

1 Article

2 In Situ and Satellite Observation of CDOM and 3 Chlorophyll-a Dynamics in Small Water Surface 4 Reservoirs in the Brazilian Semiarid Region

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17 **Abstract:** We analyzed Chlorophyll-a and CDOM dynamics from field measurements and
18 assessed the potential of multispectral satellite data for retrieving water-quality parameters in
19 three small surface reservoirs in the Brazilian semiarid region. More specifically, this work
20 comprises i) analysis of Chl-a and trophic dynamics; ii) characterization of CDOM; iii) estimation
21 of Chl-a and CDOM from OLI/Landsat-8 and RapidEye imagery. The monitoring lasted 20 months
22 within a multi-year drought, which contributed to water-quality deterioration. Chl-a and trophic
23 state analysis showed a highly eutrophic status for the perennial reservoir during the entire study
24 period, while the non-perennial reservoirs ranged from oligotrophic to eutrophic, with changes
25 associated with the first events of the rainy season. CDOM characterization suggests that the
26 perennial reservoir is mostly influenced by autochthonous sources, while allochthonous sources
27 dominate the non-perennial ones. Spectral-group classification assigned the perennial as
28 CDOM-moderate and highly eutrophic reservoir, whereas the non-perennial ones were assigned
29 as CDOM-rich and oligotrophic-dystrophic reservoirs. The remote sensing initiative was partially
30 successful: the Chl-a was best modelled using RapidEye for the perennial; whereas CDOM
31 performed best with Landsat-8 for non-perennial reservoirs. This investigation showed potential
32 for retrieving water quality parameters in dry areas with small reservoirs.

33 **Keywords:** water quality; eutrophication; trophic state index; Landsat-8; RapidEye; tropical inland
34 water bodies; Brazil

36 1. Introduction

37 The Brazilian semiarid region has a dense network of surface reservoirs, with storage capacities
38 typically ranging from 1 to 10³ hm³. The small (< 10 hm³) and medium-sized reservoirs (10 – 50 hm³)
39 prevail. They are embanked streams that collect water in the rainy season (January to May) to secure
40 water supply during the dry season (June to December). In fact, about 90 % of the water demand in
41 the state of Ceará is supplied by surface reservoirs [1]. However, these systems show low standards
42 of water quality, mainly due to eutrophication, which causes water use conflicts and calls for an
43 improvement of water resources management [2]. Due to operational and financial difficulties,
44 conventional *in situ* monitoring of water quality in such a complex system is limited in spatial
45 coverage and sample representativeness. Therefore, remote sensing is seen as a complementing for

46 traditional approaches of water resources monitoring, providing a synoptic view of environmental
47 systems.

48 Satellite remote sensing (RS) has been used for inland water monitoring in the Brazilian
49 semiarid region with regard to water extents and bathymetries of surface reservoirs [3–5]. Also for
50 quality monitoring, several authors presented encouraging results for satellite RS application in
51 rivers and large lakes of different locations [6–14]. However, there are only few studies applying
52 satellite RS to small complex inland water bodies such as the water-surface reservoirs in the
53 semiarid Brazil, where monitoring and understanding magnitude and composition of optically
54 visible water constituents, in particular the concentration of Chlorophyll-a (Chl-a) and Colored
55 Dissolved Organic Matter (CDOM), are highly relevant given the dense population that relies on the
56 water supply.

57 The photosynthetic pigment Chl-a serves as a *proxy* for phytoplankton and is one of the major
58 indicators of the trophic state [15]. RS algorithms for deriving inland water Chl-a concentration
59 mainly use the absorption in the red wavelength region versus the near-infrared (NIR) reflectance in
60 case of eutrophic lakes [16]. The maximum red-absorption around 675 nm of Chl-a has been
61 standardized as the most important wavelength to extract information on Chl-a in eutrophic waters
62 [17]. In turn, CDOM is the most abundant fraction of dissolved organic material in natural surface
63 waters, being the major light absorbing constituent in surface waters, in case of high concentrations
64 causing the yellowish-brown color of the water. Besides the light absorbance, organic matter water
65 concentration affects the ecosystem functioning through its influence on acidity, transportation and
66 reactivity of toxic substances, photochemistry and energy supply [15,18–20]. The increase in the
67 CDOM concentration lowers the reflectance values mainly in the blue and the green ranges of the
68 spectrum (especially below ~500 nm) and its absorbance increases exponentially with decreasing
69 wavelength [21]. CDOM in water bodies originates from allochthonous and autochthonous sources.
70 The main source of allochthonous CDOM are humic and fulvic substances derived from terrestrial
71 vegetation and soil in the drainage basin area, whereas the main contributors for autochthonous
72 CDOM are phytoplankton and submersed aquatic vegetation [22–25].

