

# Generating fire risk variables from satellite images: fuel and vulnerability

Patricia Oliva

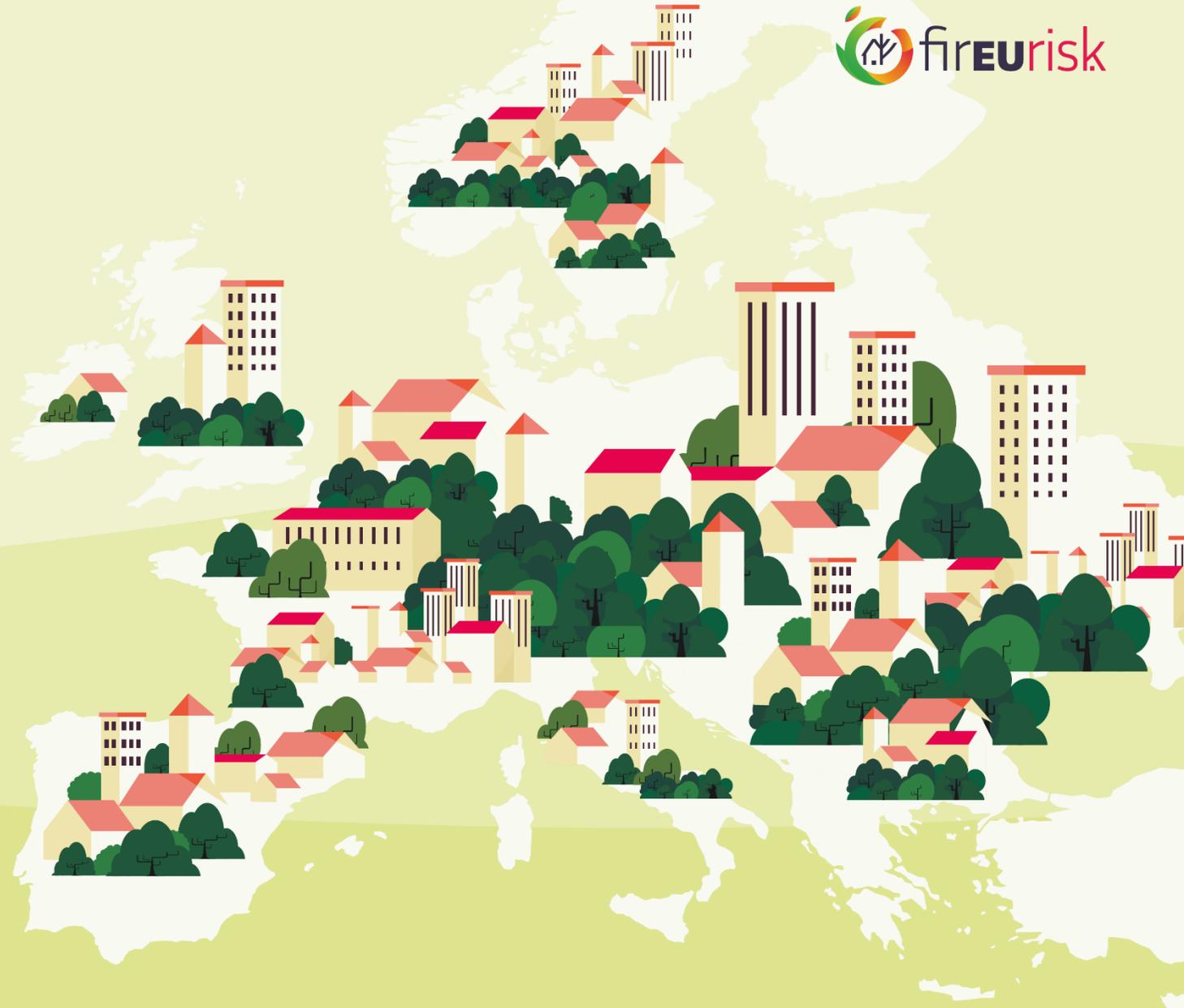
Environmental Remote Sensing Research Group,

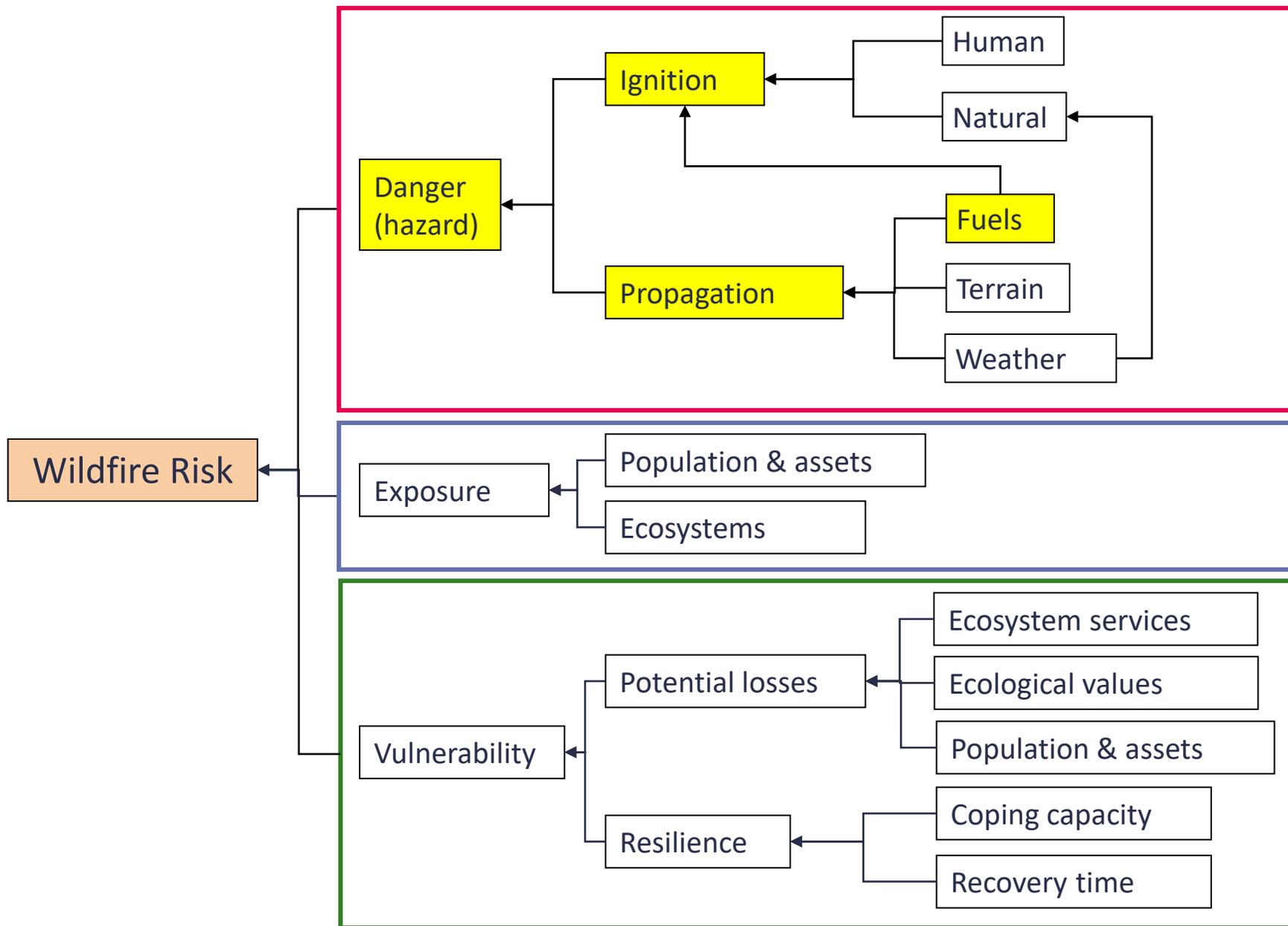
Universidad de Alcalá (Spain)

[Patricia.oliva@uah.es](mailto:Patricia.oliva@uah.es)

# Fuel types and fuel parameters

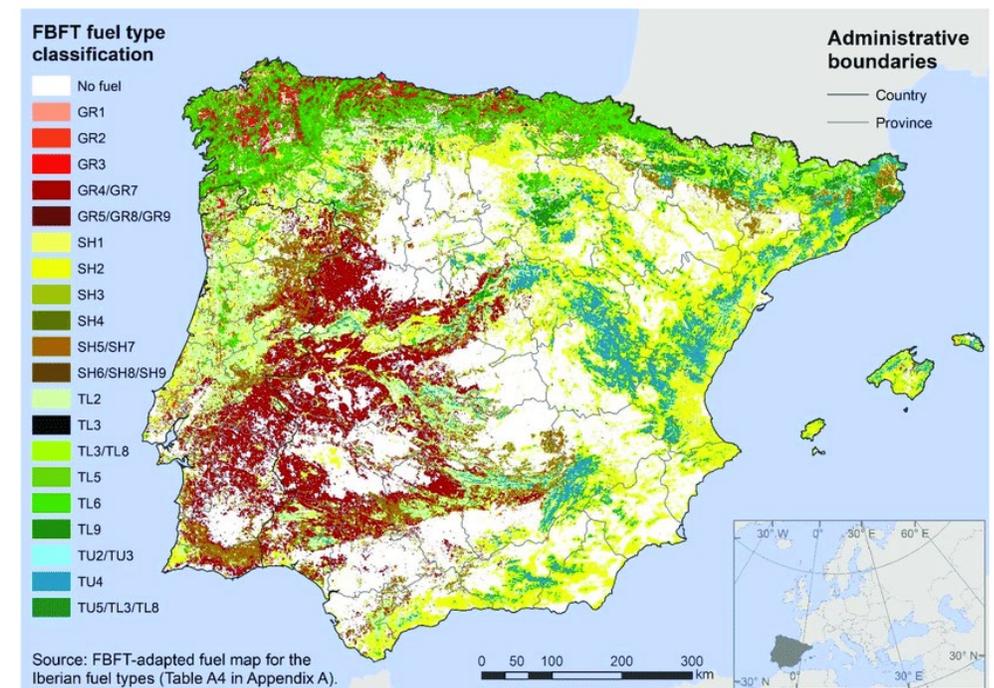
Elena Aragoneses ([e.Aragoneses@uah.es](mailto:e.Aragoneses@uah.es))  
University of Alcalá





# Fuel types and fuel parameters

- **Fuel types:** vegetation with similar fire behaviour (categories)
- **Fuel parameters:** Numerical parameters required to model fire behaviour (height, load, particle size, etcetera).



