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2 **River color monitoring using optical satellite data**

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8 **Abstract:** Knowledge of inland water quality and riverine inputs to oceans is fundamental for water
9 management, environmental monitoring and for the definition of policies and planning strategies related to the
10 sustainable use of rivers. While European Union directives aim at the conservation of inland water resources,
11 the ground operational monitoring network is often inadequate. Rivers monitoring using Remote Sensing may
12 complement in-situ measurements supplying continuous spatially explicit representation of parameters related
13 to water quality and solid transport, even if the high frequency dynamics of water parameters could be not
14 caught due to limited satellite revisit time. Sentinel-2 and Landsat-8 satellites, equipped with MSI and OLI
15 optical sensors whose spectral bands allow to perform a more accurate atmospheric correction, allow to develop
16 methodologies for monitoring river color from space thanks to high spatial resolution and short revisit time.
17 This study presents a processing chain developed to monitor water constituents in rivers using high resolution
18 satellite images. Multitemporal analysis of Chl-a and Total Suspended Matter (TSM) bio-geophysical variables
19 was performed for the case study of Po river (Italy) for the year 2017. Quantitative estimations of water
20 constituents were retrieved from Sentinel-2 optical multispectral satellite data using the C2RCC algorithm, and
21 main outcomes discussed. The developed processing chain can be used to create operational services for river
22 monitoring and represent a major improvement in the identification of spatio-temporal dynamics, like solid
23 transport, in riverine systems.

24 **Keywords:** Po river; river color; water constituents; solid transport; TSM; Chl-a; C2RCC; multitemporal analysis;
25 Sentinel-2 MSI; operational service; biophysical

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27 **1. Introduction**

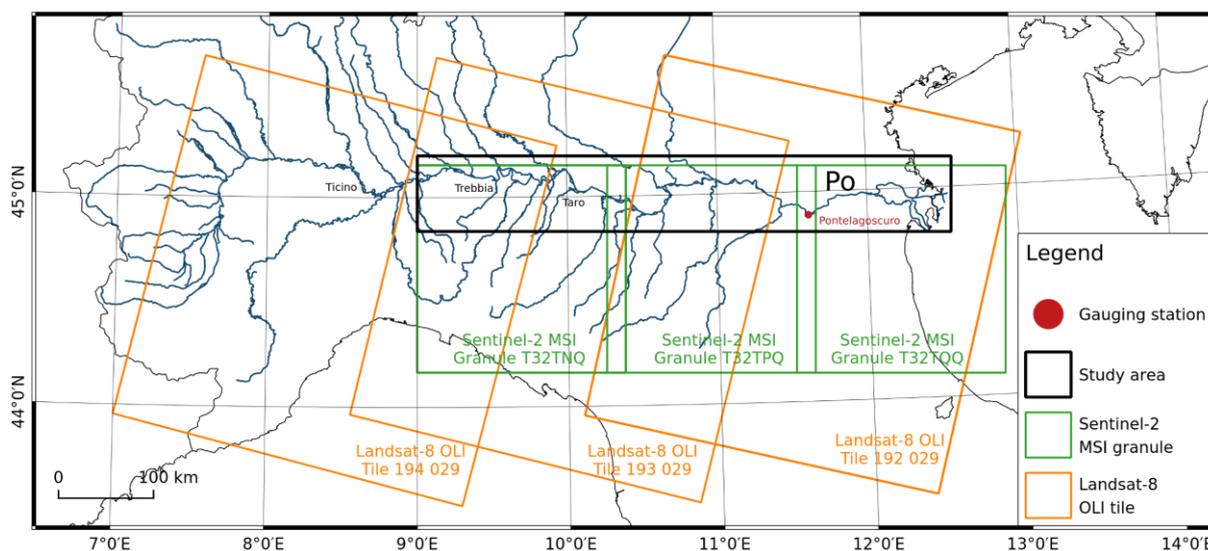
28 The EU Water Framework Directive and Marine Strategy Framework Directive, European establish that the
29 member states monitor the environmental status of their inland and coastal waters. An operational monitoring of
30 water constituents in river channels and river plumes is required to provide quantitative measurements of
31 biogeochemical cycles, explain long-term morphodynamic trends related to sediment transport, and to evaluate
32 the related ecosystem services.

