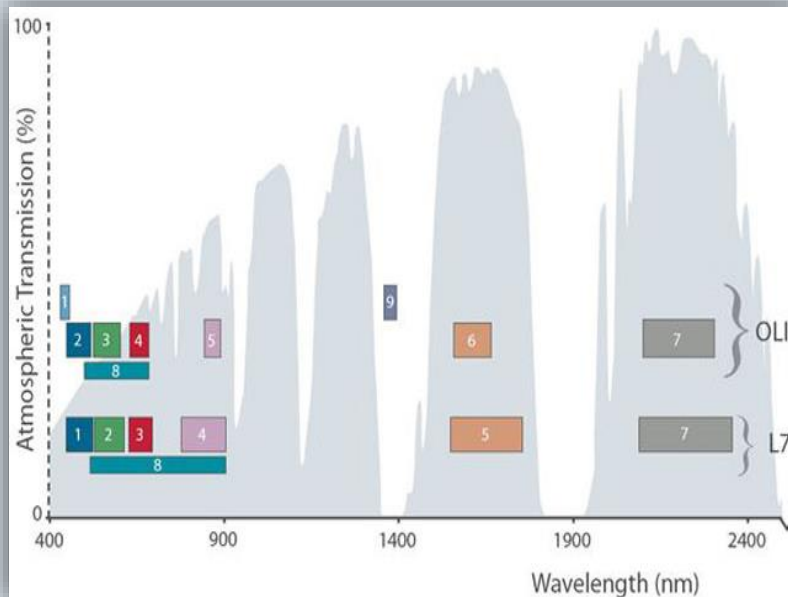
50

y
Observed
Defoliation
Degrees



X = d(Vegetation Index)

Multispectral Bands (OLI)

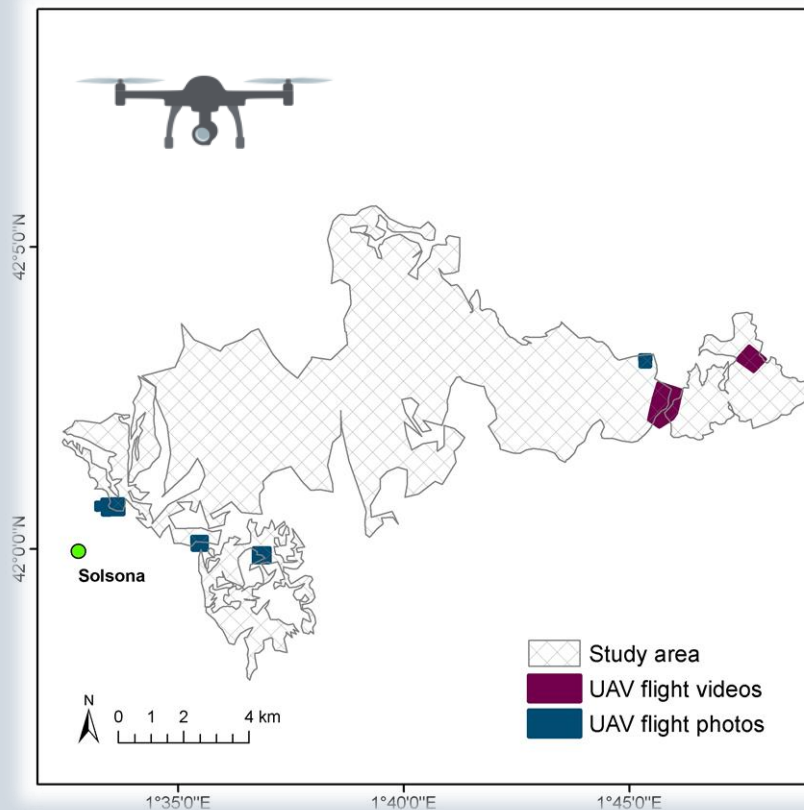


Landsat 8 Vegetation Indices

| Index | Acronym | Formula |
|--|---------|-------------------------|
| Middle Infrared Wavelengths | MID | $b6 + b7$ |
| Moisture Stress Index | MSI | $b6 / b5$ |
| Normalized Difference Moisture Index | NDMI | $(b5 - b6) / (b5 + b6)$ |
| Normalized Difference Vegetation Index | NDVI | $(b5 - b4) / (b5 + b4)$ |
| Normalized Burn Ratio | NBR | $(b5 - b7) / (b5 + b7)$ |
| Change detection in VI | dVI | $VI(2015) - VI(2016)$ |