Aragoneses and Chuvieco (2021): <https://www.mdpi.com/2571-6255/4/3/59>

Fuel Model	Typical Fuel Complex	Fuel Loadings		
		Fuel 1-Hr	Fuel 10-Hr	Fuel 100-Hr
2	Timber (grass and understory)	2	1	0.5
3	Tall grass (2.5 feet)	3.01	0	0
4	Chaparral	5.01	4.01	2
6	Dormant brush, hardwood slash	1.5	2.5	2
7	Southern rough	1.13	1.87	1.5
8	Closed timber litter	1.5	1	2.5
9	Hardwood litter	2.92	0.41	0.15

Kabli et al (2015): <https://www.mdpi.com/1999-4907/6/6/2148/htm>



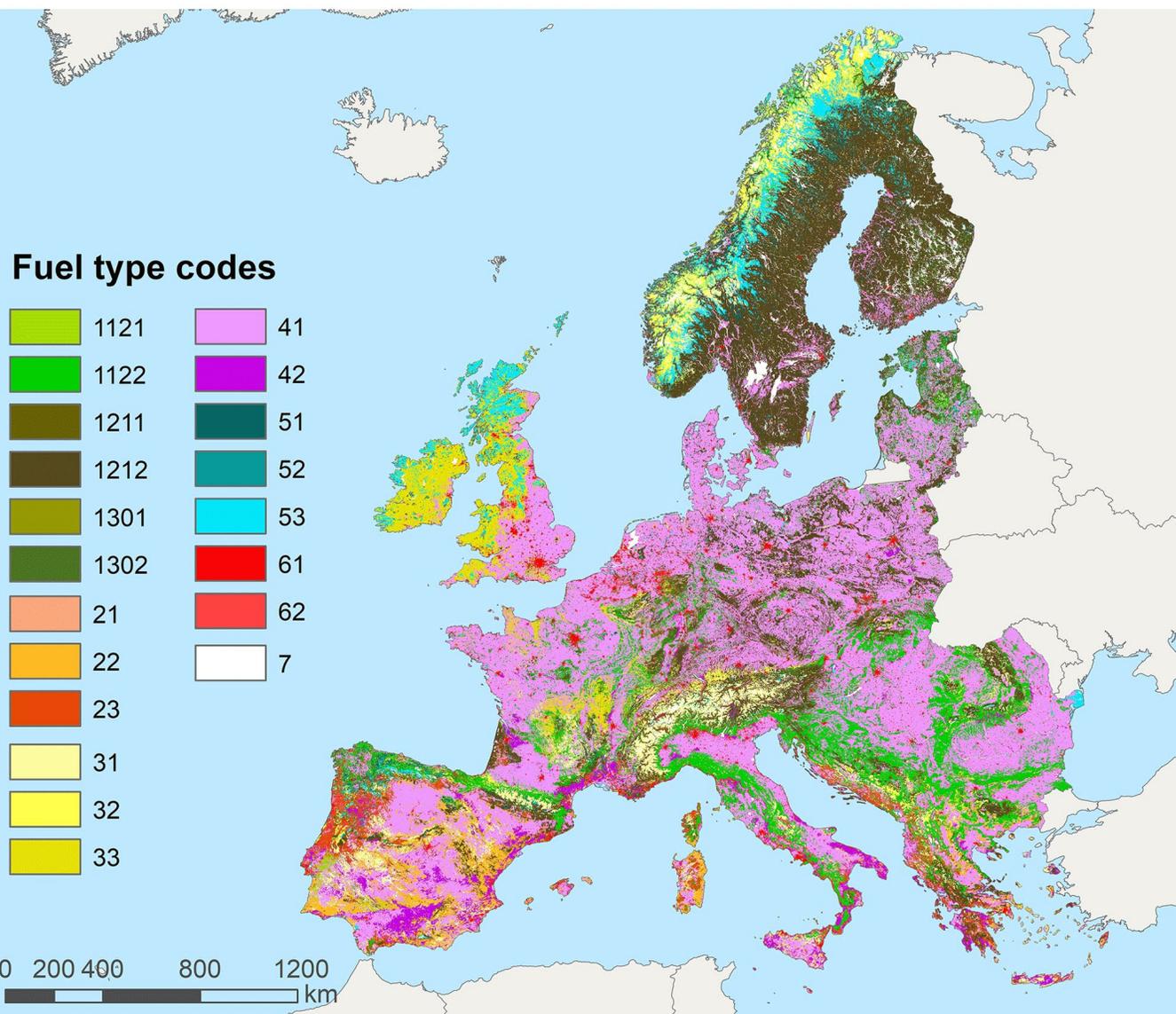
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# European fuel type map

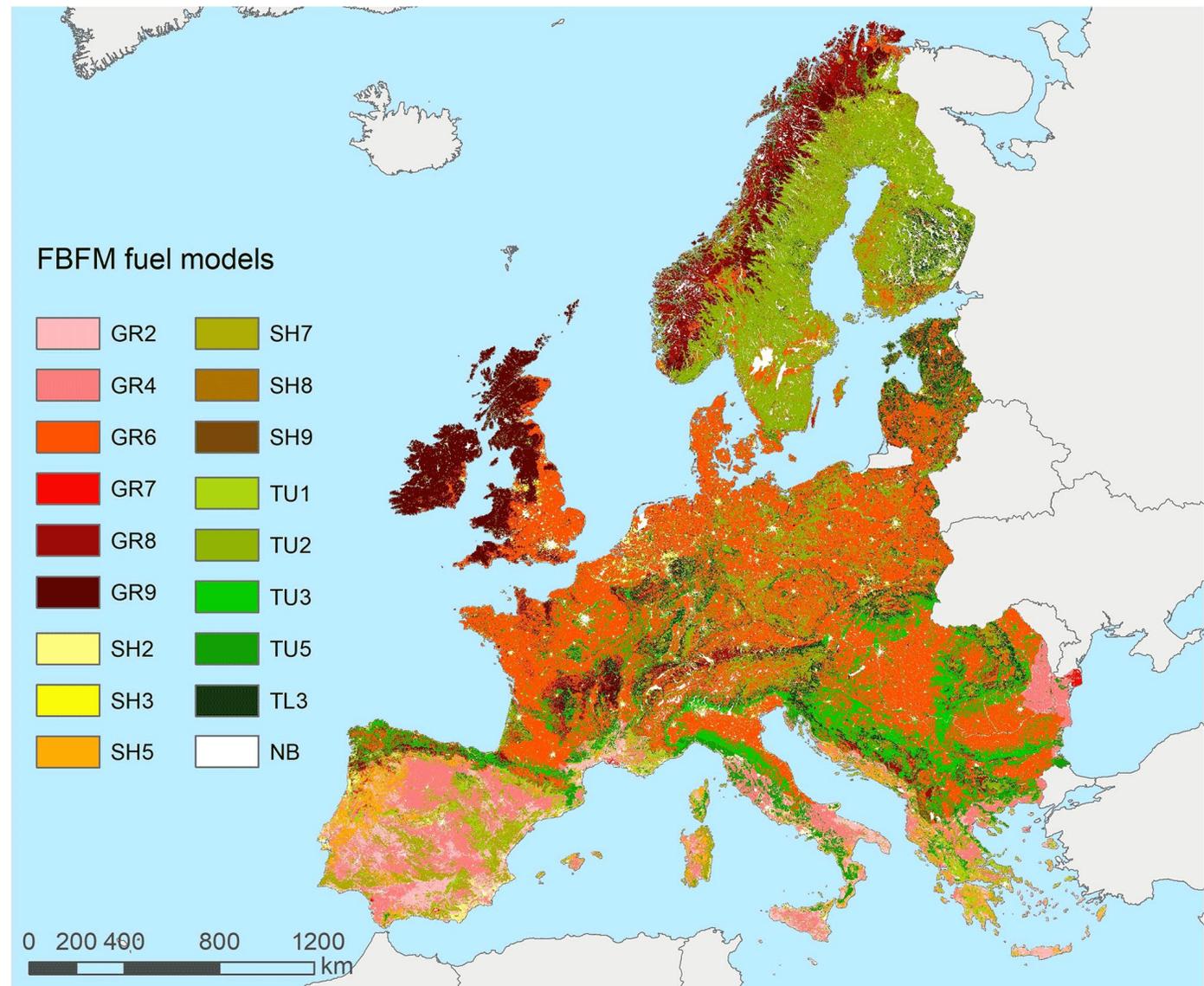


Main fuel type	Leaf type/ Type	Phenology	Fractional cover (%)
1. Forest	11. Broadleaf	111. Evergreen	1111. Open [15-70%] 1112. Closed [70-100%]
		112. Deciduous	1121. Open [15-70%] 1122. Closed [70-100%]
	12. Needleleaf	121. Evergreen	1211. Open [15-70%] 1212. Closed [70-100%]
		122. Deciduous	1221. Open [15-70%] 1222. Closed [70-100%]
	13. Mixed		1301. Open [15-70%] 1302. Closed [70-100%]
	<b>Fuel depth</b>		
2. Shrubland	21. Low [0-0.5m]		Hierarchical
	22. Medium [0.5-1.5m]		
	23. High ( $\geq 1.5m$ )		
3. Grassland	31. Low [0-0.3m]		Europe's diversity Surface and crown fuels
	32. Medium [0.3-0.7m]		
	33. High ( $\geq 0.7m$ )		
<b>Type</b>			
4. Cropland	41. Herbaceous		1 km resolution
	42. Woody (shrub-tree)		
5. Wet and peat/ semi-peat land	51. Tree		
	52. Shrubland		
	53. Grassland		
6. Urban	61. Continuous fabric: urban fabric ( $\geq 80\%$ )		
	62. Discontinuous fabric: vegetation and urban fabric [15-80%]		
7. Nonfuel			

# First assignment of standard fuel models



- **Crosswalk** from FirEUrisk fuel types to FBFM fuel models (Scott and Burgan, 2005)
- Use of **expert knowledge**
- Only **surface fuels**



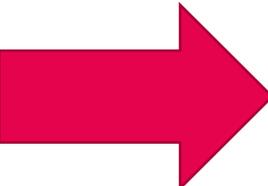
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Aragoneses et al. 2023 <https://doi.org/10.5194/esd-15-1287-2023>

# Crown fuel parameters and used data



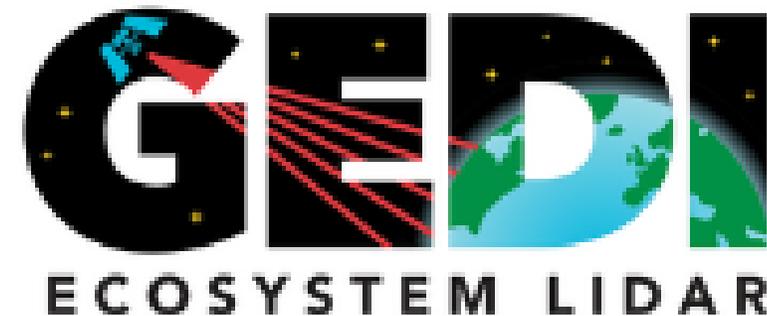
- Forest canopy fuel parameters to be estimated – **useful** for fire behaviour modelling:



<ul style="list-style-type: none"><li>• Canopy mean height</li><li>• Canopy cover</li><li>• Canopy base height</li></ul>	<b>METHOD 1:</b> Derived metrics from GEDI LiDAR
<ul style="list-style-type: none"><li>• Canopy fuel load</li><li>• Canopy bulk density</li></ul>	<b>METHOD 2:</b> Raw GEDI LiDAR waveforms

**MORE COMPLEX PARAMETERS:**  
Vertical distribution of fuel

- Forest inventory data
- Discrete waveforms: airborne LiDAR → calibration of models
- Full-waveforms (continuous pulse of energy): satellite LiDAR (GEDI mission) → interpolation for wall-to-wall maps
  - Almost global coverage



<https://gedi.umd.edu/>

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# Method 1 : Calibration of GEDI-based models from GEDI metrics



Forest inventory data and input airborne LiDAR (ALS)

