FLAGSHIP 2.6.3: AI ALGORITHM FOR (SATELLITE) IMAGING RECONSTRUCTION

REPORT FOR WP6 MEETING, 23/01/2024 A. Tricomi^{1,2,3}, G. Piparo¹, G. A. Anastasi², E. Tramontana², V. Strati⁴

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ICS



GENERAL STATUS OF THE FLAGSHIP

- The Spoke 2 general meeting was a good opportunity to get to know each other and discuss this year's work and also to show the current advances of the flagship! First discussion points also opened with related IG responsible persons.
- > The general state is **quite good** despite the somewhat slower start compared to other projects.
- > The flagship work can be divided into three interconnected branches:
 - > 1) Identification of objects using satellite images (**Emiliano and his group**).
 - > 2) Study of vineyard diseases using aerial images (Virginia and her group).
 - 3) General study of techniques for analysing satellite images (Alessia, myself and Gialex).
 Currently, attempts to replicate/integrate the results of group 2) using satellite images.
- We are confident that the coordinated work of these three groups can complete the objectives of the flagship.

ILESTONES AND KPI OF FLAGSH Centro Nazionale di Ricerca in HPC, Big Data and Quantum Computing

2.6.3

Milestones

- 1. M1-M6 (corresponding to MS7): Survey of the State-of-the-Art; tracking of R&D technologies to be used; selection of datasets for use cases (at least one).
 - D1: report on technologies to be used, selection of at least one test dataset.
- 2. M7-M10 (corresponding to MS8): first experimentation with data sources and algorithms, demonstration on the feasibility of choices
 - \succ <u>D2</u>: report on the experimentation and of technical choices; first code repository available
- 3. M11-M24 (corresponding to MS10): Implementation of the selected technology(ies); test and validation on selected dataset(s). Proof-of-Concept deployment.
 - > D3: Report on the work carried out; release of the developed code on public repository.
 - Intermediate report at MS9

KPIs

KPI ID	Description	Acceptance threshold	Current status
KPI2.6.1.1	Publications	2	1*
KPI2.6.1.2	Presentations at conferences	2	
KPI2.6.1.3	Publicly available Code repositories	1	
KPI2.6.1.4	Use case Test Datasets	1	

*D. Marletta, A. Midolo, E. Tramontana. Detecting Photovoltaic Panels in Aerial Images by Means of Characterising Colours. Technologies 2023, 11(6), 174; https://doi.org/10.3390/technologies11060174

WP6 MEETING, ONLINE, 23-01-2024

Centro Nazionale di Ricerca in HPC, Big Data and Quantum Computing

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KPIs

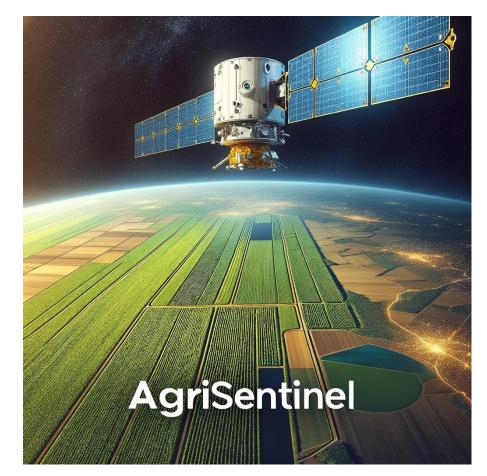
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AGRISENTINEL: A COMPREHENSIVE LIBRARY FOR DOWNLOADING, MANIPULATING AND ANALYSING SATELLITE DATA

- In some of the discussions at the Spoke 2 meeting, the difficulty of managing the downloading and manipulation of satellite data for the realization of datasets using the available frameworks emerged.
- We decided to develop a library that would serve as a unified framework for the preparation of datasets used for training AI techniques, and possibly integrate these techniques within it.
- Development is already underway and the first version of the code is already available on github (currently in private form): <u>https://github.com/gpiparo2/AgriSentinel</u>.
- In addition to being useful internally, the development of this library could be a valuable support for the completion of certain KPIs.





- Agrisentinel is a modular library in python, currently consisting of 4 modules:
- 1) AgriSentinel_Download: Allows the download of satellite data using the sentinelhub API. Currently available are Sentinel2-L2A, but other satellites (Copernicus Constellation, LandSat, other commercial satellites) can also be integrated.
- 2) AgriSentinel_DataManipulator: Allows data manipulation for the creation of datasets (with images in tiff format) of multispectral images and vegetation indices (currently 19 have been implemented). It also allows images to be loaded into lists and numpy arrays to be used for analysis, e.g. for training and testing AI models.
- 3) **AgriSentinel_Visualiser:** Allows multispectral images and vegetation indices to be printed out for visualization.
- 4) Agrisentinel: serves as an interface for the previous modules.