(b4 = Red, b5 = Near Infrared, b6 = Shortwave Infrared 1, b7 = Shortwave Infrared 2)

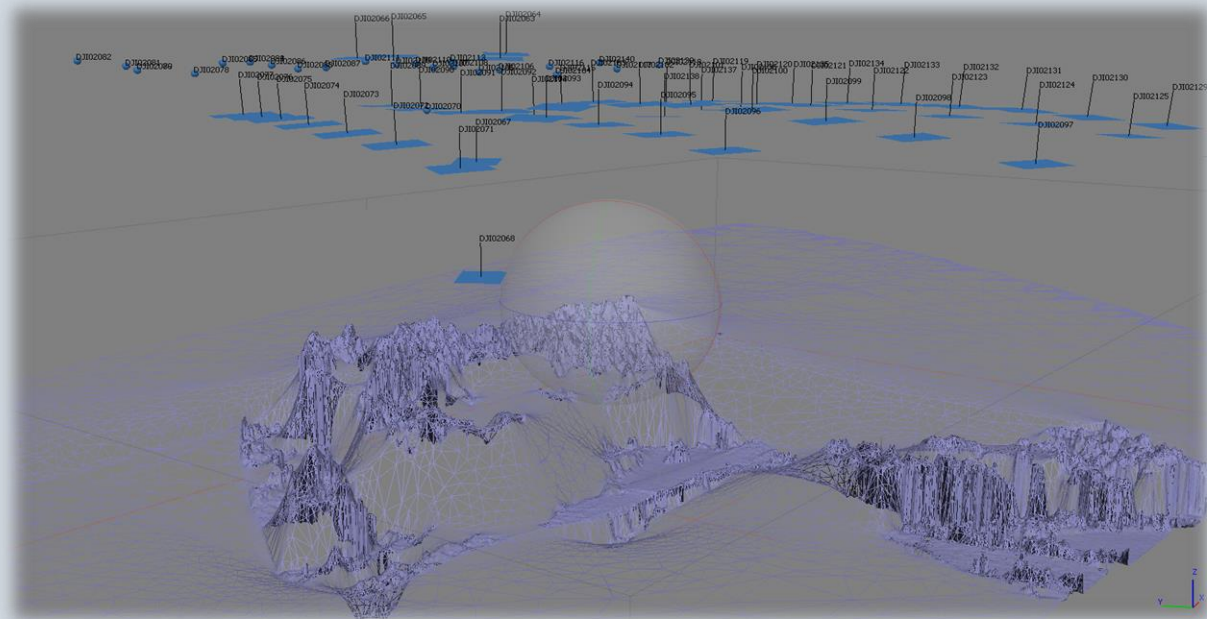
Y = UAV Images



- ❖ RGB camera
 - DJI Phantom 2 Vision FC200
- ❖ UAV flight
 - Altitude 50-100 m
 - 7 surveys in winter 2016 (post-outbreak)
 - Image processing for orthomosaic
 - Ground resolution 2.0-3.5 cm