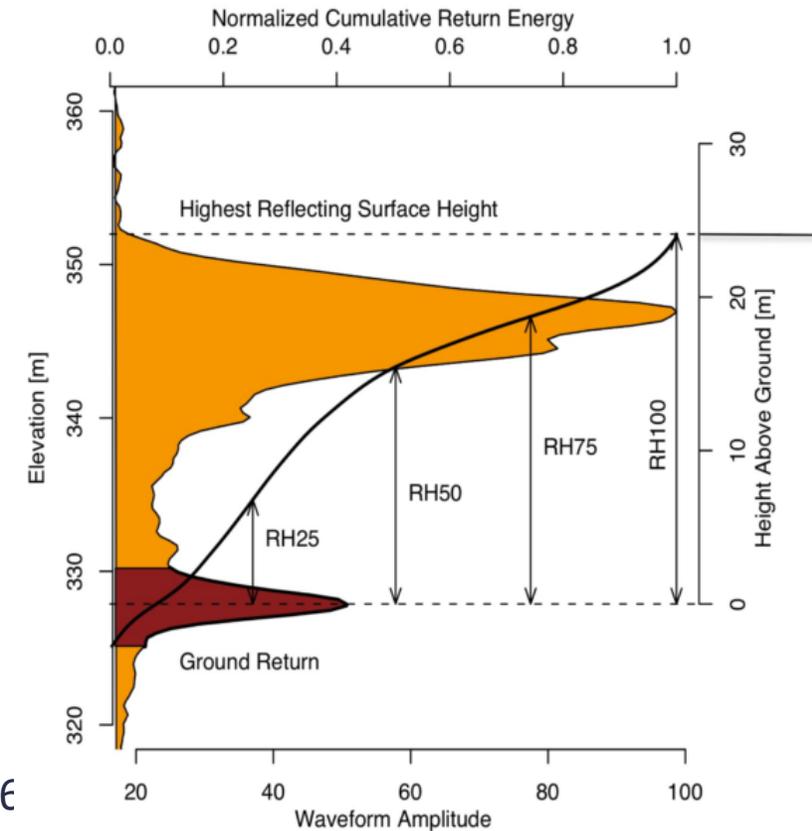
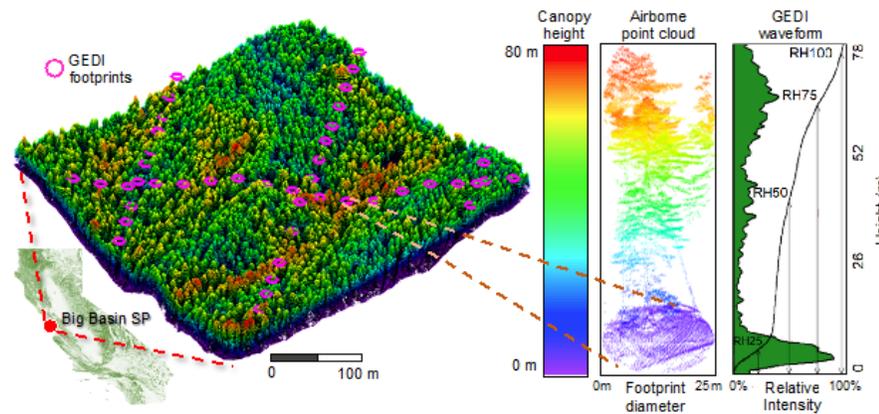
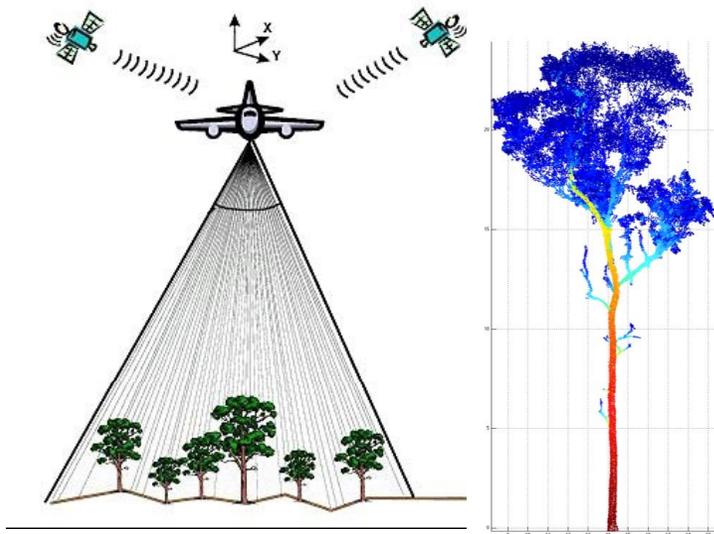
Spain -> 4 areas  
Germany -> 2 areas  
Slovenia -> 1 area

Clip by forest plots

Simulate satellite GEDI-like waveforms

Simulation -> Extract metrics (RHs, ground, canopy cover,...)

GEDI-based model using reference data from ALS and forest inventory



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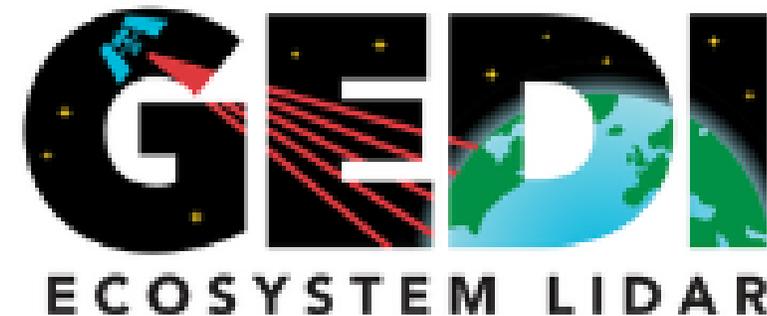
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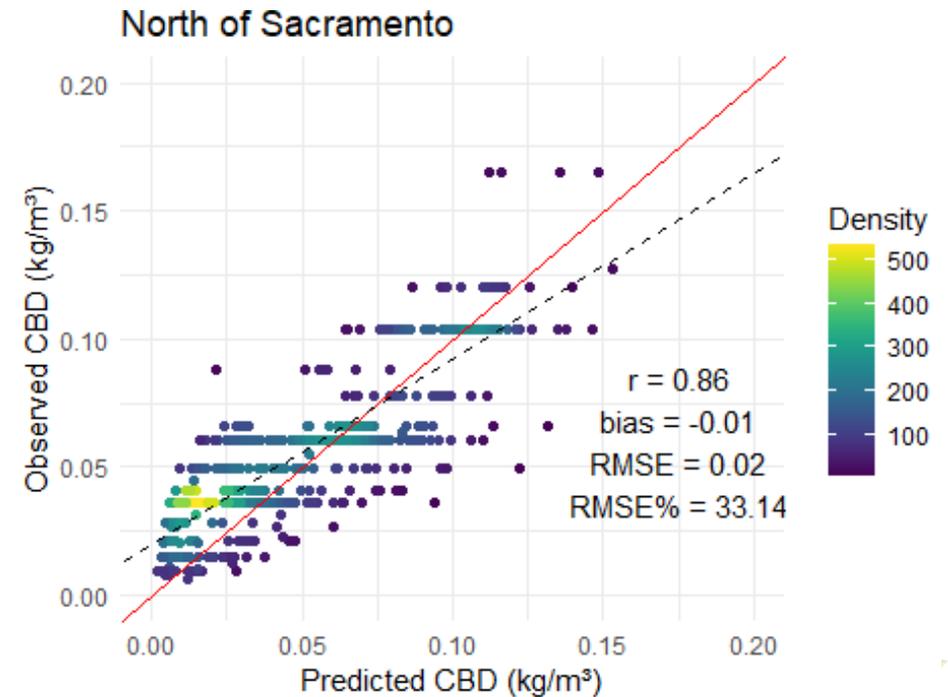
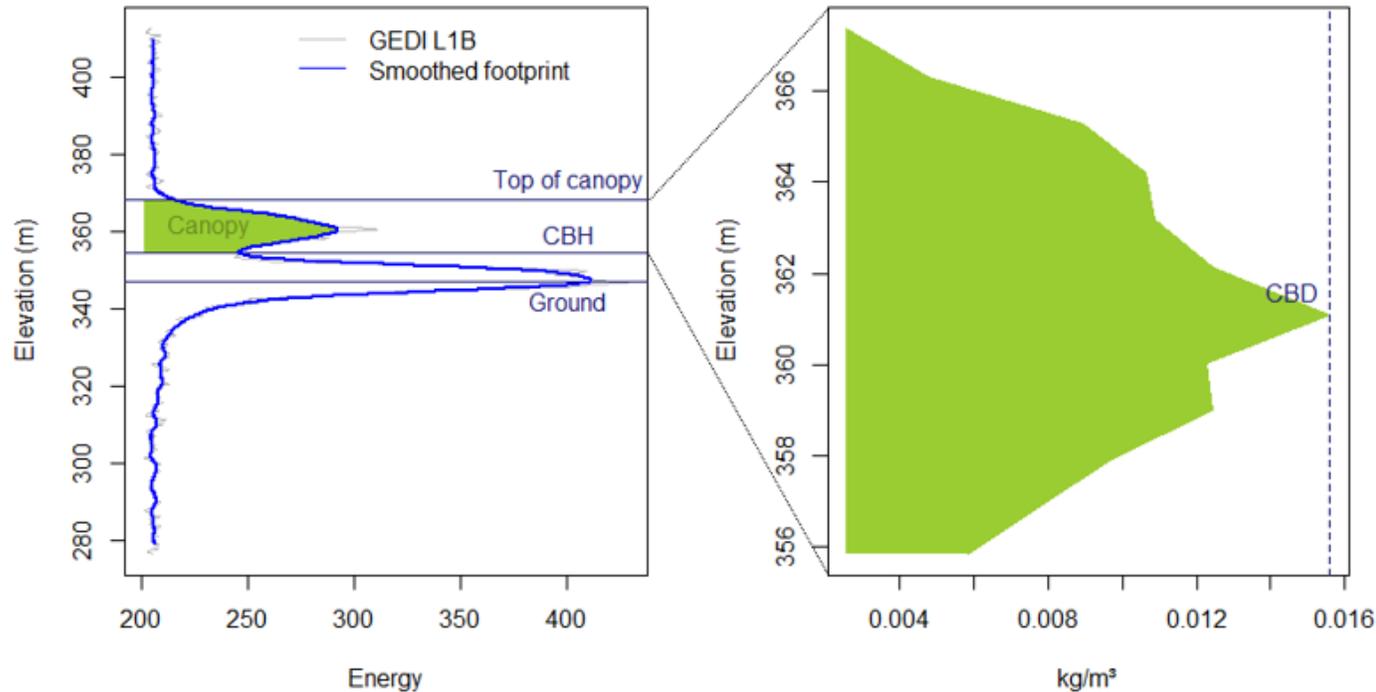
# Method 2 : Analysis of raw GEDI waveform



- Canopy fuel load
- LAI/ SLA (specific Leaf Area)

## • Canopy bulk density

- Get canopy Fuel Vertical Profile (FVP)
- CBD as maximum of the FVP/SLA



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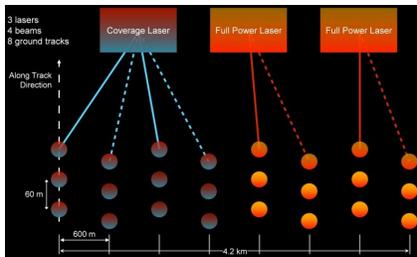
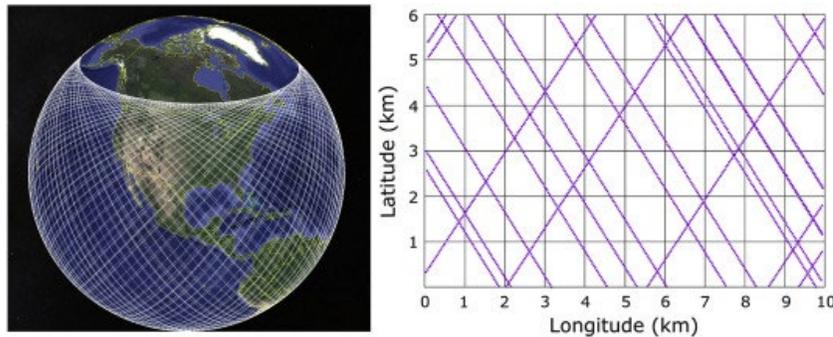
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# Final step: Interpolation model of GEDI data

