

## Inputs for fire Risk Management

Emilio Chuvieco





#### Risk concepts in Natural Hazards

- Danger: probability of occurrence in a determined space and time.
- Vulnerability: "The characteristics and circumstances of a community, system or asset that make it susceptible to the damaging effects of a hazard" (UNISDR, 2009).
  - Value of affected resource: Value \* Recovery time.

#### Risk = Danger \* Vulnerability



### **Fire Risk**

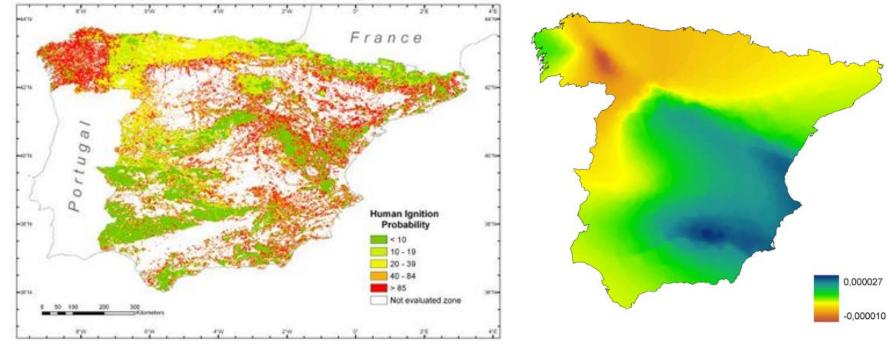
Risk = Danger \* Vulnerability

- Danger:
  - Fuel load (partial)
  - FMC (dead only)
  - Wind speed-direction.
  - Slope.
  - Lightning prob.
  - FMC of live fuels.
  - Human factors.

- Vulnerability:
  - Human values
  - Ecosystem values.



