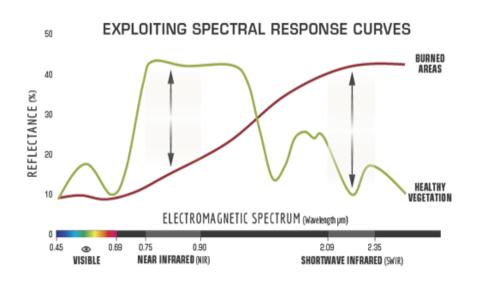
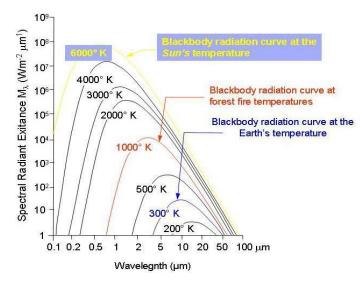


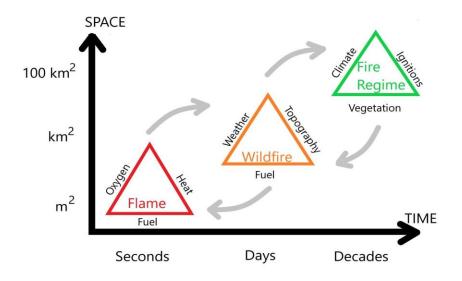
Introduction on the radiometry of forest fires





Healthy plant species reflect more energy in NIR but weakly in SWIR. This spectral characteristic is useful for detecting burned areas such as dead soil/plant material on forest floor. Source: US Forest Service.

Remote sensing observations and forest fires

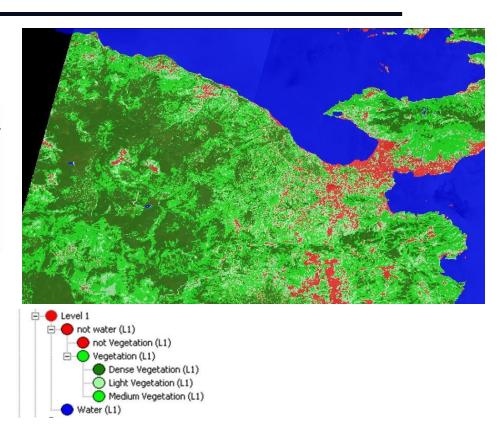


Pre-Fire Risk Assessment	Vegetation density and extent	
	Soil moisture/drought severity	
	Topography	
	Fire risk mapping	
Active Fire Detection	Hot Spot Detection	
	Total area burning	
	Fire Radiative Power and	
	Thermal Infrared	
	Pyro cloud formation	
Post Fire Assessment	Total area burned	
	Burn severity	
	Post fire vegetation regrowth	
	Landscape regeneration	

Remote sensing observations and forest fires stages

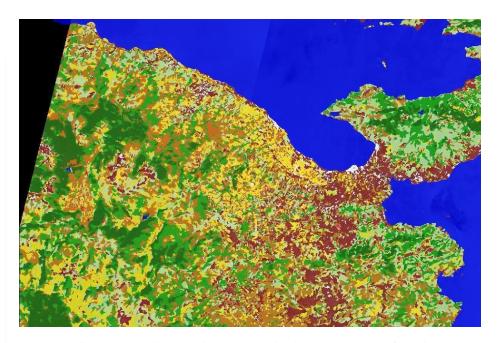
Vegetation density

- Fire risk as varies with vegetation density VD.
 VD also influences fire dispersion.
- Classification of a forest in terms of VD depends strongly on spatial resolution



Vegetation type

- Urban areas slow down forest fires, a fact which is important for fire modelling.
- Fuel behavior (ignition and dispession)
 varies with vegetation type (VT).
- For instance, areas with olive trees slow down the fire. On the contrary, areas with pine trees (typical species for the Mediterranean biodiversity), ignite and disperse easier.