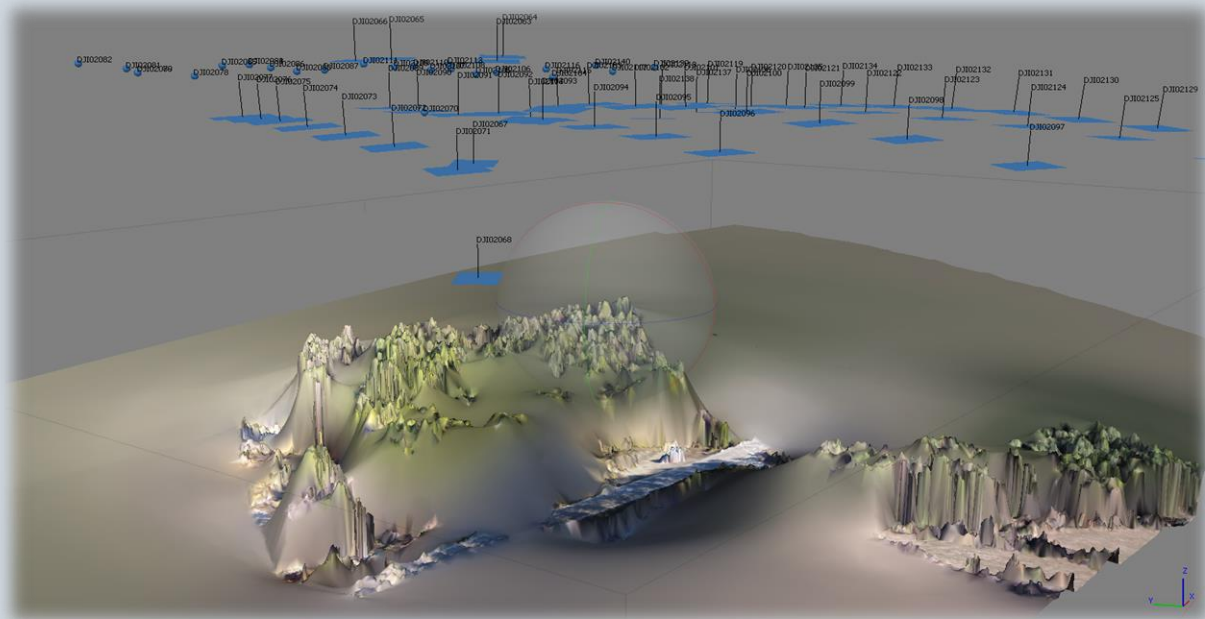
$Y = \text{UAV Images}$

3D model by PhotoScan



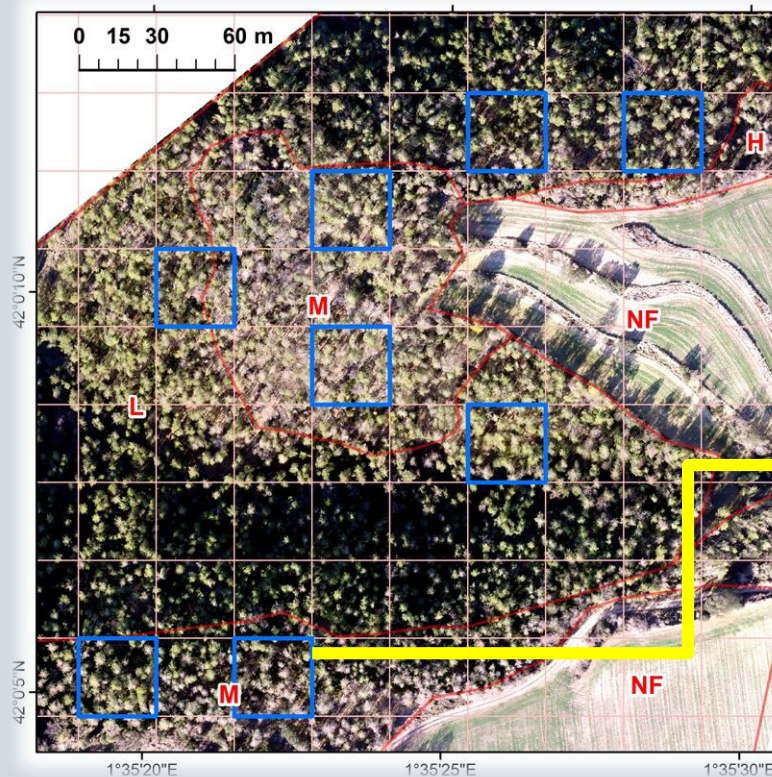
$Y = \text{UAV Images}$