73 The amount and type of Dissolved Organic Matter (DOM) are important factors for
74 understanding the sources of organic compounds and its possible effect on inland water ecology.
75 Autochthonous CDOM is more biologically labile, less chromophoric and constituted by molecules
76 of low molecular weight. Therefore, it is less affected by photodegradation and more susceptible to
77 biodegradation. Contrastingly, terrigenous CDOM is less biologically labile and has aromatic
78 molecules, being more resistant to overall degradation but more susceptible to photodegradation
79 processes [21,24–26]. The specific absorption coefficient (a_{CDOM}) provides the magnitude of DOM
80 influencing the transparency of column water [15,19]. Spectral slope (S) describes the exponential
81 reduction of absorption at a given wavelength range, being an important parameter for studying
82 CDOM type and CDOM dynamics. Spectral slope is used as an indicator for the CDOM source
83 (allochthonous *versus* autochthonous) and for the status of degradation. The spectral slope ratio (Sr)
84 between the ultraviolet region (275–295 nm) and the visible region (350–500 nm) is used as weight
85 molecular indicator, source and photobleaching of CDOM [21,25–28].

86 Presently, there is a scientific community effort in designing and validating RS algorithms to
87 assess CDOM [8–11,16,29–30]. Nonetheless, there is no universal wavelength and/or spectral range
88 to retrieve CDOM, since it may vary according to spectral interference levels of high concentrations
89 of other optically active constituents in water. Inland water RS has been a challenge, because these
90 environments are optically complex and their optical properties are highly variable (even within the
91 same water body), limiting the development of algorithms and their applicability [30,31].

92 Eutrophication corresponds to an imbalance in aquatic environments due to excess of nutrients,
93 mainly phosphorus and nitrogen [15]. This is even more harmful in drylands, such as the Brazilian
94 semiarid region, due to the low runoff coefficient and the excessive evapotranspiration leading to
95 recurrent droughts, which causes water level reduction [5]. Consequently, the water residence time
96 of the reservoirs is high (2 – 3 years on average) and water quality is often low [32–34]. Land
97 use/land cover changes in the sub-catchments, including agricultural runoff and poor sewage

98 treatment infrastructure, are interfering factors leading to environmental vulnerability of water
99 bodies and anthropogenic eutrophication [34–36].

100 The Trophic State Index (TSI) was initially proposed for temperate regions by [37] and
101 comprises Chl-a, phosphorus concentration and Secchi disk depth measurements. Later, the trophic
102 state classification was adjusted for warm-water tropical lakes [38,39] and some authors have also
103 applied it to the tropical semiarid reservoirs obtaining satisfactory results [32,34,36,40].

104 Few studies address the relation between CDOM and other trophic level indicators. However,
105 organic compounds are one important factor affecting phosphorus precipitation and mobilization in
106 the aquatic system [15,41]. Phosphorus is recognized as a limiting nutrient of eutrophication
107 process, while DOM contributes to control this nutrient, thus regulating the trophic level of the
108 water body. Moreover, CDOM concentration interferes in the transparency of the water and
109 consequently in the availability of light in the water column, regulating the photosynthetic activity.
110 The term "dystrophy" is used as a category for systems rich in humic substances and these present
111 low productivity associated with high humic color [15] and the trophic levels are associated with
112 different input loads (phosphorus or CDOM) and also related to internal processes. Generally,
113 eutrophic ecosystems receive high concentrations of phosphorus and have high primary
114 productivity, whereas dystrophic systems have a higher intake of CDOM and the bacterial
115 metabolism in these systems can be suppressed by low pH [19,20,42].