#### Human factors of fire ignition



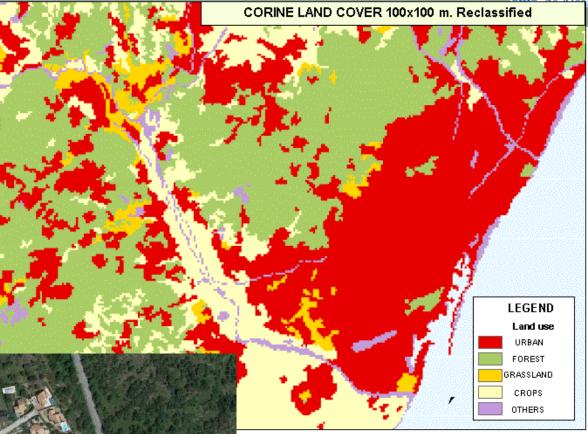
Chuvieco et al., 2014, IJWF

Urban-forest interface Geographical regression

Rodrigues et al., 2014



WUI



**CORINE 100X100 m** 

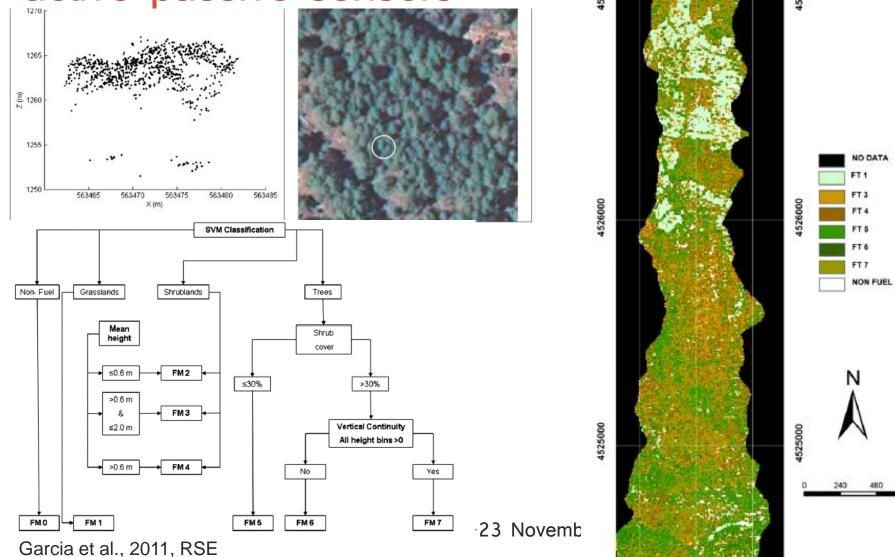
Mercado de Segunda Mano SL os Camochos

vember 2017



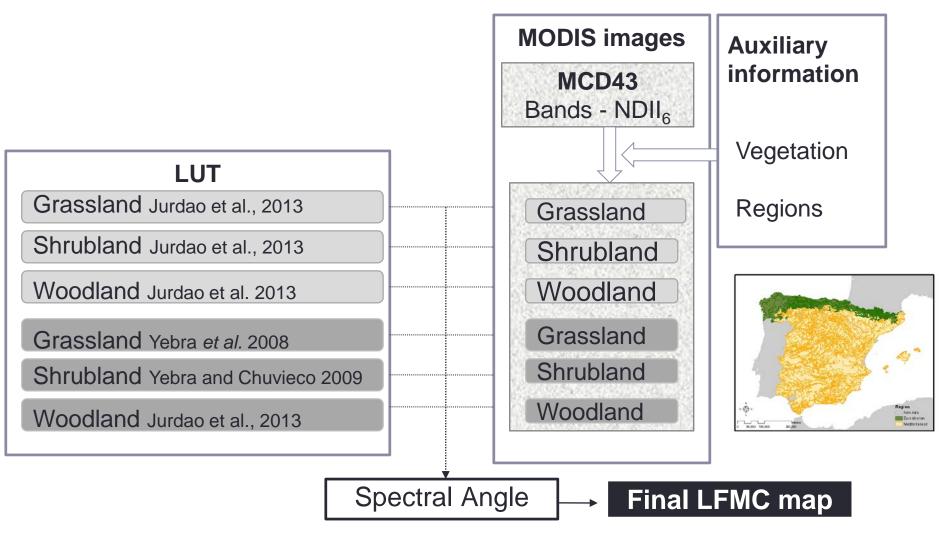
720

# Fuel type mapping from the integration of active-passive sensors



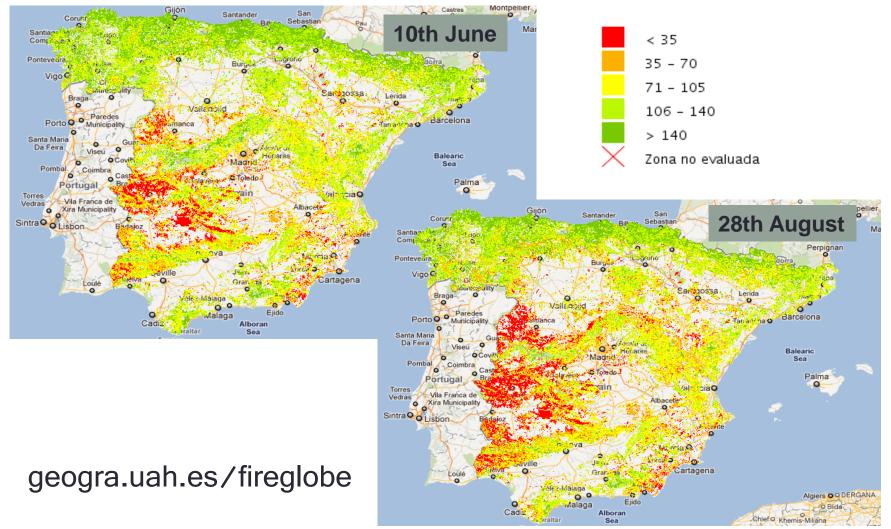


## Semi-operational LFMC estimation





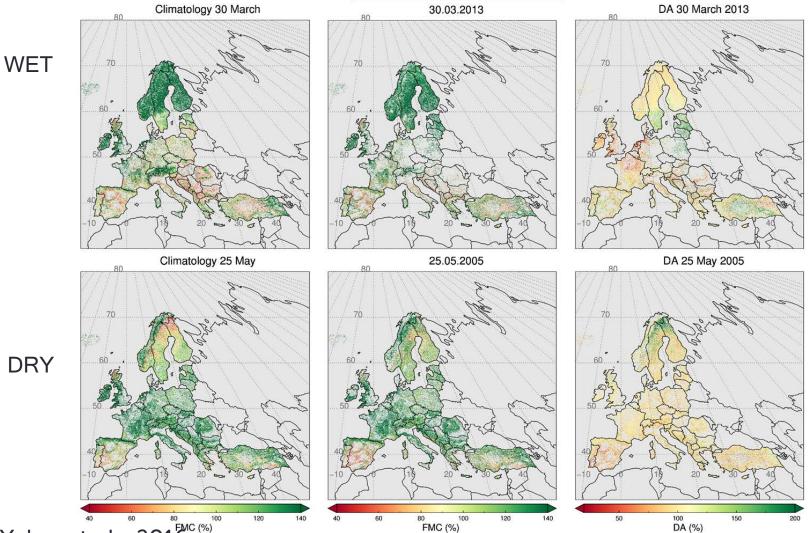
#### Examples of FMC maps





#### LFMC estimation from MODIS RTM

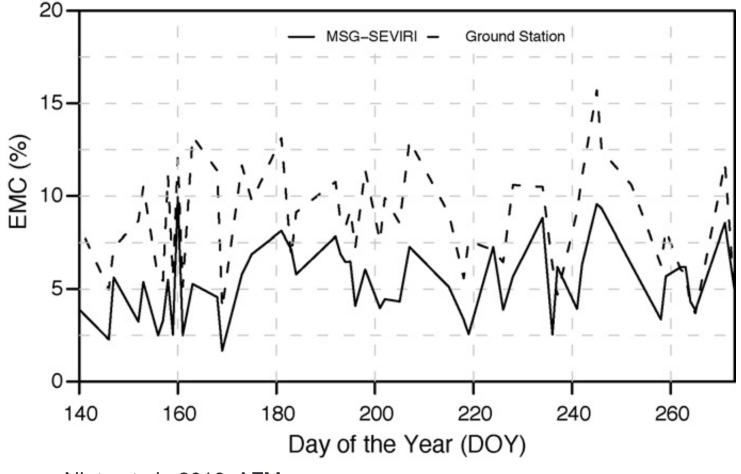
#### modelling



Yebra et al., 2016



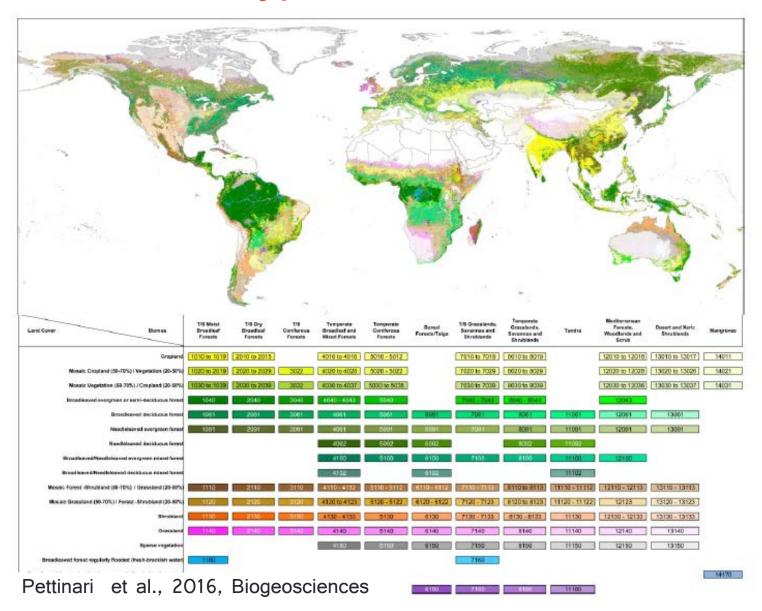
## Estimation of EMC from MSG Seviri



Nieto et al., 2010, AFM