Brown: olive trees; Yellow: agricultural cultivations; Green: conifers; light green: Shrubs

Vegetation stage - Land Surface Phenology

Unhealthy vegetation has a higher percentage of dead leaves, providing easier to burn fuel for fires

Satellites can be used to track seasonal patterns of variation

in vegetated land surfaces through indices:

- NDVI Normalized Difference Vegetation Index
- EVI Enhanced Vegetation Index
- SAVI Soil-Adjusted Vegetation Index
- Vegetation index anomalies



1/ NDVI is widely used as a metric for vegetation health

- Values range from -1.0 to 1.0
- Negative values to 0 mean no green leaves.
- Values close to 1 indicate the highest possible density of green leaves.
- NDVI Formula: (NIR Red)/(NIR + Red)

$$EVI = G * \left(\frac{(NIR-R)}{(NIR+C1*R-C2*B+L)}\right)$$

$$Constants$$

$$G = 2.5$$

$$C1 = 6$$

$$C2 = 7.5$$

$$L = 1$$

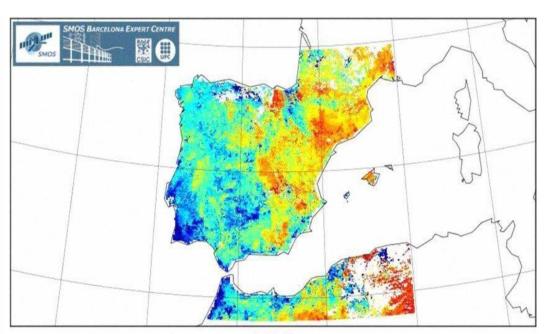
2/ Enhanced Vegetation Index (EVI)

More sensitive in areas with dense vegetation, making it better for fuels assessment in dense forests

Soil Moisture

Vegetation-Based Fire Applications:

- Vegetation Moisture: Soil moisture acts as a proxy for vegetation moisture and evaporative stress.
- Drought information can also identify areas with dry fuel.
- Soil moisture is measured by active microwave scatterometers such as ERS1&2/AMI and MetOp/ASCAT as well as by passive microwave radiometers such as Sentinel 1, Aqua/AMSR-E, Coriolis/WindSat...

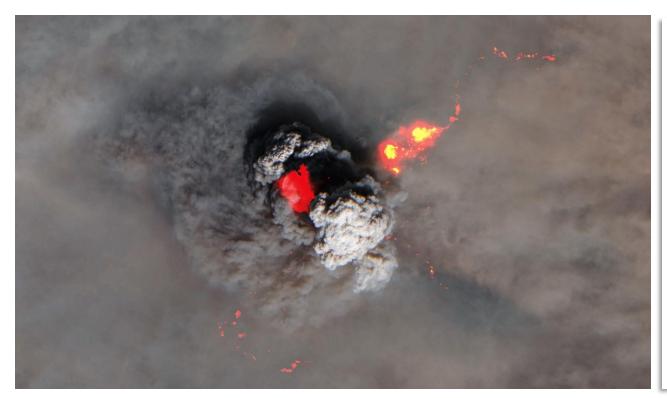


Example of high resolution (1 km) soil moisture maps of the Iberian peninsula generated from SMOS data (10 days average)

https://directory.eoportal.org/web/eoportal/satellite-missions/s/smos

• Measures the moisture in the top 5 cm of the soil globally every 3 day

Local meteorology (pyrocumulus)



Copernicus Sentinel-2 catches impressive smoke cloud

This image acquired on 9 September 2020, by Copernicus Sentinel-2 features the impressive pyrocumulus cloud forming over the complex wildfire in California. This true-colour image is combined with short-infrared bands to highlight the location of the fire hot spots.