**Next step:** Use **spaceborne GEDI data** to derive forest canopy height and cover

But we need to **interpolate** because GEDI data is not spatially continuous



## 1 Inverse distance weight (IDW) interpolation of GEDI data

Forest fuel polygons with spaceborne GEDI footprints  
92 % forest under 51.6°N

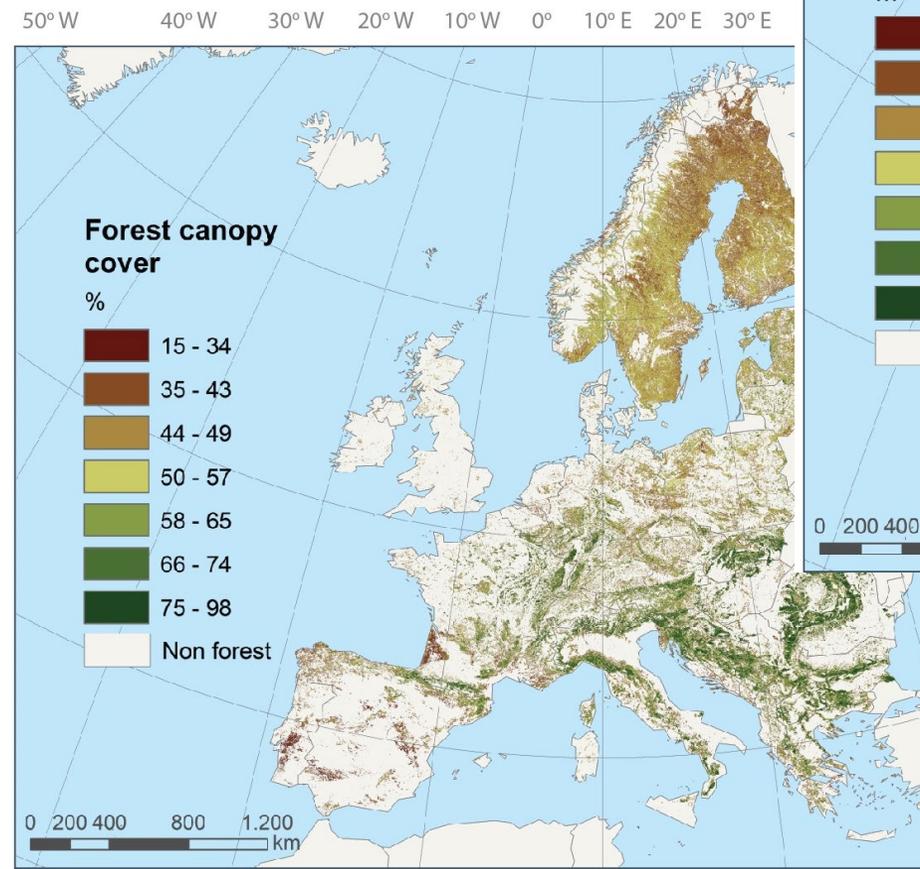
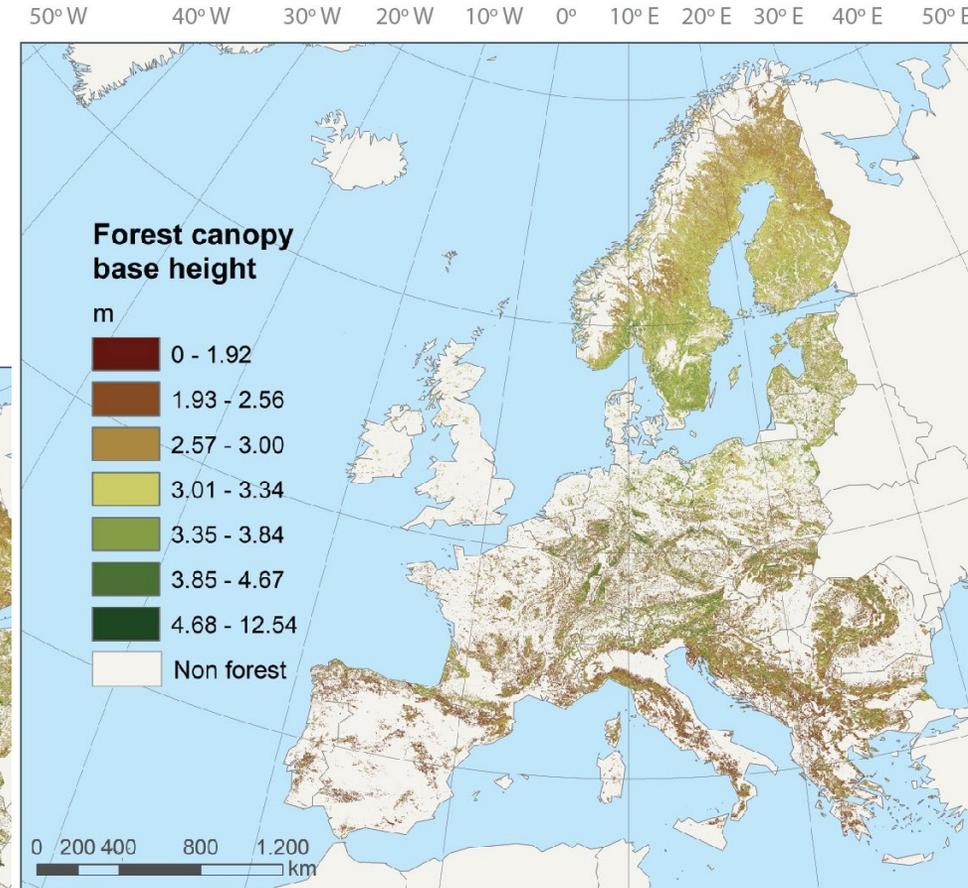
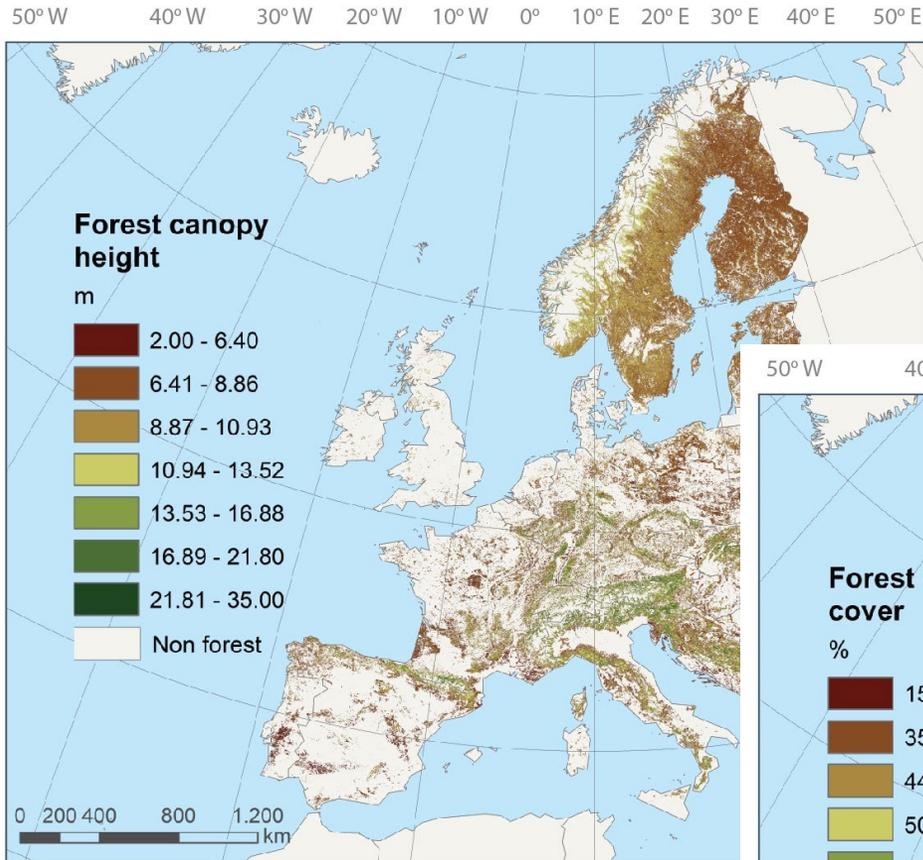
## 2 Random forest & K-nearest neighbour regression models of GEDI + multispectral+ RADAR data + biophysical variables

Forest fuel polygons without spaceborne GEDI footprints

(small polygons under 51.6°N and all polygons above 51.6°N, GEDI coverage)