33 Fluxes of water constituents at river channels, confluences and mouths are spatially and temporally highly
34 variable, with periods of high flooding events corresponding to high sediments supply, alternated with periods
35 of low discharge. Variations in discharge rates, solid transport and nutrient pathways are often only roughly
36 estimated due to the lack of available in-situ measurements. When available, field measurements are expensive
37 and specific to a time or geographical location and are thus not fully representative of the spatial and temporal
38 dynamics of water constituents in river systems [1].

39 Monitoring the temporal dynamics of sediment transport in river channels allows to characterize processes
40 like modification of river morphology due to erosion and sedimentation, river bed sediment washout, mudflat
41 and sandbar moves. Channel length and width, braiding and sinuosity indexes, and channel lateral shifting are
42 proved to be the most effective morphometric parameters for a quantitative analysis of river changes [2].

43 Remote Sensing may complement in-situ measurements and morphometric indices analysis in river
44 monitoring by supplying continuous spatially explicit representation of parameters related to water quality and
45 solid transport, even if the high frequency dynamics of water parameters could be not caught due to limited
46 satellite revisit time. Satellite optical multispectral and hyperspectral can be used to retrieve Inherent Optical
47 Properties (IOPs) of river waters, those representing parameters depending only on the dissolved and suspended
48 substances in water. The existing relation between the spectral response and the absorption and backscattering
49 of water constituents at different concentrations, allows to calibrate algorithms for the accurate estimation of bio-
50 geophysical variables like Chlorophyll-a (Chl) and Total Suspended Matter (TSM) using in-situ measurements.

51 Satellite optical multispectral data have been already analyzed in conjunction with automated in situ
52 turbidity measurements, in order to evaluate the accuracy of the TSM algorithms and to systematically monitor
53 changes in TSM distribution [3].
54



55
56 **Figure 1.** Location of the study area, shown with the extent of Sentinel-2 MSI granules and Landsat-8 OLI tiles used
57 for the analysis. Location of the gauging station, measuring discharge data used for the analysis, is represented in
58 a red dot.

59 A proper atmospheric correction procedure plays a key role for the accurate estimation of IOPs, especially
60 for very turbid waters. Sentinel-2 and Landsat-8 satellites, equipped with MSI and OLI optical sensors whose
61 spectral bands allow to perform a more accurate atmospheric correction, allow to develop methodologies for
62 monitoring river color from space thanks to high spatial resolution and short revisit time.

63 More recently, SWIR bands have been used to extend the existing turbid water atmospheric correction of the
64 ACOLITE algorithm to extremely turbid waters. Atmospheric correction based on SWIR bands has been
65 demonstrated to retrieve more accurate water leaving reflectances over extremely turbid waters and improve the
66 algorithm for detecting black suspended sediments from dredging and dumping operations [5]. The ACOLITE
67 algorithm, initially developed to perform atmospheric correction over water from Landsat-8 OLI optical
68 multispectral data and to estimate water constituents from water leaving reflectances, has been extended for the
69 use of Sentinel-2 MSI data.

70 The C2RCC algorithm [6], developed to estimate water constituents from optical multispectral data acquired
71 by many different satellite sensors, relies on a large database (derived from in-situ measurements) of radiative
72 transfer simulations, inverted by neural networks. C2RCC has been validated for the different sensors, with good
73 results for optically complex waters, and additional neural nets have been trained for extreme IOP ranges [6].

74 Time series analysis of TSM retrievals can be used to characterize the TSM dynamics and, thus solid
75 transport, like demonstrated by a recent study conducted across a wide range of Australian lakes [7]. The use of
76 multitemporal series of TSM estimates to compute discharge-based indicators in rivers like the turbidity
77 maximum zone, that can migrate upstream in large rivers under strong tidal forcing, may provide powerful tools
78 to assess future scenarios of morphodynamic trends under climate change conditions [8].

79 Objectives of this study are: i) present a processing chain developed to monitor water constituents in rivers
80 using high resolution satellite images; ii) provide an example of multitemporal analysis of Chl-a and TSM bio-
81 geophysical variables for the case study of Po river (Italy) of year 2017.

82 2. Materials and methods

83 Study area is located in Italy and represented by the last 400 km of the Po river, that collects the runoff of a
84 large drainage basin (about 71000 km²) that has a resident population of 16 million inhabitants. Po river discharge
85 has an annual average rate of 1500 m³ s⁻¹, the minimum flow rate is 275 m³ s⁻¹ and the largest is 11000 m³ s⁻¹.
86 Anthropogenic influence affects river discharge, because of filling reservoirs from snowmelt and rainfall and during
87 hydroelectric power generation for heating. Further, the Po river plain is characterized by and extensive
88 agriculture activities, that influence groundwater and river water levels.