Contains modified Copernicus Sentinel data (2020)/processed by P. Markuse

SAR in support of optical remote sensing for forest fires

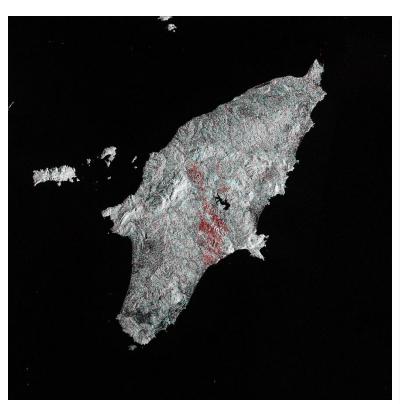
Sentinel-1 SAR (C-band SAR data, 12-day revisit, Resolution: 5 x 20 meters)

- Vegetation-Based Fire Applications:
- Vegetation Type and Extent: Land classification, fuels mapping
- Vegetation Structure: Density and height
- Vegetation Moisture: Fuel moisture content and dryness

Sentinel-2 Vegetation-Based Fire Applications (-day revisit, Resolution: 10 meters)

- Vegetation Extent and Type: Land cover classification
- Vegetation Stage and Health: Variety of vegetation indices, including NDVI, EVI, SAVI
- Vegetation Moisture: NDWI

SAR in support of optical remote sensing for forest fires



Between 18 and 28 July 2023, wildfires broke out on Rhodes. Fierce blazes ravaged almost 18,000 hectares of land, destroyed buildings, trapped animals and led to a mass evacuation of thousands of tourists.

This Copernicus Sentinel-1 image shows the burn scars left by fires on the Greek island of Rhodes.

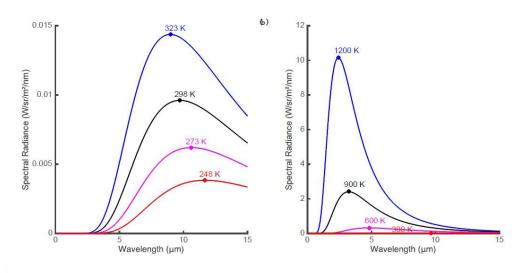
Fire Risk Mapping	Ignition	NO	-
	Land cover	YES	VIS and
			SAR
	Soil moisture and	YES	Microwaves
	drought severity		
	Vegetation type and	YES	VIS
	stage		
	Burning fuel	YES	VIS
	Topography	YES	VIS and
			SAR
	Meteorological	LIMITED	VIS and TIR
	parameters		
	Land surface	YES	TIR
	temperature		

To provide reliable Fire Risk Mapping, spatial resolution needs to be high

As fires burn much hotter than the typical temperature of surfaces on the Earth, heat provides a strong signal for the detection of fire.

The total energy radiated from a surface increases rapidly with its temperature (proportional to the fourth power of temperature as described by the Stefan-Boltzmann law).

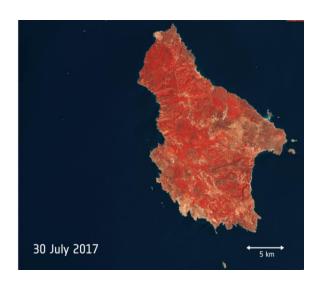
However, the radiance is not uniform across wavelength and the distribution peaks at a wavelength that varies inversely with the temperature.

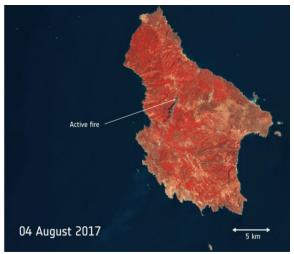


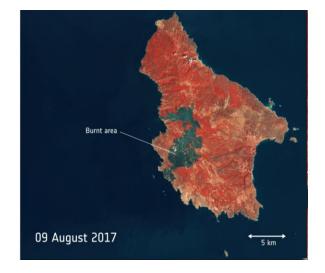
At normal ambient temperature, the peak is in the range 8–12 μ m and most of the radiant energy lies at wavelengths greater than 5 μ m (left image).

At higher temperatures typical of forest fires, the peak of the response shifts to mid-wave infrared (MWIR, 3-5 μ m) or shorter wavelengths (right image).

