## LPVE23 - WORKSHOP ON LAND PRODUCT VALIDATION AND EVOLUTION

# The Fire Radiative Power (FRP) Inter-comparison framework: an approach to identify product differences for nonsimultaneous detections

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### INTRODUCTION

The current method to systematically validate Earth Observation (EO) based Fire Radiative Power (FRP) products relies mostly on the intercomparison between products **without clearly identifying** one as the **reference dataset**. In addition, due to the highly dynamic nature of fire, comparisons are mostly restricted to near-simultaneous detections between polar and geostationary based products. Here, we propose a new comparison framework that overcomes these limitations by comparing each product's capability in capturing the full fire event **power detection range** and its **characterized fire signature**.

## DATA & METHODS

In this framework, we inter-compare eight operational remotely sensed Fire Radiative Power (FRP) products: eight polar-orbiter products derived from active fires and five geostationary products. This novel approach is based on a **robust statistical analysis of the frequency density (f-D) distributions** of each product's active fire detections, whereby an **Inverse-gamma distribution** is used to characterize the **fire statistical signature**, providing a reference baseline on to which FRP products can be compared, and their representation uncertainty assessed.

	Platform/		Spatial	Temporal	Spatial			
Product Name	Sensor	Spectral channel	resolution	coverage	coverage	Units	Unc	Reference/Data provider
MSG FRPPixel	MSG/SEVIRI	3.92 [range 3.48-	3-5.3 km	2005-2021 (15 min)				
MSG IODC FRPPixel	MSG/SEVIRI	4.36]		2017-2021 (15 min)	Hemispheric		Y	
KCL/IPMA_GOES17	GOES-17/ABI	3.90 [range 3.80- 4.00]	2-3.5 km	2020-2021 (10 min)				Wooster et al., 2012 LANDSAF ( <u>https://landsaf.ipma.pt/</u> )
KCL/IPMA_GOES16	GOES-16/ABI			2018-2021 (10 min)				
KCL/IPMA_HMWR	HIMAWARI/ABI			2019-2021 (10 min)				
MOD14ML	TERRA/MODIS	3.95 [range 3.39 - 3.99]	1 km	2001-2021 (Daily-N/D)	Global	MW	Ν	Giglio et al., 2016 NASA/UMD
MYD14ML	AQUA/MODIS			20032021 (Daily-N/D)				
VNP14ML	SUOMI-NPP/VIIRS	3.74 [bw 0.38]	750 m	2012-2019 (Daily-N/D)				
VNP14IMGML	SUOMI-NPP/VIIRS	3.70 [bw 0.18]	375 m	2014-2021 (Daily-N/D)				( <u>Iuoco.geog.uma.eau</u> )
			1 km	04/2021-06/2022	16 ROI		Y	Wooster et al., 2012
S3A_L2_FRP	Sentinel-3A/SLSTR	S7 or F1 at 3.74 [bw 0.38]		(Daily-N)				EUMETSAT
				04/2021-06/2022				(https://eoportal.eumetsat.i
S3B L2 FRP	Sentinel-3B/SLSTR			(Daily-N)				nt/cas/)
				03/2020-02/2021				Xu et al., 2020
C3S FRP V1.1	Sentinel-3A/SLSTR			(Daily-N)	Global			C3S
				03/2020-12/2022				(https://cds.climate.coperni
C3S FRP V1.1	Sentinel-3B/SLSTR			(Daily-N/D)				cus.eu/)

## RESULTS

 The rollover parameter show the degree of under-representation of each sensor's detections, which typically precludes the detection of a proportion of the highly numerous but individually relatively small and/or low intensity fires. Shape parameter identify the relation between each detection within the same sensors.



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• Spatial variability of the product's **shape parameter are similar**, but can differ in magnitude. Differences are mainly driven by what each product grid-cell coverage, with SOUMI/VIIRS and MSGs having highest coverage, and characterized by higher shape parameter values. In terms of the **rollover parameter**, the products magnitudes are driven by each **sensor' spatial resolution**.



• For Geostationary-based FRP detections there is a **dependency of the pixel size** on the position of the roll-over

We focus on **annual detection periods** and perform the analysis at **0.5° grid cell** resolution (with 100 min counts), for the full time-series of each product/ The results are analysed for their **temporal and spatial consistency**, and inter-product differences are analysed in the context of its drivers.



**Shape parameter,** (ρ) Right hand side exponent associated with events. powerlaw relation associated with nature's self-organized critical behaviour.

#### Location parameter, (a)

Identifies the FRP threshold below which detection relations detach from a power-law relation

#### **Rollover parameter, (r)**

Identifies the FRP threshold below which a detection algorithm/sensors are limited

#### Scale parameter, (s)

Left hand size exponent associated with the rate of decay of detections

## **CONCLUSIONS & FUTURE WORK**

 $p(FRP \mid \rho, a, s) = \frac{1}{a\Gamma(\rho)} \left[ \frac{a}{FRP - s} \right]^{\rho + 1} \exp\left[ -\frac{a}{FRP - s} \right]$ 

This proposed framework is a useful tool to compare non-simultaneous based FRP products, contribute to identification of the main FRP product uncertainty sources, and provide a knowledge management tool for identifying where and when product inconsistencies occur. Preliminary results show that:

parameter, meaning that detections below this **threshold** are **under-represented**. The pattern can be clearly seen in the mean rollover parameter by VZA bins. This means that **compatible representation** between the MSG AF and IO is only possible within a narrow longitude band.



Differences in the fitted parameters **between identical sensors** can be identified. Although minimal compared with non-identical sensors, these need to be further investigated for their causes as they can be attributed to the **environment**, **sensor or algorithm**.





- **Spatial consistency** in the shape parameter between the products
- **Temporal consistency** in the rollover parameter for each product.
- Geostationary satellite based detections are under-represented at larger VZA angles.
- Differences can be identified between detections made by Identical sensors.
- Large shape parameter differences can highlight potential algorithm/sensor issues. Further work should focus on assessing the impact of some of the key non-fire effects, such as: pixel size, pixel area growth off-nadir, algorithm limitations, quality information, and its uncertainty sources. Understanding each product limitations is essential for deriving higher level (L3/L4) products, in this context, this approach is a contribution for the retrieval of conversion factors for product harmonization.

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 Large differences in the shape parameter could identify potential error, either refer to false detections or due to sensor saturation. The example shows suspicious areas where the SUOMI/VIIRS – S3A/SLSTR-NTC differences are higher than 1.5