89 While recent works estimated Po river discharge from satellite radar altimetry data [9] and Po river plume
90 dispersion patterns from satellite optical multispectral data [10,11,12,13], analysis of Po river color still must be
91 produced. Ground operational monitoring network exists for the Po river and its tributaries, measuring the water
92 level from many gauging stations. On the other hand, no measurements of turbidity levels or TSM and Chl-a
93 concentrations are systematically acquired.

94 A processing chain has been developed in order to systematically generate maps and multitemporal statistics
95 of water constituents from Sentinel-2 MSI and Landsat-8 OLI. The processing chain can make use of both C2RCC
96 and ACOLITE algorithms [5,6] for the atmospheric correction and IOPs estimation.

97 Essential step in the processing chain is the definition of a water mask formula, based on different rules
98 applied to input TOA radiance spectral bands. A proper water mask should be adopted in order to remove from
99 the analysis those pixels whose spectral contamination of land or terrestrial targets can lead to incorrect estimation
100 of bio-geophysical parameters. For this reason, the default expression for the generation of the water mask
101 available in C2RCC algorithm was refined using the following formula for Sentinel-2 MSI data:

$$B8 < 0.15 \text{ and } B11 < 0.03 \text{ and } B8 < (B3 + 0.03) \text{ and } B8 < (B2 + 0.02) \text{ and } ((B2 + B3 + B4) > 0.20) \quad (1)$$

102 Similarly, the following formula was used to perform water masking in C2RCC with Landsat-8 OLI data:

$$((\text{near_infrared} < 20) \text{ and } (\text{swir_1} < 3) \text{ and } (\text{near_infrared} < \text{blue}) \text{ and } ((\text{coastal_aerosol} + \text{blue} + \text{green} + \text{red}) > 140)) \quad (2)$$

103 Formulas (1) and (2) are reported as implemented in the SNAP processor. Later, bio-geophysical variables
104 of interest are extracted from the resulting algorithm outputs and an optional image sieve filter is applied in order
105 to remove spurious pixels. Multitemporal statistics are finally computed from the multitemporal estimates.

106 3. Results and Discussion

107 With increasing availability of broadband high resolution optical multispectral satellite data it is time to
108 speak more broadly about the concept of "water color", introduced by [14], extending the range of applications to
109 inland water monitoring, with special concern to rivers.

110 The present work shows an application example for the "river color" monitoring using satellite optical
111 multispectral data, carried on using multitemporal analysis of freshwater water constituents for medium to large
112 rivers using high revisit time and high spatial resolution optical satellite data, like Sentinel-2 MSI and Landsat-8
113 OLI.

114 Spatial representation of Chl-a and TSM average concentration computed from all the 326 available cloud-
115 free Sentinel-2A and Sentinel-2B MSI granules acquired during year 2017 (for a total of 166 observed days) is
116 shown in Figure 2. The spatial variability shows how the concentration of suspended sediments is not
117 homogeneous along the river channel, indicating that resuspension and sedimentation processes occur at
118 different discharge regimes.

119 Multitemporal statistics report Chl-a ranges between 13 and 20 mg × m⁻³ on average, and TSM ranges
120 between 15 and 50 g × m⁻³ (Figure 3). The maximum and 90th percentile values calculated from time series of TSM
121 reveal a strong heterogeneity along the river channel in terms of suspended solid transport.

