

Consiglio Nazionale delle Ricerche

# KNOWLEDGE BASED FUSION OF REMOTE SENSING TIME SERIES WITH HETEROGENEOUS SPATIO TEMPORAL RESOLUTION: AN APPLICATION FOR CROP CLASSIFICATION AND MONITORING

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H0

H1

### 1. Problem Faced

Appropriate agricultural Remote Sensing monitoring needs high resolution multispectral data with dense revisiting cycles, which are actually not operational. Currently, only high spatial resolution (*H*) data (< 30 *m*) but temporally sparse due to both revisiting time (16 to 26 days) and cloud contamination or quasi-daily data with low or very low spatial resolution (*L*) (250-1000 *m*) are available for agricultural Remote Sensing monitoring.

# 2. Objectives of the Work

Exploiting the agronomists expert knowledge on the crops growth dynamics and practices to perform the fusion of the available heterogeneous Remote Sensing time series so as to generate new temporal series of data characterized by better spatio-temporal information than the original ones with a desired timestamp.

### 3. Study Area and Dataset

A fusion experiment has ben designed considering as input two time series in an **agricultural area of Lombardy (Italy)**. The **H series has been partitioned into two sub-series**, one for the fusion (**input series**) and the second for the testing (**target series**).

# **4. Proposed Fusion Procedure**

**Time validity weights** 

SERIES	SENSOR	# IMAGES	SPAT. RES.	FREQUENCY
<b>H</b> <sub>input</sub>	Landsat8	7	30 <i>m</i>	16 days
<b>H</b> <sub>target</sub>	Landsat8	6	30 <i>m</i>	16 days
<b>L</b> <sub>input</sub>	MODIS	14	500 <i>m</i>	8 days



# $\mu t1(th0) \text{ of } H0$ $\mu t1(tL2) \text{ of } L2$ $\mu t2(tH1) \text{ of } H1$ $\mu t2(tL3) \text{ of } L3$ $\mu t1(tH0)$ $\mu t1(tL0)$ $\mu t2(tH1)$ $\mu t2(tL3)$ $\mu t_{t}(t_{H}) =$ $\left| t_{0} - t \right| \text{ a depend}$

 $\mu_t(t_H) = \begin{cases} \frac{t_H - t_O}{t - t_O} & if \ t_O \leq t_H < t\\ \frac{t_H - t_E}{t - t_E} & if \ t \leq t_H < t_E \end{cases}$ 

 $|t_o - t|$  and  $|t - t_{E}|$  can change depending on the seasonal variation of crop growth

- **1.** Definition of desired Output Timestamps:  $t_1, t_2, ..., t_n$
- 2. Definition of temporal Validity of Input images at output Timestamps based on Knowledge

For each timestamp *t*, the validity of each input image  $H_i$  or  $L_i$  acquired before  $t_o$  and after  $t_E$  is Null, otherwise is  $m_t(t_H)$  and  $m_t(t_L)$ 

### 3. Pixel level Fusion

At each timestamp *t*, the fusion is a weighted average of input images within  $t_0$  and  $t_E$ , using validity as weights.

### **5. Results**

We applied the fusion defining a desired timestamp corresponding to the dates of the target series and generating the simulated series which has been compared with the target series.

Statistics (Simulated vs Target)				
R	RMSE	Accuracy		
0.783	0.105	0.931		



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