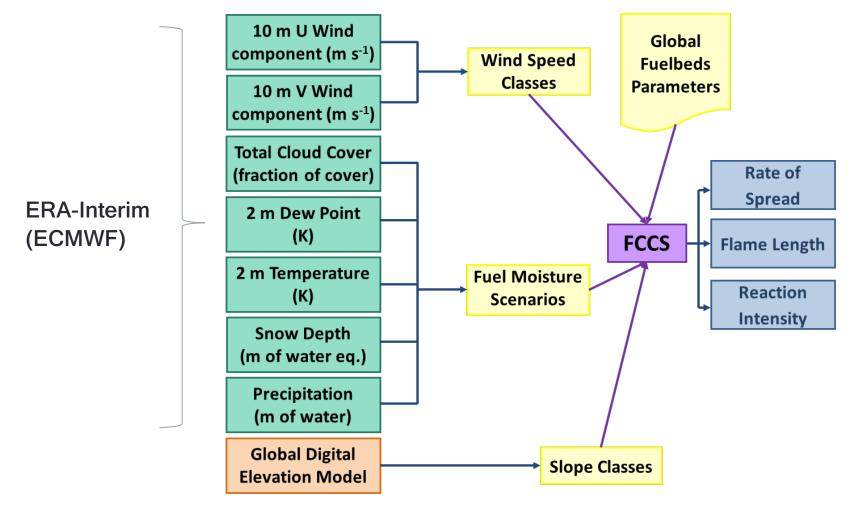


#### Global Fuel type characterizaton





### Estimation of fire danger parameters



Pettinari & Chuvieco, 2017, Forests



Rate of

Spread

Flame Leng

Reactio

Fuelbeds

Parameters

FCCS

#### Fuel moisture scenario maps

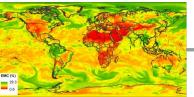
#### Precipitation

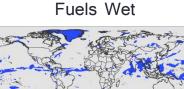


Snow Depth

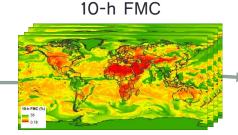


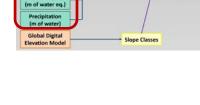






Wet: 10-h FMC = 35 % Dry: 10-h FMC = 1.28 \* EMC





Wind Spee

Classe

Fuel Moistur Scenarios

10 m U Wind mponent (m s<sup>-1</sup>)

10 m V Wind component (m s

fraction of cove

2 m Dew Point (K)

2 m Temperatu

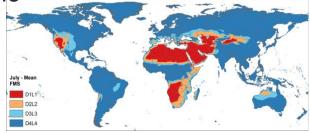
Snow Dept

Man July FMC (%)

Mean July 10-h

July Mean Fuel Moisture Scenario