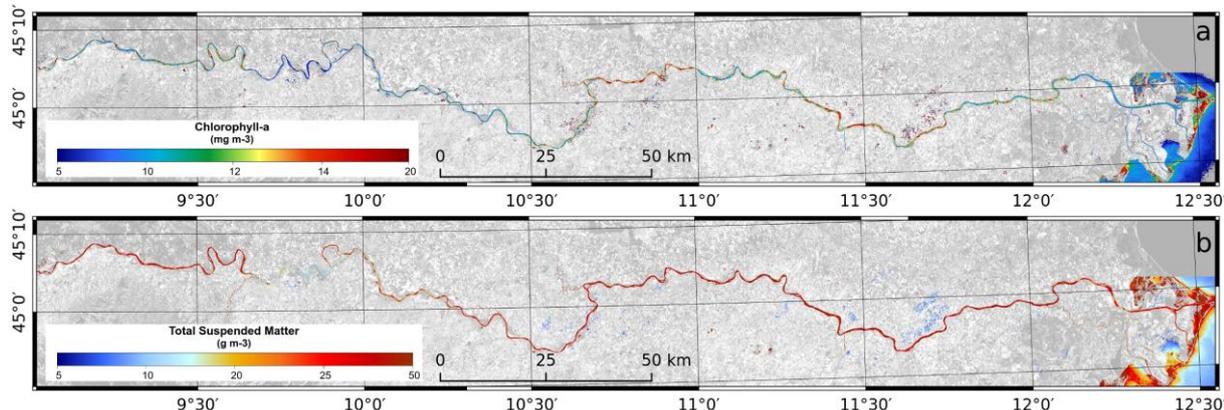


Figure 2. (a) Spatial representation of Chl-a average concentration computed from all the available cloud-free Sentinel-2A and Sentinel-2B MSI data acquired during year 2017; (b) Spatial representation of TSM average concentration computed from all the available cloud-free Sentinel-2A and Sentinel-2B MSI data acquired during year 2017.

This evidence is partly due to the influence of high TSM concentration of river tributary during flood regimes joining Po river. After mixing, the reduced river turbulent flow is not any more capable to keep particles in motion and to transport them, resulting in a subsequent sediment deposition on riverbed and river bars. The fingerprint of such process is visible as a gradient of decreasing TSM concentration along the river channel (e.g. visible from Sentinel-2 acquired on 05/03/2017 and 02/01/2018 at Po river confluences with Trebbia and Taro rivers). From a comparison between Chl-a and TSM concentrations and Po river discharge (Figure 4), it is possible to notice a weaker correlation between physical forcing and system response after the month of June 2017.

A processing chain has been developed in order to generate time series analysis of TSM and Chl-a estimates in river channels. The processing chain needs a proper water mask formula to be defined prior the analysis, in order to avoid the estimation of water constituents in pixels not representing pure water (e.g. river banks, river bars). The default water masking formulas available in C2RCC and ACOLITE algorithms have been refined in order to better account for the removal of not pure water pixels. The pixels used for water constituents retrieval and multitemporal series analysis should be located as far as possible from the shore to minimize the impact of adjacency effects [7] and to avoid the pixel contamination with land or terrestrial targets.

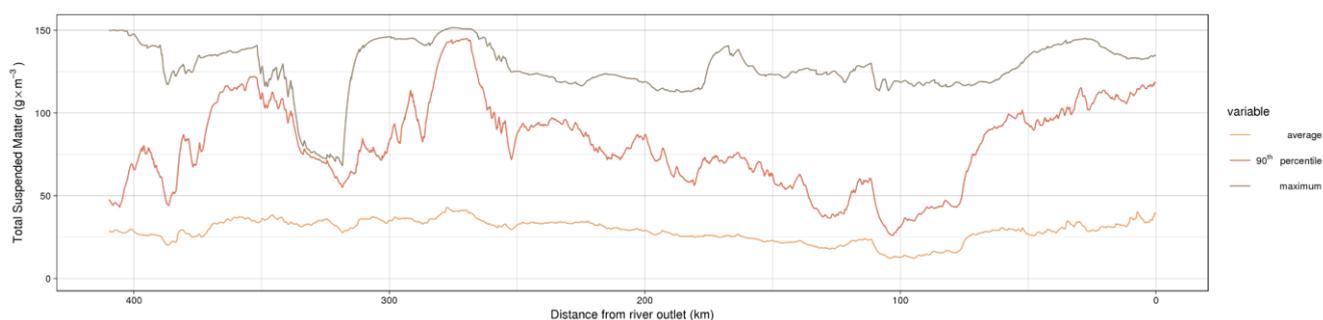


Figure 3. Spatial profile along the last 400 km of Po river showing multitemporal statistics of TSM concentration computed from all the available cloud-free Sentinel-2A and Sentinel-2B MSI data acquired during year 2017.

C2RCC and ACOLITE algorithms, compared to previously developed algorithms for water constituents retrieval, offer the possibility for specific coefficients customization to be applied in the retrieval formulas, in order to better account for regional water cases. The unavailability of in-situ data for the analyzed study area, made coefficient customization and assessment of estimation accuracy impossible and stimulates the need for the establishment of a measurement network of turbidity levels and other bio-geophysical parameters in the Po river. Continuous measurement networks, like for instance the MAGEST and SYVEL in the Gironde and Loire estuaries,

allow to monitor the high-frequency dynamics of the temperature, salinity, oxygenation and turbidity of the water, besides supplying data to calibrate regional algorithms. The insufficient in-situ data-to-imagery matchups availability to evaluate the IOPs retrieval accuracy from multispectral optical data acquired by recent satellites (like Landsat-8 OLI and Sentinel-2 MSI) has been already highlighted in a research work [7].