Detecting hotspots and Total area burning







Kythira wildfires

Southern Europe experienced a relentless heatwave this summer, fuelling wildfires in a number of countries. The Copernicus Sentinel-2 satellite pair captured the start of a fire on the Greek island of Kythira on 4 August. Five days later, a huge burn scar is visible across the western part of the island.

Source: modified Copernicus Sentinel data (2017), processed by ESA

Fire Radiative Power and Thermal Infrared

Comparing window channels in near and thermal infrared

Near infrared (1.6 μm)	More adequate for smoke detection than 3.9 μm Small fires not visible No CO2 absorption (higher fire temperature) High sub pixel sensitivity
Middle infrared (3.9 μm)	High temperature sensitivity - major sub pixel effects (hotspots are easily detected) Negligible absorption by atmospheric humidity Close to a CO2 absorption band, 4-7 Kelvin signal reduction Brightness is temperature of the CO2 layer above the fire
Thermal infrared (10.8 μm)	1-2 Kelvin absorption by atmospheric humidity No signal reduction by CO2 Lower temperature sensitivity (small subpixel effects) No risk of sensor blinding by fires Low values compared with 3.9 µm due to semi transparent cloud or smoke

Temporal resolution – a critical parameter

The majority of satellites providing earth imagery are either geostatic or in the near-polar sunsynchronous orbit and include multispectral imaging sensors.

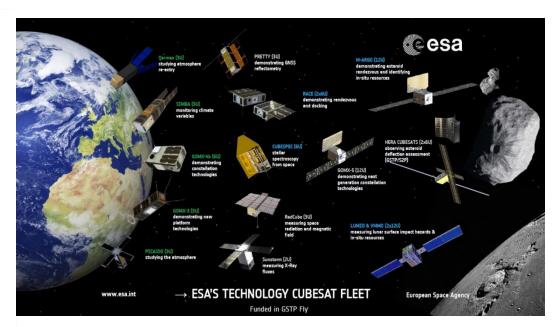
Sun-synchronous satellites provide data with **high spatial resolution but low temporal resolution** while geostationary satellites have **high temporal resolution but low spatial resolution.**

Sensor/Satellite	Channels	Product	Spatial/Temporal
SEVIRI/Meteosat	3.9µm, 10.8µm	FIR (Active Fire Monitoring)	3 km/5 min
MODIS/Aqua and Terra	4μm, 11μm	Active Fire	1km/1-2 days per satellite
SLSTR/SENTINEL 3	3.7μm, 10.8μm	Active Fire	1km/approx. 1 day
SENTINEL 1	Radar	Burned area	5m/2 days at mid-latitudes
SENTINEL 2	Vis	Burned area	10 m/2-3 days at mid-latitudes
AVHRR/NOAA	3.7μm	FIMMA	1km/5-6 times per day
VIIRS/Suomi-NPP	4μm, 11μm	Active fire	375m/3-4 times per day

Temporal resolution – a critical parameter

Recently, advances in nanomaterials and micro-electronics technologies have allowed the use of tiny low-Earth-orbiting satellites, known as CubeSats.

CubeSats by launched in constellations succeed in improving considerably the temporal resolution while at the same time they reflect high spatial resolution (due to their low orbit).



The post-fire stage

Normalized Burn Ratio (NBR)

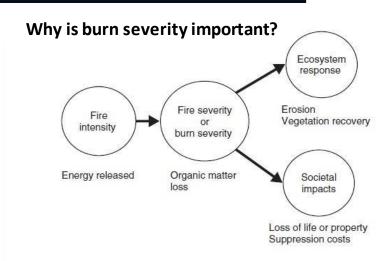
- Used to measure burn severity by distinguishing areas that have been significantly altered in their spectral signature after a wildfire event
- It is calculated using the energy intensity from the NIR and SWIR wavelength bands from the remotely sensed satellite imagery.
- Healthy vegetation has very high NIR reflectance and low reflectance in the SWIR portion of the spectrum
- Burned areas on the other hand have relatively low reflectance in the NIR and high reflectance in the SWIR band

$$NBR = \frac{(NIR - SWIR)}{(NIR + SWIR)} = \frac{(Band\ 8 - Band\ 12)}{(Band\ 8 + Band\ 12)}$$