- <u>D1L1</u>: 10-h FMC < 5.5%
- <u>D2L2</u>: 5.5% <= 10-h FMC < 8.5%
- <u>D3L3</u>: 8.5% <= 10-h FMC < 10.5%
- <u>D4L4</u>: 10-h FMC >= 10.5%

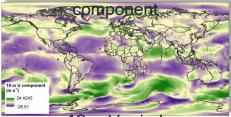


Pettinari & Chuvieco, 2017, Forests

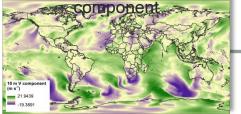


#### Wind speed

10m U wind



10m V wind



10 m Wind Speed = 
$$\sqrt{W_U^2 + W_V^2}$$

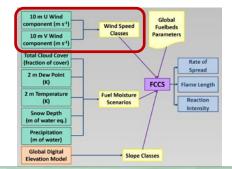


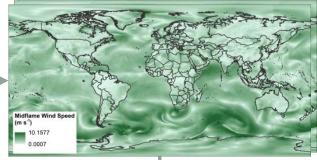
Wind adjustment factor: 0.348 (intermediate between sheltered and unsheltered fuels) *Midflame Wind Speed* = *Wind Speed* \* 0.348

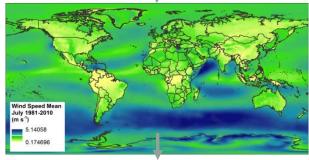
#### July Mean Wind Class

- <u>Class 1</u>: 0 1.0 m s<sup>-1</sup> (0 2.24 mph)
- <u>Class 2</u>: 1.0 − 2.5 m s<sup>-1</sup> (2.24 − 5.59 mph)
- <u>Class 3</u>: 2.5 5 m s<sup>-1</sup> (5.59 11.18 mph)