# Final results: wall-to-wall maps on canopy parameters



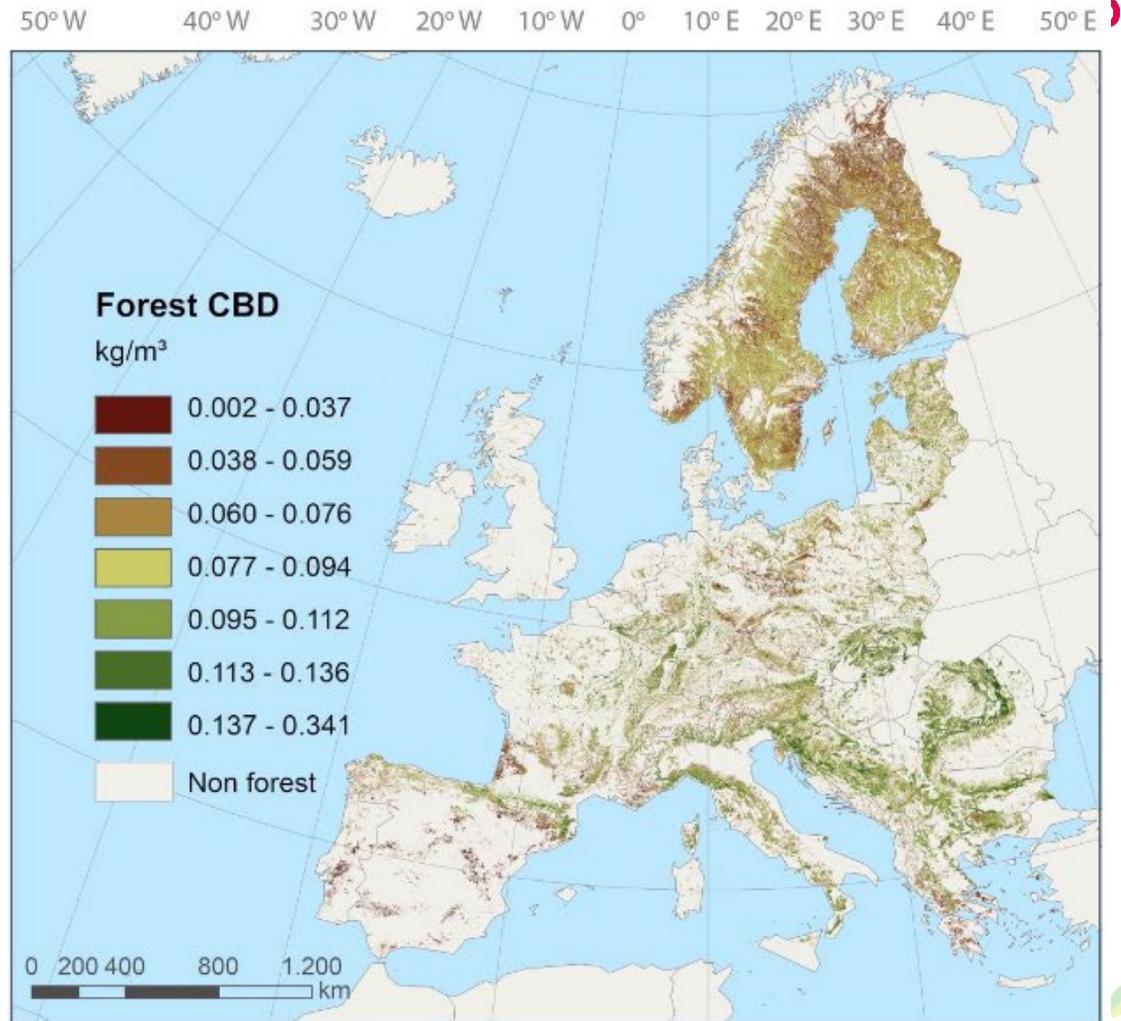
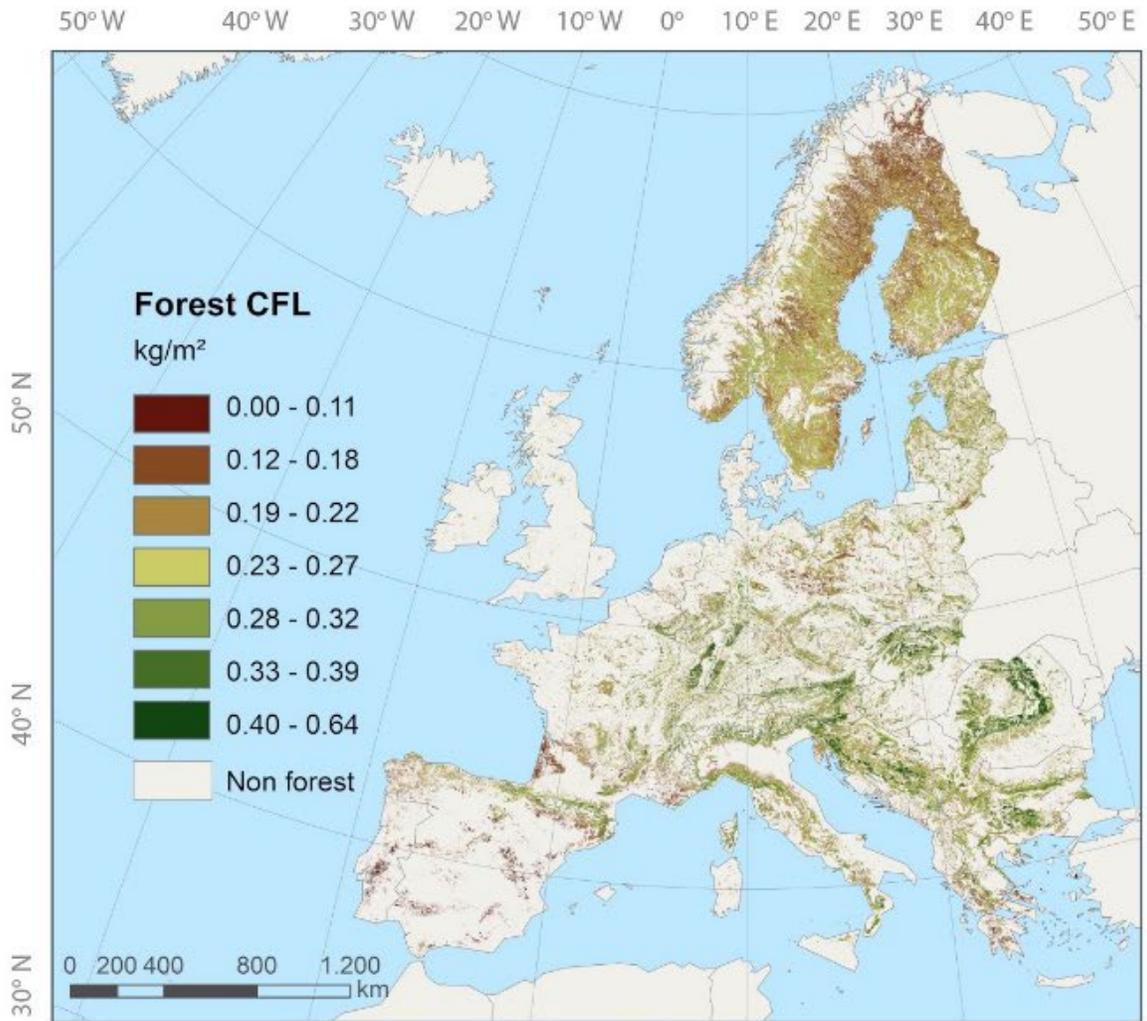
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# Final results: wall-to-wall maps on canopy parameters



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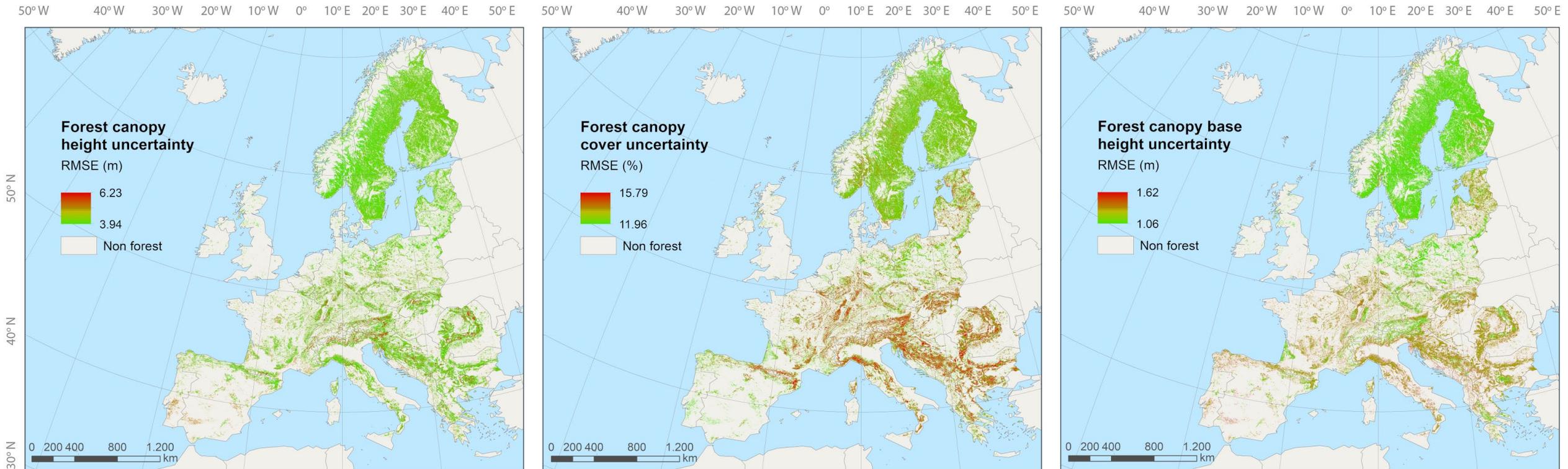
# Generation of uncertainty maps



Steps

- Generation of reference data
- Calibration of GEDI models
- Extrapolation (IDW or RF)

→ Propagation of independent errors



# Final set of available layers



- **Open Access in public repository**

- Aragoneses, Elena; Garcia, Mariano; Chuvieco, Emilio, 2022, "FirEUriSk\_Europe\_fuel\_map: European fuel map at 1 km resolution", <https://doi.org/10.21950/YABYCN>, e-cienciaDatos

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Elena Aragoneses; Mariano García; Emilio Chuvieco, 2024, "FirEUriSk\_canopy\_fuel\_parameters: canopy height, canopy cover and canopy base height", <https://doi.org/10.21950/KTALA8>, e-cienciaDatos, V1

- **Methodology and details in scientific papers**

- Aragoneses, E., García, M., Salis, M., Ribeiro, L. M., & Chuvieco, E. (2023). Classification and mapping of European fuels using a hierarchical, multipurpose fuel classification system. *Earth System Science Data*, 15(3), 1287-1315.
- Aragoneses, E., García, M., Ruiz-Benito, P., & Chuvieco, E. (2024). Mapping forest canopy fuel parameters at European scale using spaceborne LiDAR and satellite data. *Remote Sensing of Environment*, 303, 114005.