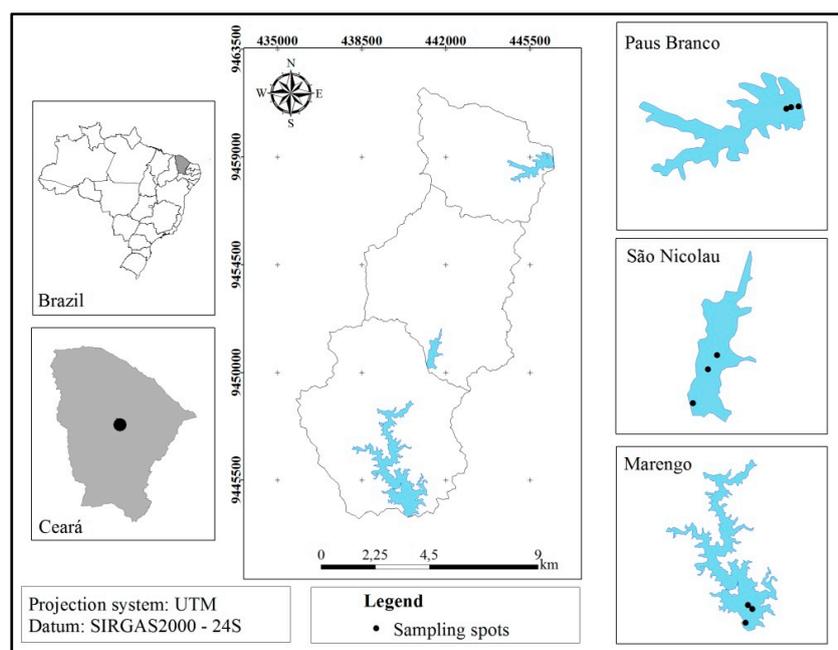
116 This study aims at analyzing the dynamics of Chl-a and CDOM from *in situ* measurements as
117 well as at assessing the potential of multispectral satellite data for retrieving Chl-a concentration and
118 specific absorption coefficient of Colored Dissolved Organic Matter (aCDOM) in three small surface
119 reservoirs in the semiarid Brazil region. More specifically, this work comprises three steps: i)
120 analysis of *in situ* Chl-a and dynamics of attributes related to the eutrophication process; ii)
121 characterization of *in situ* Dissolved Organic Matter (DOM) obtained from absorption and spectral
122 slope of CDOM; iii) Estimation of Chl-a and CDOM through OLI/Landsat-8 and RapidEye satellite
123 data. Furthermore, limitations of satellite monitoring of small complex tropical inland water bodies
124 are discussed and recommendations for future in-situ data collections and satellite monitoring
125 strategies presented.

126 2. Materials and Methods

127 2.1. Study area

128 The study focuses on three reservoirs (Marengo, Paus Branco and São Nicolau) in the Madalena
129 Basin (124 km², see Figure 1), nested into the Banabuiú River Basin, State of Ceará, Brazil. This
130 region has a semiarid climate with water scarcity in the dry season (most prominently between July
131 and December) and high inter-annual rainfall variability. The total average precipitation is
132 approximately 600 mm.year⁻¹, the potential evaporation is around 2500 mm.year⁻¹, the monthly
133 temperatures range from 26°C to 28°C. The basin is vulnerable to droughts, and the investigation
134 occurred during a pluriannual recorded drought, which started in 2012 [43].

135 The Madalena Basin is located within the Land Reform Settlement "25 de Maio". The main
136 water source for the local population (around 600 families), for animal supply, for fishing and for
137 small-scale irrigation are the surface reservoirs. The three reservoirs were chosen for this study
138 because they are distinct in terms of size and limnological characteristics. Marengo is a perennial
139 (PR) and highly dendritic reservoir with 75.3 km² of hydrographic basin and a maximum storage
140 capacity of 15.3 hm³, whereas Paus Branco and São Nicolau are non-perennials (NPR) and have 36.1
141 km² and 22.5 km² of hydrographic basin and a maximum storage capacity of 5.5 hm³ and 0.89 hm³,
142 respectively.



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Figure 1. Study site and sampling spots in the reservoirs Paus Branco, São Nicolau and Marengo in the Madalena Basin, Ceará, Brazil.

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The catchment areas of the focus reservoirs are subject to a similar land use and land cover: free access of animals to water bodies, floodplain crops (called “vazantes”), agrochemical use, practice of soil burning, exposed soil (areas prepared for agriculture or livestock farming) and poor sanitary infrastructure. These practices favor the process of eutrophication. No occurrence of macrophytes is noticeable in the Marengo reservoir (Figure 2a), whereas there was a dense presence of emerged macrophytes in Paus Branco (Figure 2b) and occurrence of submersed species in São Nicolau (Figure 2c) during the investigation period.



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Figure 2. (a) Absence of macrophytes in the water surface in Marengo reservoir; (b) High density of macrophytes (free and rooted species, floating, free and submersed) in Paus Branco reservoir; (c) Presence of submersed macrophytes in São Nicolau reservoir (2015).