Pettinari & Chuvieco, 2017, Forests



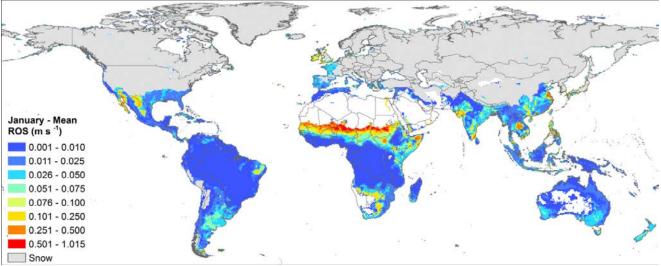


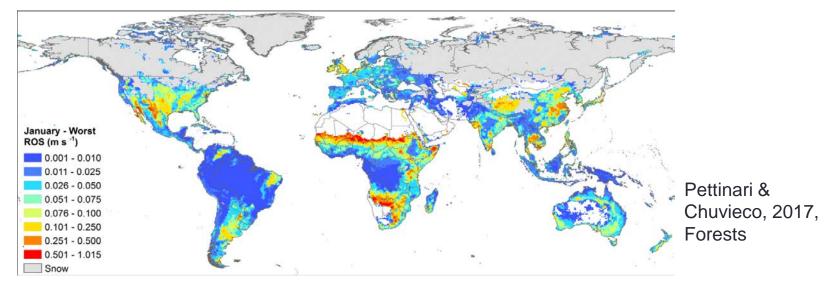






# Estimation of fire danger parameters from forecasted data (RoS)







#### Global fire vulnerability

Global Ecology and Biogeography, (Global Ecol. Biogeogr.) (2014) 23, 245-258



#### Integration of ecological and socio-economic factors to assess global vulnerability to wildfire

Emilio Chuvieco<sup>1\*</sup>, Susana Martínez<sup>2</sup>, María Victoria Román<sup>3</sup>, Stijn Hantson<sup>1</sup> and M. Lucrecia Pettinari<sup>1</sup>

<sup>1</sup>Environmental Remote Sensing Research Group, Department of Geography and Geology, Universidad de Alcalá, Colegios 2, 28801 Alcalá de Henares, Spain, <sup>2</sup>GI-TB Lab. Botánica y Biogeografía, IBADER, Campus de Lugo, Universidad de Santiago de Compostela, 27002 Lugo, Spain, <sup>3</sup>Department of Economícs, Universidad de Alcalá. Colegios 2, 28801 Alcalá de Henares, Spain

#### ABSTRACT

**Aim** This paper presents a map of global fire vulnerability, estimating the potential damage of wildland fires to global ecosystems.

Location Global scale at 0.5° grid resolution.

**Methods** Three vulnerability factors were considered: ecological richness and fragility, provision of ecosystem services and value of houses in the wildland–urban interface. Each of these factors was estimated from existing global databases. Ecological values were estimated from biodiversity relevance, conservation status and fragmentation based on Olson's ecoregions. The ecological regeneration delay was estimated from adaptation to fires and soil erosion potential. The former was assessed by comparing actual land cover with fire-off simulations based on a dynamic global vegetation model (ORCHIDEE). The annual loss of ecosystem



#### Present Marginal Loss (PML)

$$PML = ML * \frac{1 - (1+r)^{-\log n}}{r}$$

• ML = 
$$V_1 - V_0$$

- V<sub>0</sub> = Initial value of resource, related to economic assessment
- $V_1$  = Value after fire, related to fire behavior.
- n: number of years to regenerate, related to ecological conditions and fire behavior.
- r: Discount rate (2%).

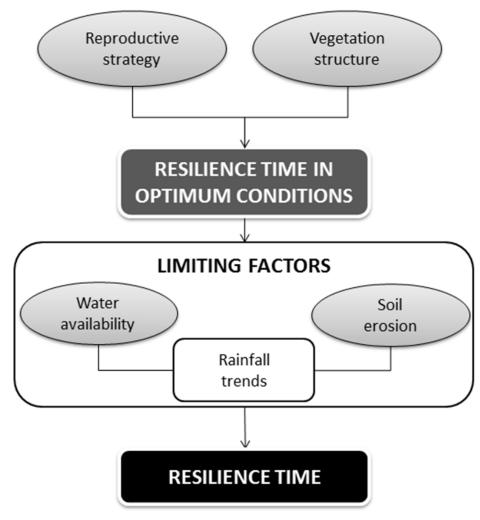
Chuvieco et al., 2014, IJWF

GOFC-GOLD Fire IT - 20-23 November 2017

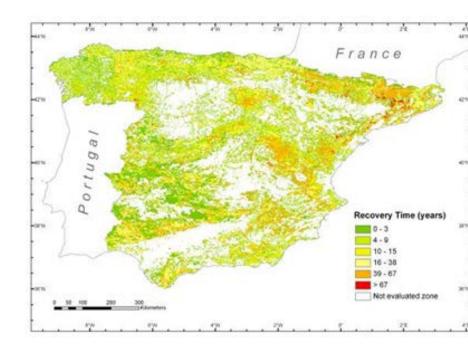
Assumed Estimated



## Recovery time (n)



Rodrigues et al., 2014, PPG



Chuvieco et al., 2014, IJWF



#### RS in Fire risk assessment

- Improve methods to generate input datasets:
  - Fuel types.
  - Fuel moisture content.
  - Human factors.
- Better link RS data to ignition and propagation models.
- Consider both danger and vulnerability:
  - Mitigating risks.