# Wildfire Vulnerability

Fátima Arrogante Funes

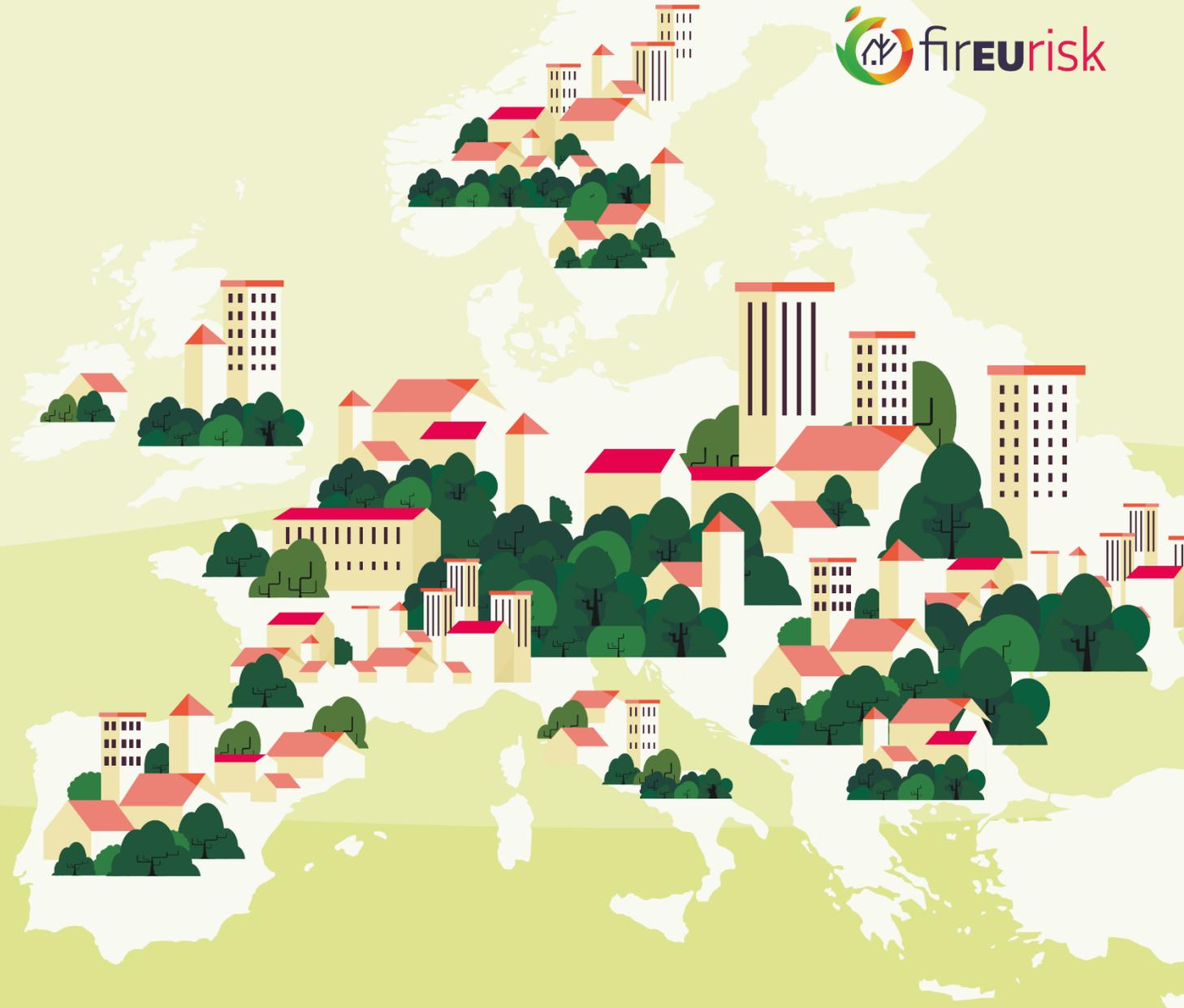
Universidad de Alcalá

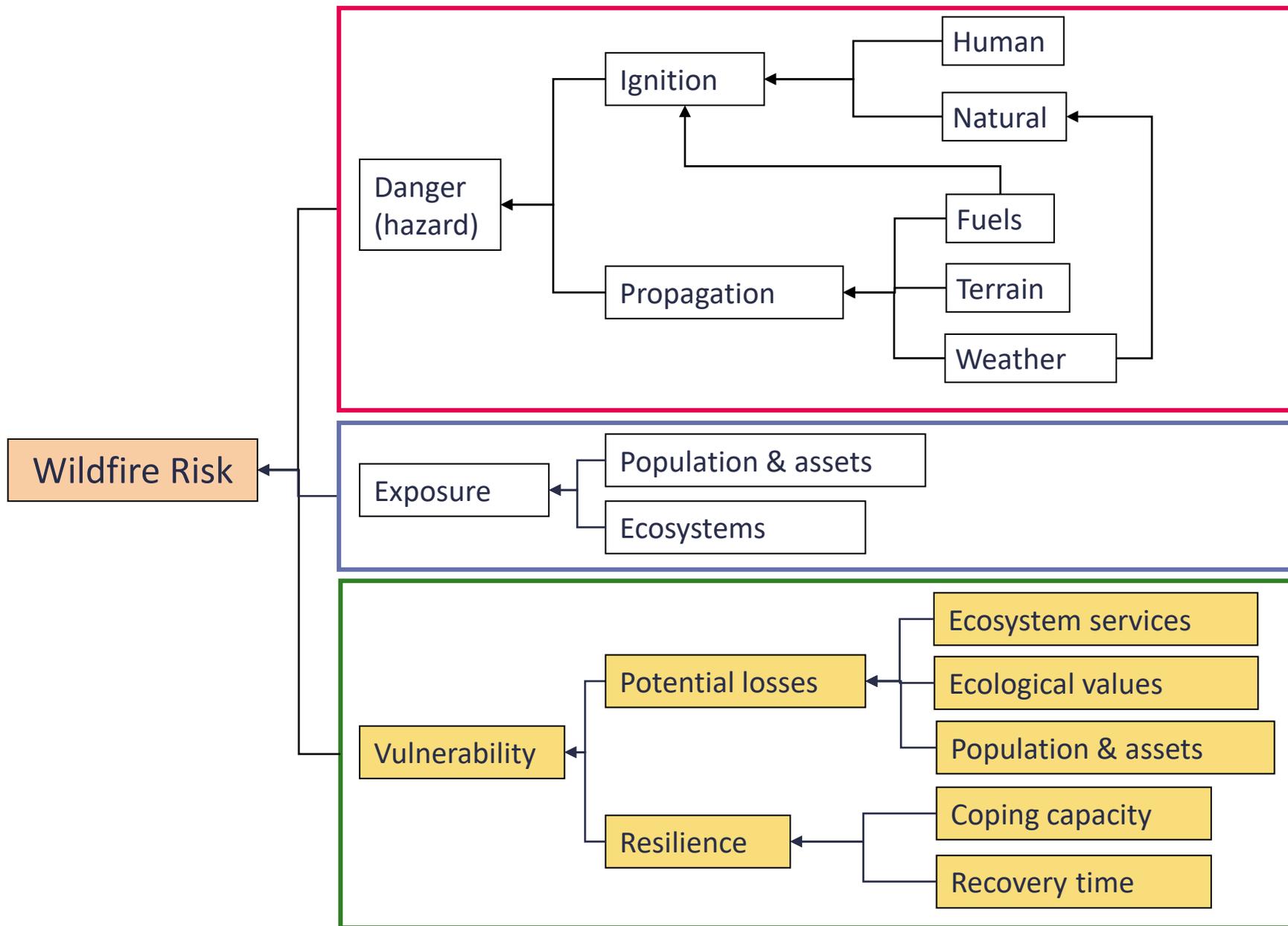
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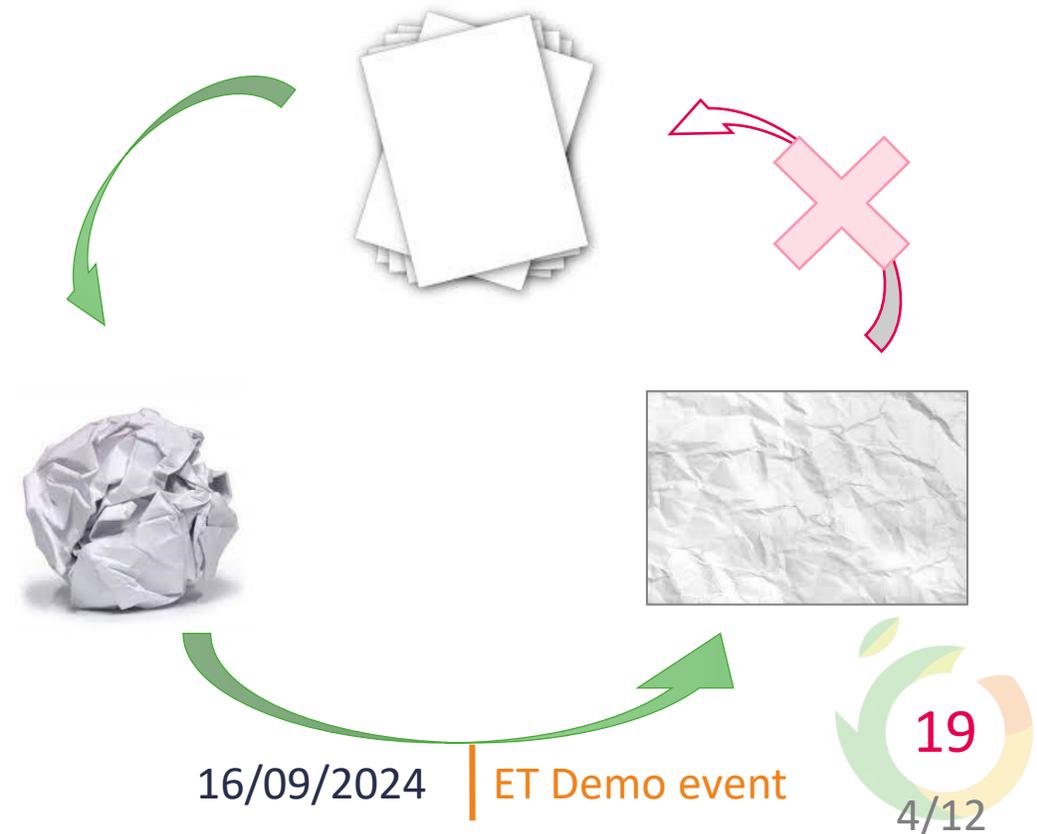
# Objective



Wildfires have a significant impact on **ECOLOGICAL VULNERABILITY**.

... Vulnerability framework refers to a structured approach or model used to **UNDERSTAND AND ASSESS THE POTENTIAL LOSSES AND RESILIENCE OF COMMUNITIES AND SYSTEMS TO THE IMPACT OF NATURAL HAZARDS** such as wildfires

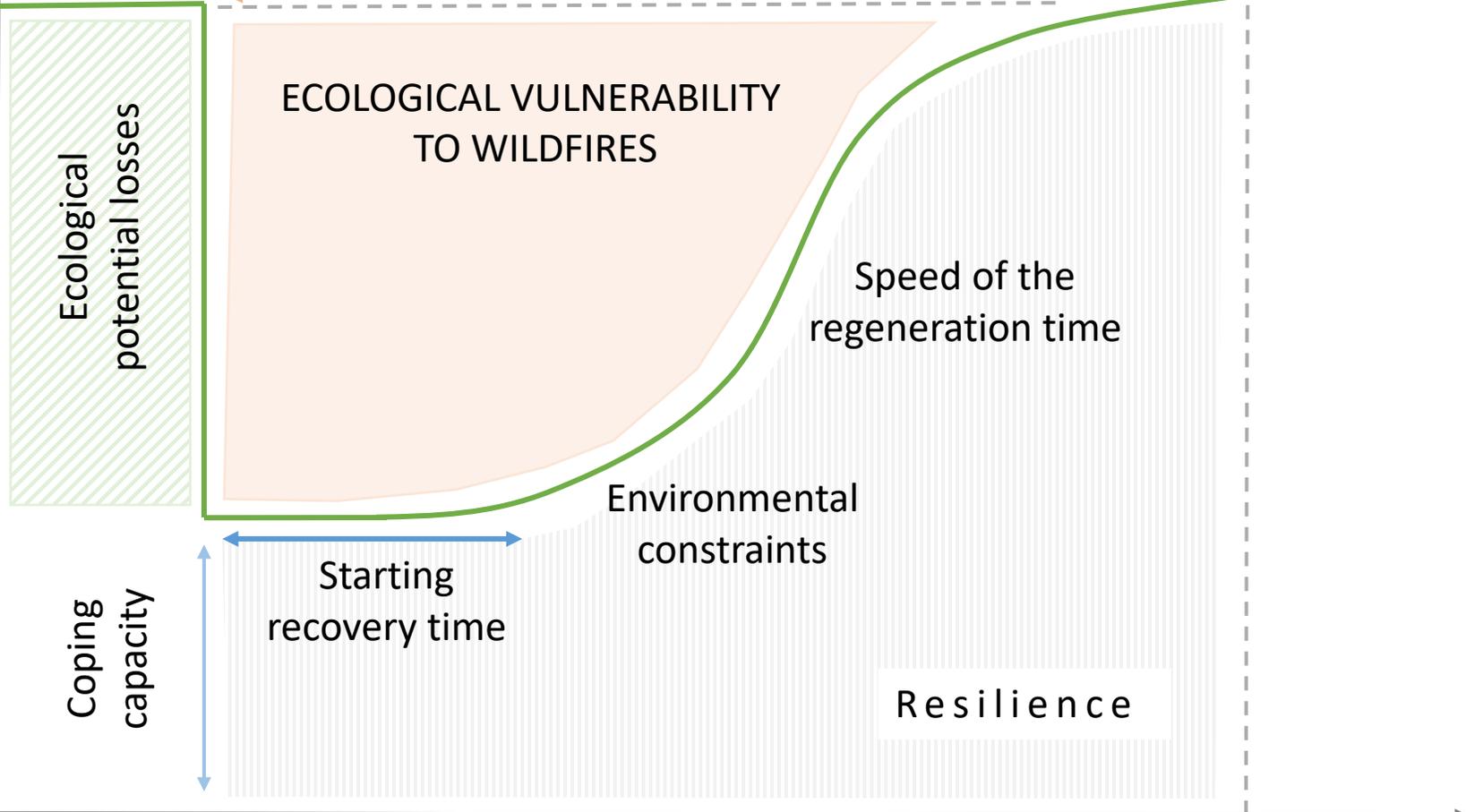
This study focuses on **developing a method to integrate and assess the ecological vulnerability to wildfires** at the European scale by characterising the **ecological values**, the **coping capacity** and **resilience (recovery)** of the ecosystems.