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2.2. Limnological and Environmental data

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2.2.1. Hydrochemistry and Hydro-optics

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Water samples were collected from boat at a water depth of 30 cm from the surface, stored in cooled styrofoam containers, and carried to the laboratory for the analysis of total phosphorus (TP), orthophosphate (PO_4^{3-}) and chlorophyll-a (Chl-a). The water filtration for the subsequent CDOM lab analysis was carried out in loco through cellulose acetate membranes (pore size $0.45 \mu\text{m}$). Water transparency was measured with Secchi disc (25 cm diameter). The sampling was conducted from May 2014 until January 2016, totaling 16 field campaigns (Table 1) and covering both dry and rainy seasons.

165 *In situ* reflectance spectra were measured with an ASDFieldSpec®3 Hi-Res spectroradiometer
 166 with a field-of-view of 25°, spectral resolution of 1.4 nm and 350-2500 nm spectral range. A
 167 Spectralon reference panel was used to represent a Lambertian surface and to provide the reflected
 168 measurement of the downwelling component. Measurements of the upwelling signal were taken
 169 between 10 a.m. and 2 p.m. and with an azimuth angle away from the sun in accordance with
 170 NASA's protocols. Measurements of water radiance were carried out simultaneously to the water
 171 samplings on 27 December 2014 (in Marengo), on 03 February 2015 (in Marengo and São Nicolau)
 172 and on 11 November 2015 (in Marengo and Paus Branco). The values of the upwelling and
 173 downwelling measurements were normalized into the reflectance, R, using Equation 1, according to
 174 [44].

$$175 \quad R_{\lambda} = \frac{DN_{a,\lambda}}{DN_{r,\lambda}} \quad (1)$$

176 Where R_{λ} represents the reflectance factor (dimensionless); $DN_{a,\lambda}$ is the upwelling spectral
 177 radiance as Digital Number DN measurement; and $DN_{r,\lambda}$ is the downwelling spectral radiance
 178 reflected from the reference panel.

179 Concentration of Chl-a was obtained using a spectrophotometry method from extraction with
 180 acetone 90%, while TP was measured by digestion with persulfate followed by ascorbic acid method
 181 [45]. PO_4^{3-} and Total Suspended Solid (T.S.S.) concentration were also obtained, according to APHA
 182 (op. cit.). The absorbance of CDOM (A_{λ}) was measured within wavelength range 250-800 nm by a
 183 Lambda 950 UV-VIS spectrometer with a 5 cm quartz cuvette. This measurement was converted to
 184 CDOM absorption (m^{-1}) according to Equation 2:

$$185 \quad aCDOM(\lambda) = \frac{2.303 * A_{CDOM}(\lambda)}{l} \quad (2)$$

186 Where aCDOM corresponds to the absorption coefficient of CDOM per wavelength (λ); A_{CDOM}
 187 stands for spectral Absorbance per wavelength (λ) and 'l' is the cuvette path length in meters [46].
 188 Also according to the authors op cit., we applied Equation 3 to determine the spectral slope in both
 189 the ultraviolet SUV (S275-295) and in the visible SVIS (S350-500) domains. Subsequently, the slope
 190 ratio S_r (S275-295: S350-500) was obtained to characterize CDOM types.

$$191 \quad aCDOM(440) = aCDOM(440)^{-S(\lambda_0-\lambda)} \quad (3)$$

192 In Equation 3, aCDOM(440) is the specific absorption coefficient at 440 nm (m^{-1}); and $S(\lambda_0-\lambda)$
 193 equals to the spectral slope over a particular wavelength domain of aCDOM(440) (nm^{-1}).

194 2.2.2. Calculation of the Trophic State Index (TSI)

195 The TSI proposed by [37] was applied, as modified by [38], as expressed in the Equations 4 – 6,
 196 which take into account Secchi dick transparency depth (m); as well as TP and Chl-a concentration
 197 ($mg.m^{-3}$)

$$198 \quad TSI_{SD} = 10 * (6 - \frac{0.64 + Ln(SD)}{Ln(2)}) \quad (4)$$

$$199 \quad TSI_{TP} = 10 * (6 - \frac{Ln(\frac{80.32}{TP})}{Ln(2)}) \quad (5)$$

$$200 \quad TSI_{Chl-a} = 10 * (6 - \frac{2.04 - 0.695 * Ln(Chl-a)}{Ln(2)}) \quad (6)$$

201 The average value among Equations 4, 5 and 6 was assumed as the representative TSI. The
 202 criterion for the classification of the trophic state is: ultra-oligotrophic ($TSI \leq 24$), oligotrophic ($24 <$
 203 $TSI \leq 44$), mesotrophic ($44 < TSI \leq 54$), eutrophic ($54 < TSI \leq 74$) and hypereutrophic ($TSI > 74$) as
 204 adopted by [38].