3D model by PhotoScan



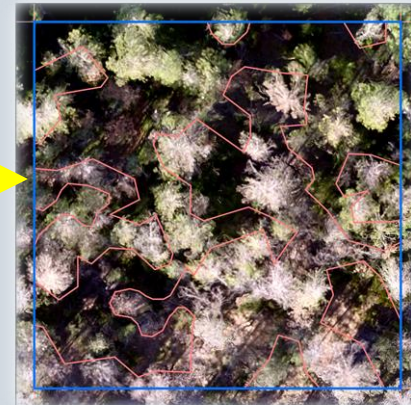
Y = UAV Images

Orthomosaic

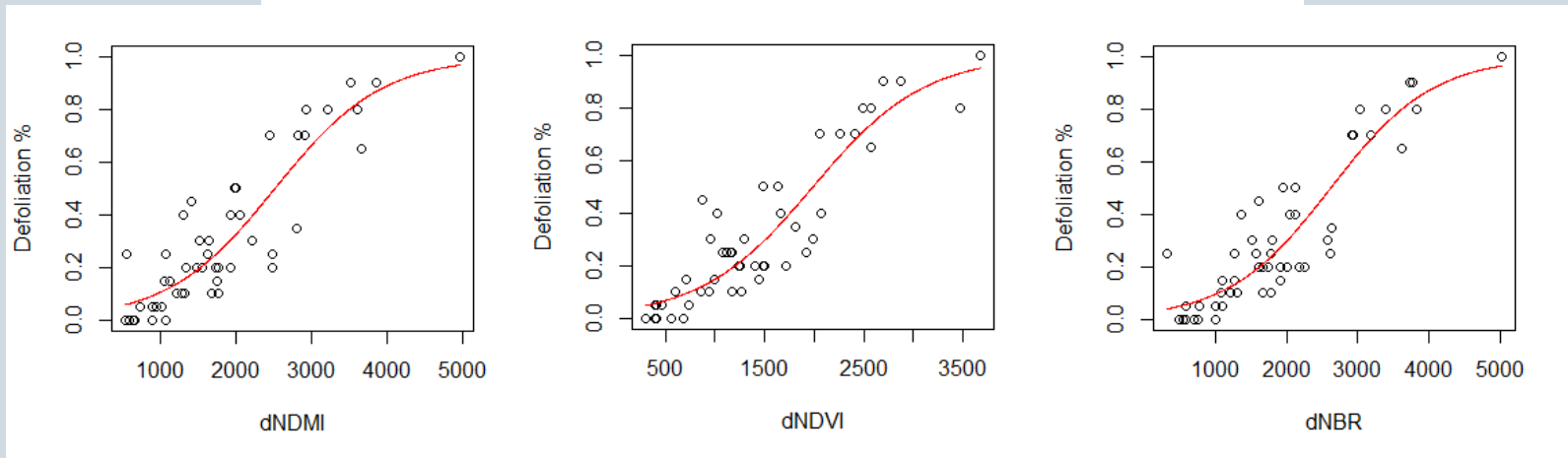
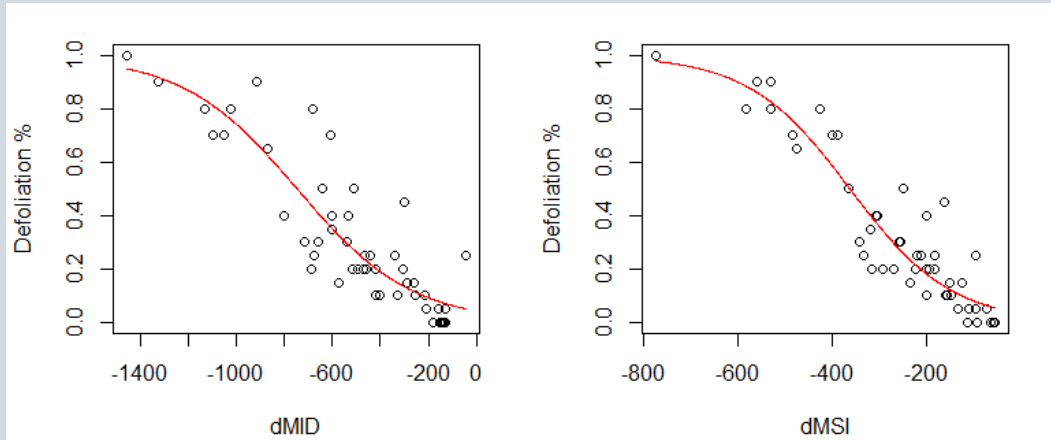