PCA Pre-fire Ecological Values



$$EVW = (EVA * (1 - CC)) \times \frac{1 - (1 + r)^{-\ln RT}}{r}$$

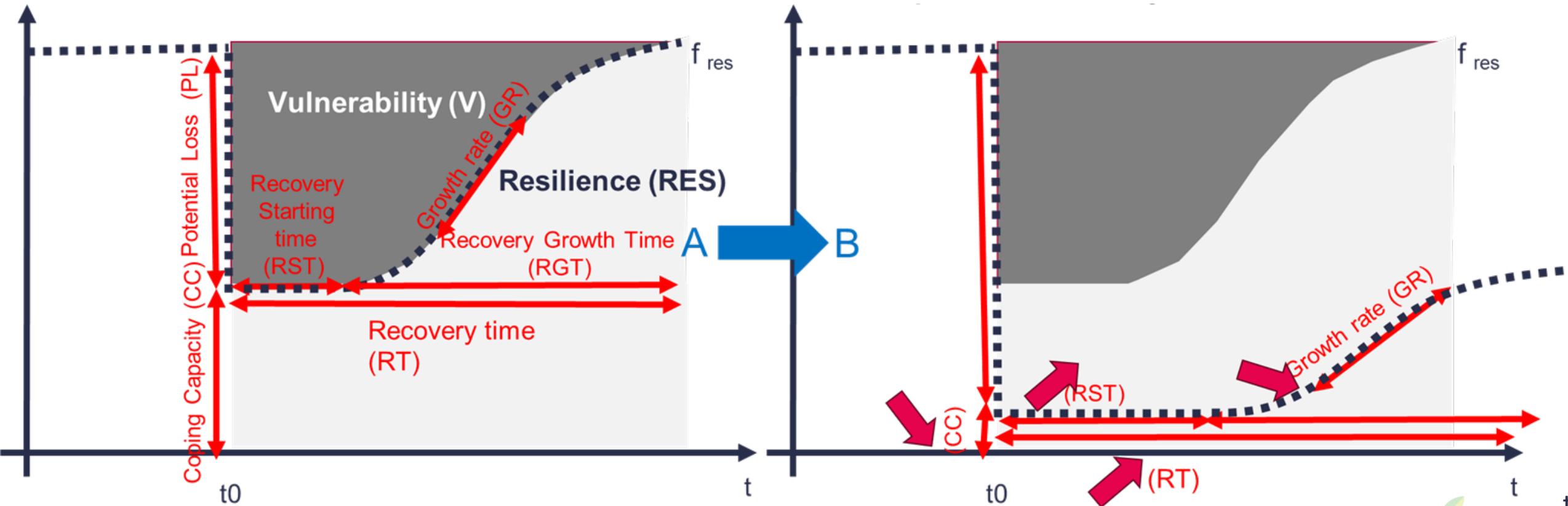


$$CC = RVW . (1 - FLI)$$

$$RT = 1 / \sum_{i=1}^n TPHMi \cdot \sum_{i=1}^n TPHMi \times (RGT . ARTm . ARTt + RT . ARTc)$$

# General concepts & background

Potential variabilities in CC & RT and the subsequent Vulnerability



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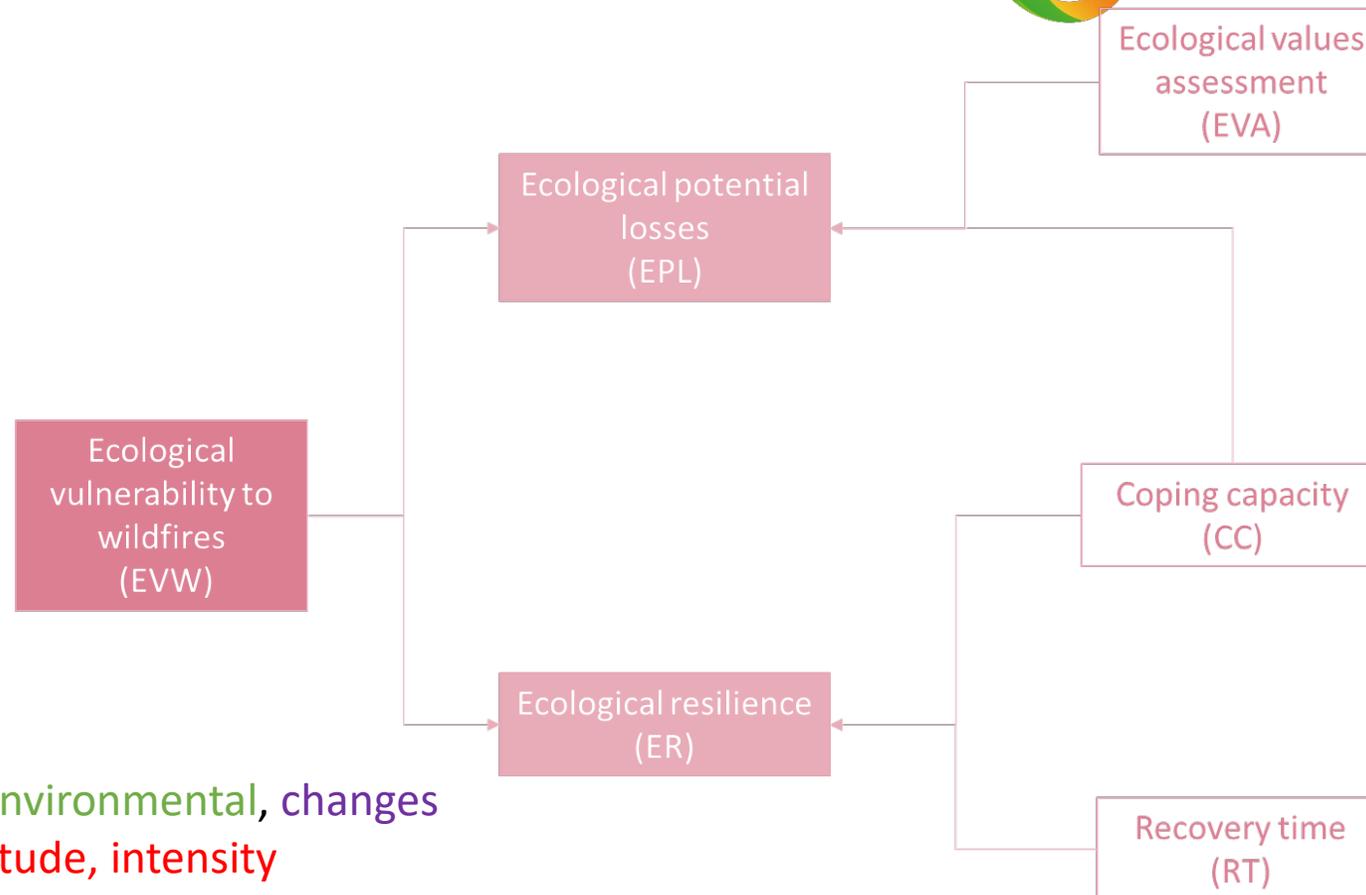
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# Ecological vulnerability to wildfires



Several **data**, several **interactions**...



## Variables brief summary

+ 50.000 sp modelling → Maxent

10 Time series databases → Climatic, Environmental, changes

Fire Regime variables → Severity, magnitude, intensity

Conditions → Forest Management, topography, geological, conservation management



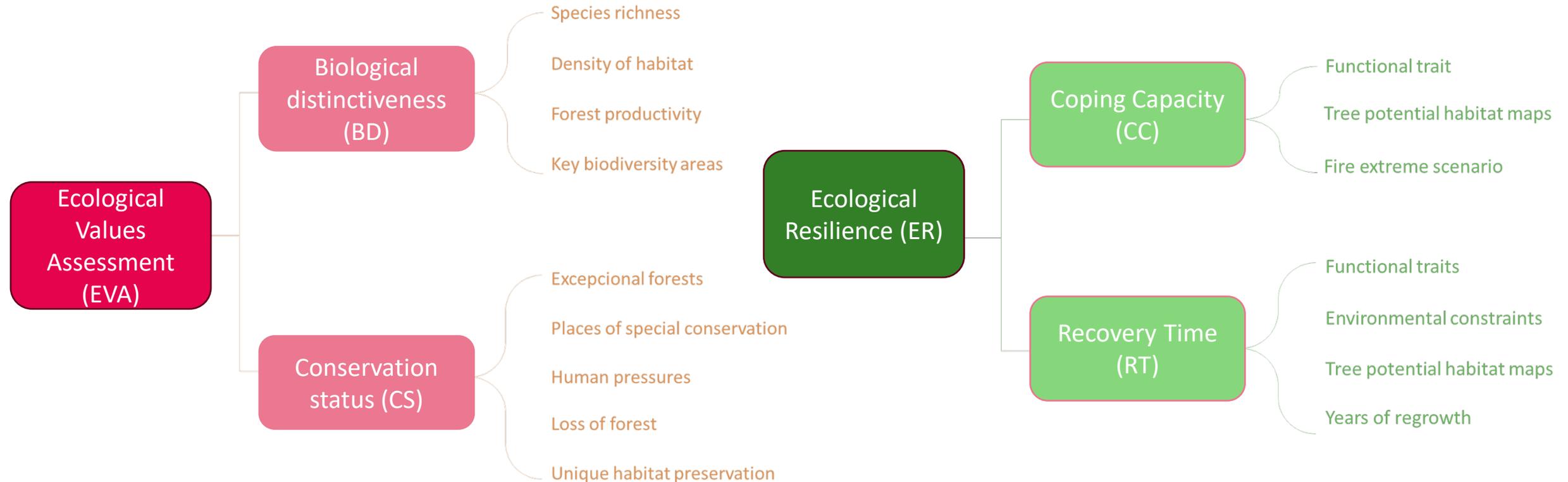
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# Ecological vulnerability to wildfires: components



# Ecological value map



The **Ecological Value Map (EVA)** is a tool designed to assess and represent the ecological importance of different areas by evaluating two primary factors: **Biological Distinctiveness (BD)** and **Conservation Status (CS)**.

## BIOLOGICAL DISTINCTIVENESS (BD)

Refers to the **uniqueness** and **structural biodiversity** of an ecosystem.

Ecosystems with **high BD** provide **critical ecosystem functions** and **contribute to biodiversity conservation**.

Species Richness

**Input:** vectorial layers of mammals (278 sp.), birds (711sp.) reptiles (156 sp.), amphibians (96 sp.), vascular plants (20.000 sp.)

**Source:** UICN

**Method:** Raster Spatialization of occurrence based on IUCN data. Sum.

**Up-to-date:** Last updated 27<sup>th</sup> June 2024

Density of Habitat

**Input:** raster layer of biomass (g/Cm<sup>2</sup>)

**Source:** JRC-Forest

**Method:** Transformed to 1 kilometre using a weighted average approach

**Up-to-date:** 2010

Forest Productivity

**Input:** raster layer of specific leaf area (mm<sup>2</sup> mg<sup>-1</sup>), leaf dry matter content (g g<sup>-1</sup>), leaf nitrogen content and leaf phosphorus content (mg g<sup>-1</sup>)

**Source:** Moreno-Martínez et al., (2018)

**Method:** normalising each variable, sum at the pixel level of the carbon, nitrogen and phosphorus cycle production values

**Up-to-date:** 2018

Key Biodiversity Areas

**Input:** vectorial layer

**Source:** Potapov et al., (2008)

**Method:** to 1km resolution

**Up-to-date:** 2020



# Ecological value map



## KEY COMPONENT: CONSERVATION STATUS (CS)

Assesses the current condition of ecosystems and the threats they face.