A (BRIEF) EXPLANATION OF HOW AGRISENTINEL WORKS

config=SHConfig("peppe") agri_sentinel = AgriSentinel(config=config, data_folder="prova/data", processed_data_folder="prova/data_off", images_folder="prova/images") geojson_path='geojson_archive/vineyard_area.geojson' with open(geojson path) as f: geojson = json.load(f) # Name of the area area_name = "FormigineVineyard" #Downloading Images agri_sentinel.download_images(date_from=datetime(2023, 8, 1), date_to=datetime(2023, 8, 31), interval_days=5, location_geojson=geojson, mosaicking_order="leastCC", resolution=10) [] agri_sentinel.manipulate_data(area_name=area_name, config_file="configuration/indices_config.json") [] \triangleright \sim agri_sentinel.visualize_data(area_name, bands='RGB', crop_mask_path=geojson_path, crop=True) [] ms_array=agri_sentinel.prepare_spectral_bands_dataset_array(area_name, crop_mask_path=geojson_path, crop=True) vi_array=agri_sentinel.prepare_vegetation_indices_dataset_array(area_name, crop_mask_path=geojson_path, crop=True)



A (BRIEF) EXPLANATION OF HOW AGRISENTINEL WORKS

		"calculate": true,	
		"description": "Modified Chlorophyll Absorption in Reflectance Index",	
		"formula": "((RedEdge1-RED)-0.2*(RedEdge1-GREEN))*(RedEdge1/RED)"	
		}.	
\triangleright ~	# Credentials configuration	"MCARI1": {	
		"calculate": false,	
	<pre>config=SHConfig("peppe")</pre>	"description": "Modified Chlorophyll Absorption in Reflectance Index 1",	
		"formula": "1.2*(2.5*(NIR1-RED)-1.3*(NIR1-GREEN))"	
	# AgriSentinel object creation	}.	
	agri sentinel = AgriSentinel(config=config, data folder="prova/data", processed data folder="	"MCARI2": {	
		"calculate": false,	
	# Definition of the area of interest via a GeoJSON object	"description": "Modified Chlorophyll Absorption in Reflectance Index 2",	
		"formula": "1.5*(2.5*(NIR1-RED)-1.3*(NIR1-GREEN))/sqrt(((2.0*NIR1+1)**2)-(6*NIR1-5*sqrt(RED))-0.5)"	
	<pre>geojson_path='geojson_archive/vineyard_area.geojson'</pre>	Ъ	
	with open(geojson_path) as f:	"NDRE": {	
	<pre>geojson = json.load(f)</pre>	"calculate": true,	
		"description": "Normalized Difference Red Edge Index",	
	# Name of the area	"formula": "(NIR1-RedEdge1)/(NIR1+RedEdge1)"	
	area name = "FormigineVineyard"	},	
		"NDVI": {	
		"calculate": true,	
	#Downloading Images	"description": "Normalized Difference Vegetation Index",	
	<pre>agri_sentinel.download_images(date_from=datetime(2023, 8, 1), date_to=datetime(2023, 8, 31),</pre>	"formula": "(NIR1-RED)/(NIR1+RED)"	
[]		b	
	# Manipulation of downloaded data		
	<pre>agri_sentinel.manipulate_data(area_name=area_name, config_file="configuration/indices_config.</pre>	ison"	
[]			
N			
\triangleright ~	# Visualisation of Results		
	agri sentinel.visualize data(area name, bands='RGB', crop mask path=geojson path, crop=True)		
	abit_senernernernernernernernernernernernerner		
[]			
	# Preparation of MS and VI arrays		
	<pre>ms_array=agri_sentinel.prepare_spectral_bands_dataset_array(area_name, crop_mask_path=geojson_path, crop=True)</pre>		
	vi_array=agri_sentinel.prepare_vegetation_indices_dataset_array(area_name, crop_mask_path=geojson_path, crop=True)		
[]			



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- Make the code more general to deal with images from multiple satellites (commercial and free).
- Develop a module to manage image analysis (time series, index averages) and AI methods.
- Choosing or developing algorithms for cloud management.
- Adding more vegetation indices or other variables that can be calculated from satellite images.
- > Improving the library structure and writing documentation (ongoing).