Visual interpretation

| Severity | Defoliation (%) | Samples |
|----------|-----------------|---------|
| Nil | 0 - 5 | 10 |
| Low | 10 - 30 | 23 |
| Medium | 35 - 65 | 8 |
| High | 70 - 100 | 9 |



Regression Analysis



Regression Analysis

Logistic Regression Models

| Index | Equation | R ² (McFadden's) |
|-------|--|--------------------------------|
| dMID | $Y = \frac{1}{1 + e^{-(-3.1299111 - 0.0041928X)}}$ | 0.740 |
| dMSI | $Y = \frac{1}{1 + e^{-(-3.3570352 - 0.0092755X)}}$ | 0.815 |
| dNDMI | $Y = \frac{1}{1 + e^{-(-3.5552389 + 0.0014107X)}}$ | 0.749 |
| dNDVI | $Y = \frac{1}{1 + e^{-(-3.509468 + 0.001767X)}}$ | 0.787 |
| dNBR | $Y = \frac{1}{1 + e^{-(-3.6323329 - 0.0013874X)}}$ | 0.776 |

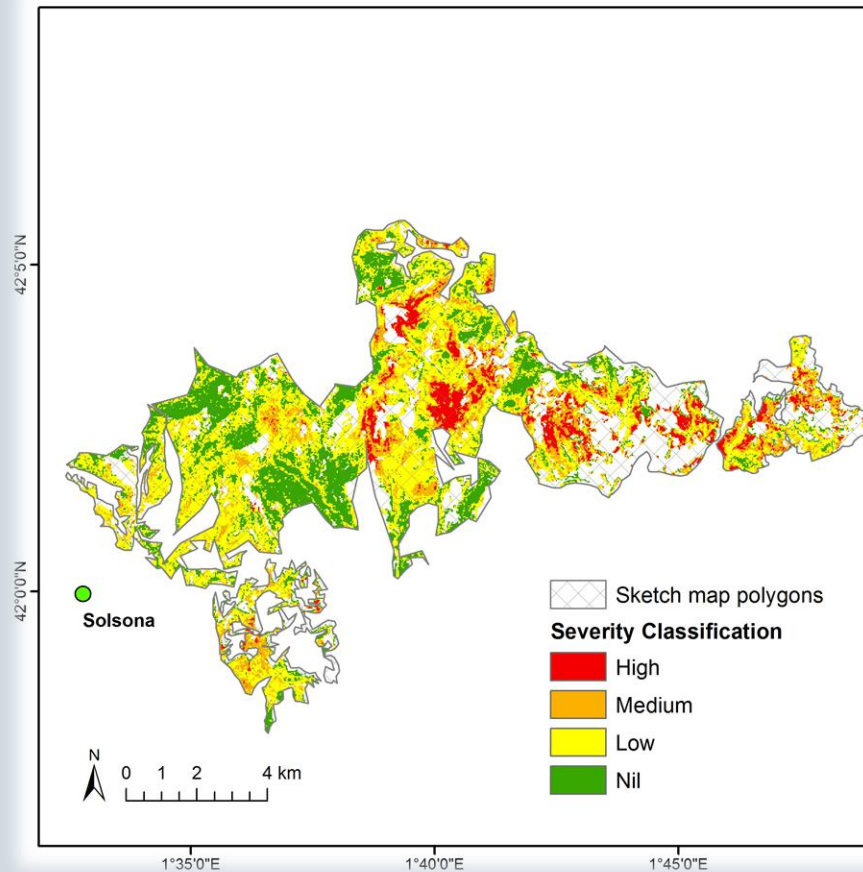
$$Y = \frac{1}{1 + e^{-(a+bX)}}$$

Threshold Classification

| X | Y = Defoliation (%) | | |
|-------|---------------------|----------------|--------------|
| | Low (10) | Medium (35) | High (70) |
| dMID | -222 | -599 | -949 |
| dMSI | -125 | -295 | -453 |
| dNDMI | 963 | 2081 | 3121 |
| dNDVI | 743 | 1636 | 2466 |
| dNBR | 1034 | 2172 | 3229 |