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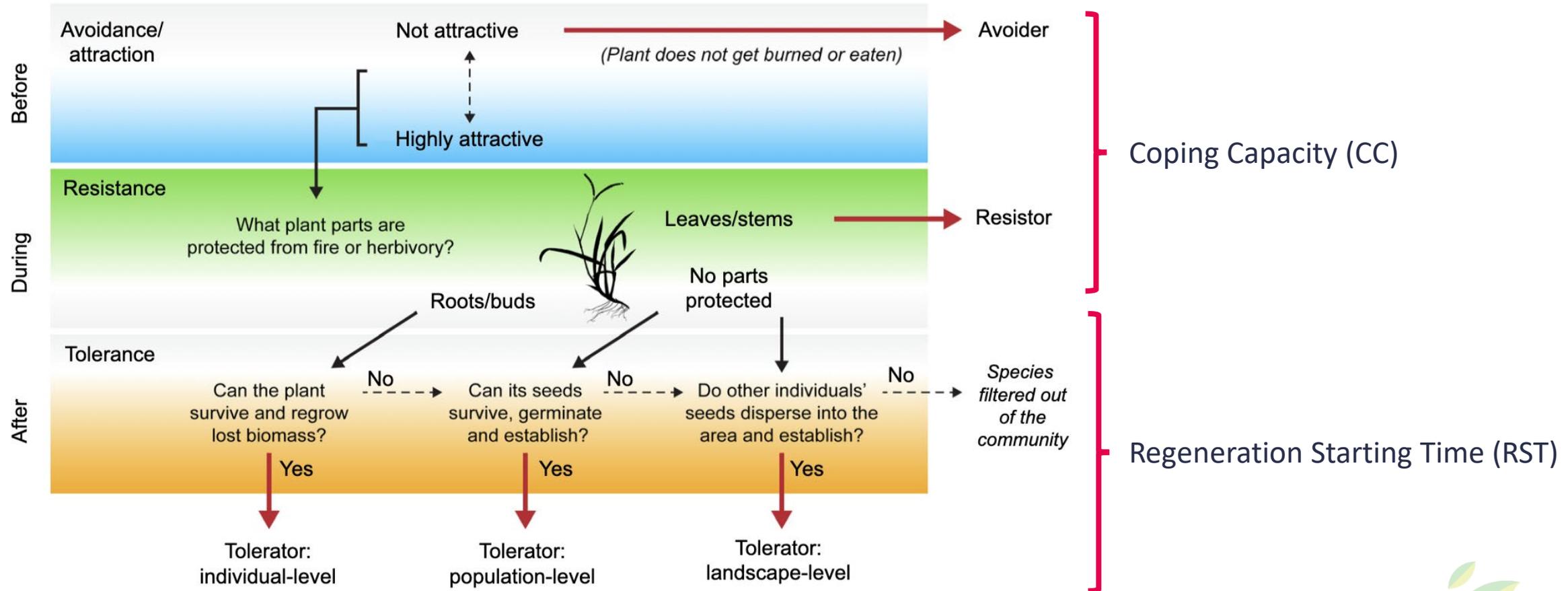
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# Ecological basis for post-fire resilience



Plant life-history strategies (Archibald et al. 2019, New Phytol.)



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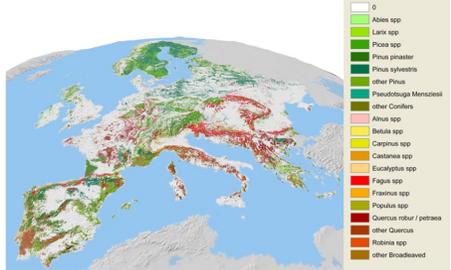


# Processing Chain for EU Resilience Map



COPING CAPACITY

Tree species distribution



Corine Land Cover



Resistance traits:

69 tree species

*Bark Thickness*

*Tree height*

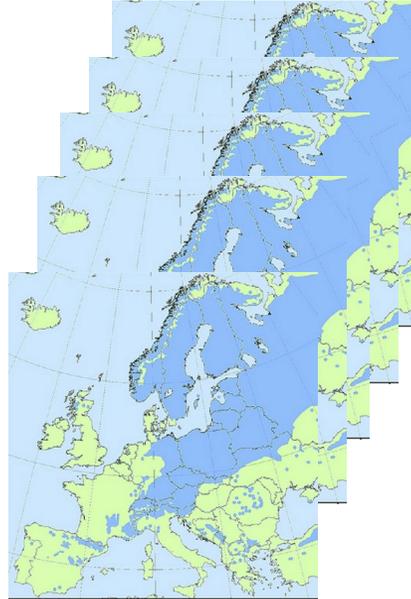
*Basal Crown Height*

Fire Line Intensity Map

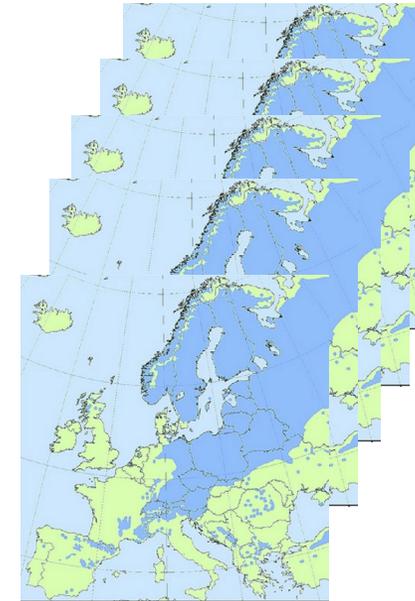


Rescaling 0-1 → Mean (species)

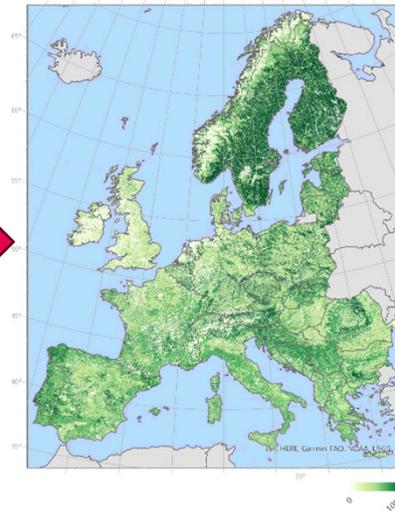
Species presence maps



Species Resistance (0-1)



Coping Capacity Map (0-1000)



Mean



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# Processing Chain for EU Resilience Map



RECOVERY TIME

Tree species distribution



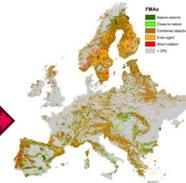
Corine Land Cover



Species presence maps



Environmental constraints

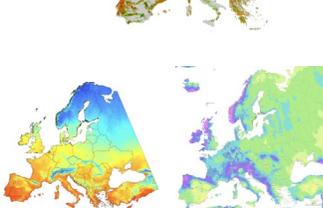


Regeneration traits (69 trees):

*Sprouting*  
*Dispersal*

Recovery Starting Time (RST)

+



Growth traits (69 trees):

Tree height  
*Growth rate*

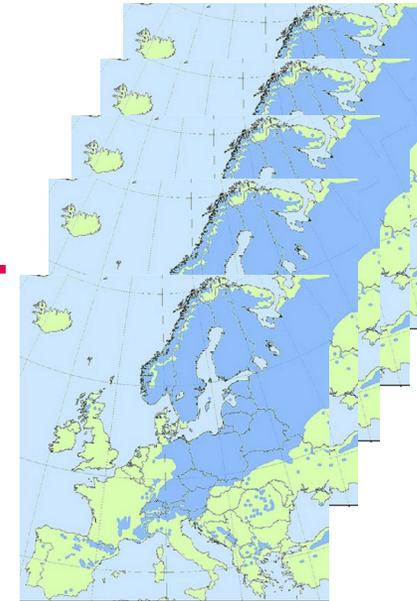
Recovery Growth time (RGT)



+

Recovery Time (RT)

Species RT (years)



Recovery Time (years)

Mean



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# Product

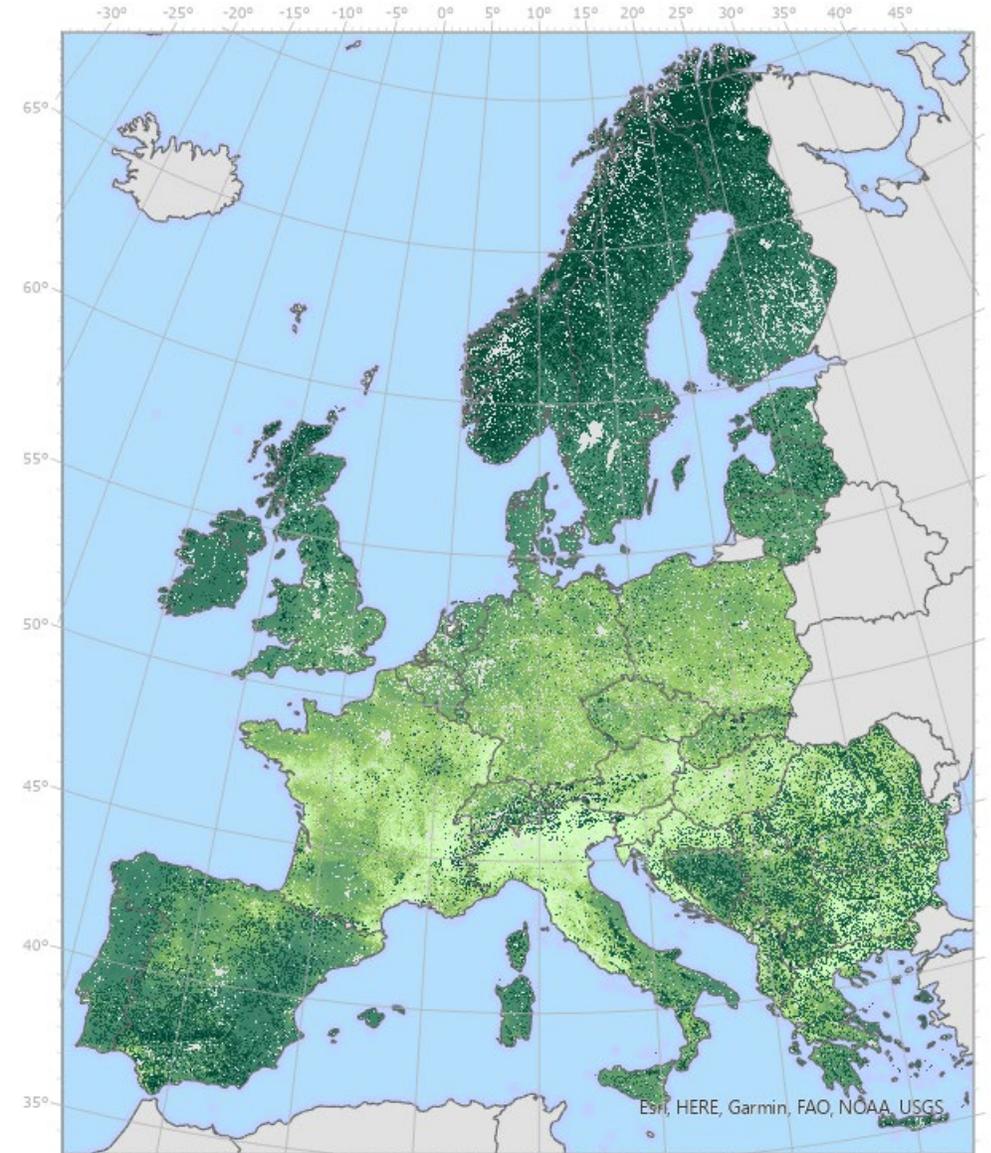
## INTEGRATION METHOD

- **First principal component** → **Synergies** between **BD** and **CS**
- **Ecological Value Map**: Continental level: 1km
- **Adapted to different spatial levels**: Versatile tool for various stakeholders
- **Up-to-date**: Depends of the refresh sources

## AVAILABILITY

Arrogante-Funes, F., Mouillot, F., Moreira, B., Aguado, I. & Chuvieco, E. (2024). Mapping and assessment of ecological vulnerability to wildfires in Europe. *Fire Ecology*. In press.

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**Thank you!**