It is generally acknowledged that the accuracy of satellite matchups is affected not only by uncertainties in bio-optical algorithms, but also by errors in field measurements, spatial and temporal differences between sampling and satellite overflight, and atmospheric correction uncertainties [3].

High revisit time and high spatial resolution of Sentinel-2 MSI and Landsat-8 OLI can be seen as a virtual constellation for observing the rapidly varying aquatic systems [16], even if a proper comparison and intercalibration of algorithms is required to generate a consistent virtual constellation bio-geophysical variables set. Comparison of results from C2RCC and ACOLITE algorithms, done for a total of 10 acquisition dates between 2016 and 2018 in the study area, revealed a weak correlation for estimated bio-geophysical variables. Further analyses are undergoing, even if a proper uncertainty estimation is not possible due to the unavailability of simultaneous satellite in-situ measurements. On one side, the Sentinel-2 MSI 10 m spatial resolution of VNIR bands, allows to extend water color to medium rivers and other small water bodies. On the other hand, Landsat-8 is significantly less affected by sun-glint effects and it has the advantage of higher SNR in its 30 m bands, resulting in a less noisy signal for low reflectances characterizing liquid water.

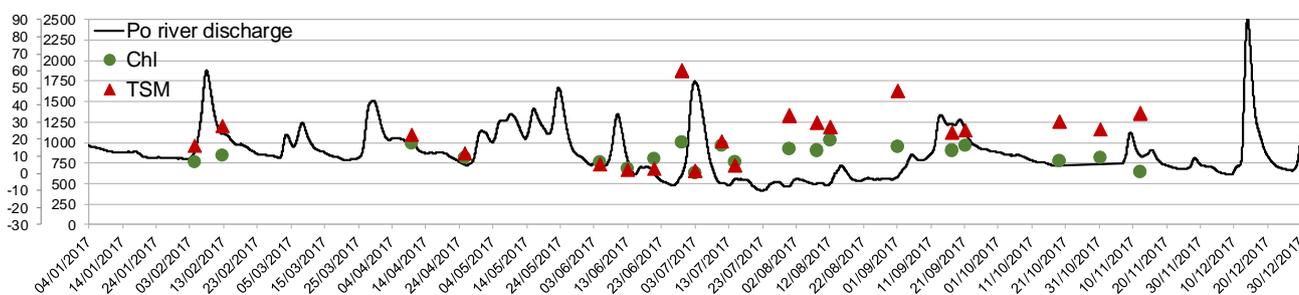


Figure 4. Comparison of Po river discharge and estimated Chl-a and TSM concentration throughout year 2017.

Sun-glint affects significantly Sentinel-2 MSI data at specific viewing azimuth angles, especially when approaching summer solstice. In Figure 4 it is possible to notice a biased estimation of both Chl-a and TSM during the period going from June to September, probably due to the sun-glint effect on liquid water surfaces. Considering that Sentinel-2 acquisition mode adopt sub-swaths with different viewing azimuth angles, the sun-glint is also inhomogeneous in the spatial domain, and therefore the observed scene is not affected the same way.

Using Sentinel-2 MSI it is possible to map at least one pure water pixel in a river channel of width 50 m, instead using Landsat-8 OLI it is possible to map at least one pure water pixel in a river channel of width 75 m. Minimum river channel width required for the estimation may vary due to the presence of river bars and riparian vegetation on the river shores that, due to the likely presence of trees with great canopy height, can generate large shadows over the river channel, especially at high latitudes and for temporal frames characterized by large solar zenith angles. Further, there is the need of defining a low concentration thresholds, in order to avoid invalid concentration estimates due to the influence of the riverbed spectral response in the detected spectrum.

4. Conclusions

In this study a processing chain has been developed in order to systematically generate water constituents for the analysis of identification of spatio-temporal dynamics, in order to facilitate the analytical procedures and stimulate the development of operational services for river monitoring. The multitemporal analysis of the Po river for the year 2017 allowed to characterize the spatio-temporal dynamics of TSM and Chl-a bio-geophysical variables. Finally, key points related to technical feasibility of river color monitoring using satellite optical multispectral data are discussed and directions for future development are provided.

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195 **Conflicts of Interest:** The author declares no conflict of interest.

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