$$X = \frac{\ln\left(\frac{Y}{1-Y}\right) - a}{b}$$

Predicted Defoliation Map



Classification Accuracy

Confusion Matrix

| Class | | Predicted (Landsat 8) | | | | | Producer's Accuracy |
|----------------|-----------------|-----------------------|-----------|----------|----------|-----------|---------------------|
| | | Nil | Low | Medium | High | Total | |
| Observed (UAV) | Nil | 9 | 1 | 0 | 0 | 10 | 0.90 |
| | Low | 2 | 17 | 4 | 0 | 23 | 0.74 |
| | Medium | 0 | 3 | 4 | 1 | 8 | 0.50 |
| | High | 0 | 0 | 3 | 6 | 9 | 0.67 |
| | Total | 11 | 21 | 11 | 7 | 50 | |
| | User's Accuracy | 0.82 | 0.81 | 0.36 | 0.86 | | 0.72 |

Discussions

❖ Robust VI for defoliation

- Moisture Stress Index
- Normalized Difference Vegetation Index

❖ Classification accuracy

- Sample size
- Non-parametric algorithms
- Spectral bands of dVIs

❖ Data resolution trade-off = spatial + temporal + spectral

- Ground & aerial sketch mapping
- Spaceborne Landsat
- Airborne UAV

Conclusions and Future Study

Validation

Additional UAV images in a new study area may increase the robustness of VI models.

Ground Truth

The UAV technology holds great potential for cost-effectively monitoring forest health.

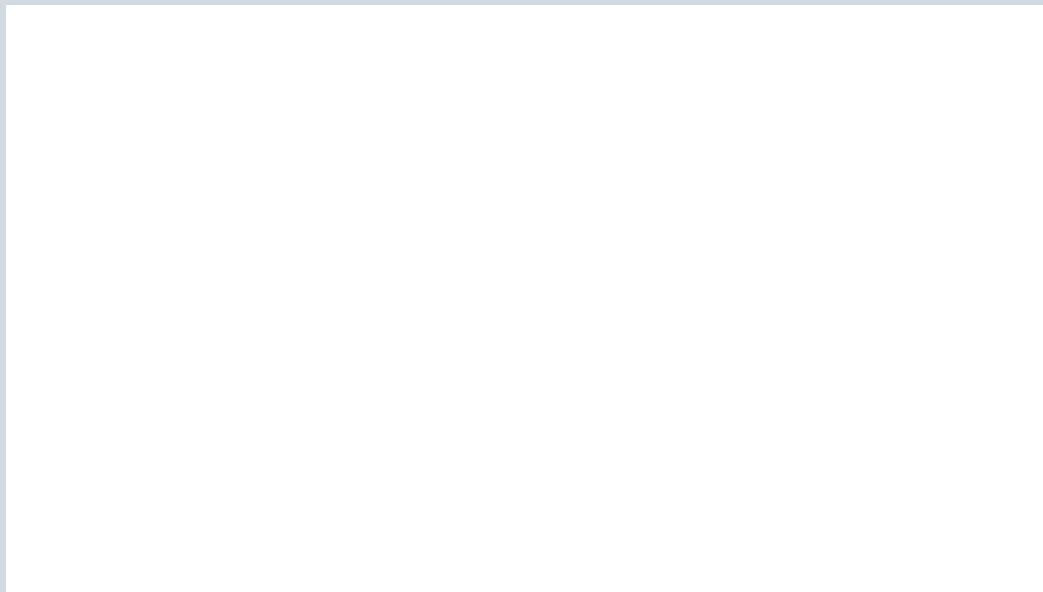
Ecosystem Service

Combining satellite and UAV data may serve as a tool to provision pest distribution and forest production.

Publication

Otsu, K.; Pla, M.; Vayreda, J.; Brotons, L. Calibrating the Severity of Forest Defoliation by Pine Processionary Moth with Landsat and UAV Imagery. *Sensors* **2018**, *18*, 3278.

<https://www.mdpi.com/1424-8220/18/10/3278/htm>



ALERTAFORESTAL

supported by



Obra Social "la Caixa"

<http://www.alertaforestal.com/alertas/procesionaria>
k.otsu@creaf.uab.cat

PPM Life Cycle (*Thaumetopoea pityocampa*)