The post-fire stage

Burn Severity and the delta normalized burn ratio

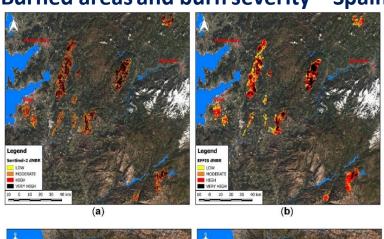
- Burn severity degree to which an ecosystem is impacted by a wildfire event. It is estimated as the difference between pre-fire and post-fire NBR derived from satellite images.
- The difference between pre-fire and post-fire NBR (the delta normalized burn ratio (dNBR) index) - frequently used to identify recently burned areas and differentiate them from other non-vegetated areas.
- Areas with high dNBR value correspond to a higher degree of damage or burn severity. In contrast, low dNBR values represent areas that are unaffected from the fire event or regions that have rebounded via regrowth of plant species following a wildfire incident.



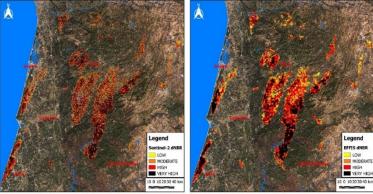
EFFIS thresholds	Severity Level
dNBR < 0.100	Unburned / Very Low
$0.100 \le \text{dNBR} \le 0.255$	Low
$0.256 \le \text{dNBR} \le 0.419$	Moderate
$0.420 \le dNBR \le 0.660$	High
dNBR > 0.660	Very High

The post-fire stage

Burned areas and burn severity - Spain and Portugal 2017



- (a) Sentinel-2 dNBR image using Post-1 images in Galicia wildfires;
- (b) EFFIS dNBR image in Galicia wildfires;
- (c) Sentinel-2 dNBR image using Post-1 images in Portugal wildfires;
- (d) EFFIS dNBR image in Portugal wildfires. Water areas are masked (blue color).



from:

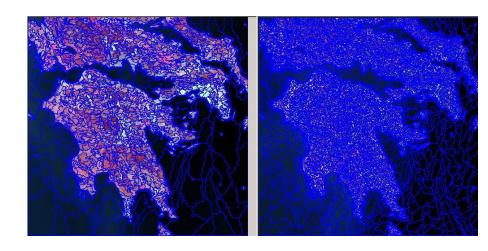
Rafael Llorensa, José Antonio Sobrino, Cristina Fernández, José M.Fernández-Alonso, José Antonio Vega, A methodology to estimate forest fires burned areas and burn severity degrees using Sentinel-2 data. Application to the October 2017 fires in the Iberian Peninsula, International Journal of Applied Earth Observation and Geoinformation Volume 95, March 2021, 102243 https://doi.org/10.1016/j.jag.2020.102243

Methodology and steps for estimating burned areas from forest fires

Step 1: Mapping the area before and after the forest fire

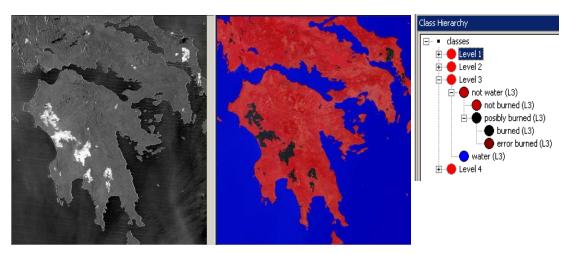


Step 2. Image segmentation



MODIS images prior and post a forest fire (RGB-124)

Step 3. Change detection



Step 4. Potential errors in the delineation of burned areas



Final map of burned areas



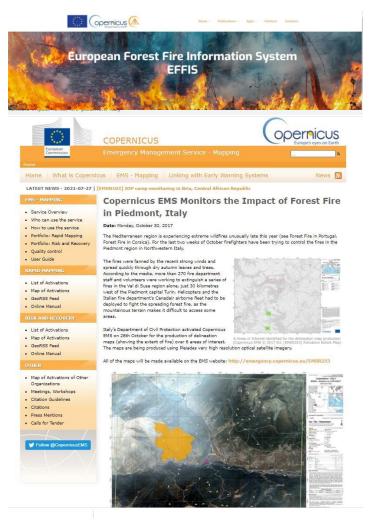
Two days later ... 26,000 m^2 of the forest were burned

Sources of information

The European Forest Fire Information System (EFFIS)

 The European Commission has developed the European Forest Fire Information System (EFFIS)
 (http://effis.jrc.ec.europa.eu/)
 to provide a fire risk forecast and a fire danger assessment in EU countries.

 EFFIS is one of the Copernicus Emergency Services and becomes an essential tool for providing most up-to date information on fire danger in EU



The Global Wildfire Information System (GWIS)



Joint initiative of the Group on Earth Observations (GEO), the NASA Applied Research and the EU Copernicus work programmes. Using advanced methods on data processing for wildfire detection and monitoring, numerical weather prediction models, and remote sensing, GWIS enables enhanced wildfire prevention, preparedness and effectiveness in wildfire management.

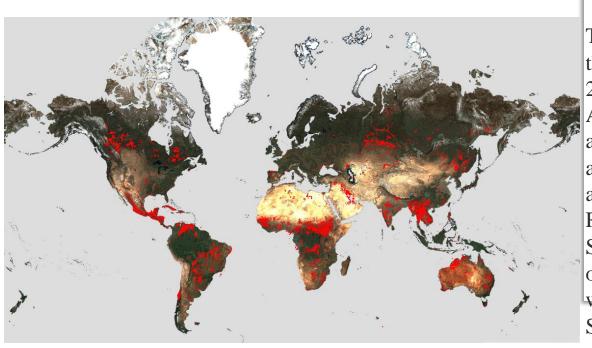
Fire Information for Resource Management System (FIRMS)



NASA's FIRMS distributes Near Real-Time (NRT) active fire data within 3 hours of satellite observation from NASA's Moderate Resolution Imaging Spectroradiometer (MODIS) and NASA's Visible Infrared Imaging Radiometer Suite (VIIRS).

28
https://firms.modaps.eosdis.nasa.gov/

Worldwide fires from ESA's World Fire Atlas



The map shows fires taking place across the globe between May 2016 and June 2023, using data from the World Fire Atlas. The atlas provides a detailed analysis of wildfires across the globe and utilised nighttime data from the Sea Land Surface and Temperature Radiometer (SLSTR) onboard the Sentinel-3A satellite. The data have been overlaid onto ESA's World Cover map which uses data from the Copernicus Sentinel-2 mission from 2021.

Conclusions

- Optical and thermal infrared RS observations → supportive for the pre-fire and post-fires stages
- Fire detection is technically feasible (in mid and thermal infrared; Sentinel 3, SEVIRI on Meteosat, Landsat TM), yet satellites with good temporal resolution have poor spatial one and vice versa → contribution to operational plans in the active fire stage is constrained
- Sentinel 2 and 3 facilitate research and operation applications with respect to forest fires.
 Results are complemented by Sentinel 1 SAR observations
- Low spatial resolution satellites/sensors (VIIRS, MODIS) used for pre-fire risk mapping; satellites of high spatial resolution - during the post-fire stage - may be used instead (Sentinel- 2, Landsat, Worldview, etc.)
- Cubesats reflect a promising development to improve both temporal and spatial resolution
- Several forest fire related applications have been developed in the framework of the EU, ESA, as well in other parts of the world.











Thank you for the attention