205 2.2.3. Environmental data

206 Precipitation: monthly data on precipitation were taken from the Madalena city rain gauge
 207 because it is the nearest gauge to the reservoirs. The period of data coincides with that of the field
 208 campaigns (from May 2014 until January 2016), and the data were obtained from FUNCEME [47].

209 2.3. Optical satellite data processing

210 Assessing water quality of inland waters by means of optical RS requires sensors which operate
 211 in the visible to near infrared wavelength range with a sufficient spatial/temporal resolution to
 212 adequately capture the magnitude of the optically active constituents of the water body. The
 213 Landsat program provides more than 40 years of satellite observations and is currently continued
 214 with LDCM (Landsat Data Continuity Mission) operating with Operational Land Imager (OLI) on
 215 board of Landsat-8 (L8). L8 was launched on February 11, 2013 with a revisit time of 16 days, a
 216 spatial resolution of 30 m for bands 1-7 [48]. It has five bands in the visible to near near-infrared
 217 wavelength range: b1 (blue; 435–451 nm), b2 (blue; 452–512 nm), b3 (green; 533–590 nm), b4 (red;
 218 636–673 nm) and b5 (NIR; 851–879 nm).

219 While the Landsat series is considered as a medium resolution mission, RapidEye Earth
 220 Imaging System (REIS) provides a much higher spatial resolution. The RapidEye (RE) products used
 221 in this study have a pre-processing level 3A, i.e., they are orthorectified resulting spatial resolution
 222 of 5 m [49]. RE imagery have five bands in the VNIR interval: Blue (440 - 510 nm), Green (520-590
 223 nm), Red (630-685 nm), Red Edge (690-730 nm) and Near Infrared (760-850 nm). The main feature
 224 that distinguishes RE from most other multispectral satellites is the Red Edge band. It is spectrally
 225 located between the Red band and the NIR band without overlap and it covers the portion of the
 226 spectrum where reflectance of vegetated surfaces drastically increases from the red wavelength
 227 region towards the NIR reflectance plateau providing additional information about vegetation
 228 characterization [50]. The optical satellite datasets were selected on dates closest to the field
 229 samplings, with the restriction of cloud cover lower than 20%. The satellite dataset used in this
 230 investigation consisted of six L8 scenes (path 217/row 63) and eleven RE tiles (2436712 and 2436812),
 231 as indicated in Table 1.

232 **Table 1.** Sampling and image acquisition dates, time interval between sample collection and image
 233 acquisition, and percentage of cloud coverage per scene or tile. The absence of images is indicated by
 234 N.A. (not available)

FIELD SAMPLING DATE	SATELLITE PRODUCT					
	OLI/Landsat-8			REIS/RapidEye		
	Acquisition date	Time interval (days)	Cloud cover (%)	Acquisition date	Time interval (days)	Cloud cover (%)
2014-05-31	2014-06-02	2	9.8	2014-06-09	10	15 (N) / 20 (S)
2014-07-02	N. A.			N. A.		
2014-08-02	2014-08-05	3	3.4	2014-08-05	3	0.3 (N) / 0.9 (S)
2014-08-30	N. A.			N. A.		
2014-09-27	2014-09-22	5	12.5	N.A.		
2014-10-31	N.A.			2014-10-16	15	4.4 (N) / 4.7 (S)
2014-11-29	N.A.			2014-12-07	8	5 (N), 17 (S)
2014-12-27	N.A.			N.A.		
2015-02-03	2015-01-28	6	8.8	N.A.		
2015-03-13	N.A.			N.A.		
2015-18-04	N.A.			N.A.		
2015-05-23	N.A.			N.A.		
2015-06-19	N.A.			2015-06-18	1	0 (N), 2 (S)
2015-09-03	2015-08-24	9	< 1	2015-09-07	4	2 (N)
2015-11-11	2015-11-12	1	16.7	N.A.		
2016-01-13	N.A.			N.A.		

235 Note: N corresponds to North (RapidEye tile 2436712) and S corresponds to South (RapidEye tile 2436812).

236 L8 satellite images were obtained from the United States Geological Survey (USGS) as Landsat
237 Surface Reflectance L1T product that had been preprocessed (georeferenced and atmospherically
238 corrected) using the L8SR software [51]. The atmospheric correction of the RE images was carried
239 out in ATCOR-2 within ERDAS IMAGINE® using the MODTRAN-4 code (MODerate Resolution
240 Atmospheric TRANsmittance Algorithm). For this processing, a pre-defined calibration file was
241 used for flat terrain with an elevation of 200 m and the option for the atmospheric model for "dry
242 tropical areas"[52].

243 We masked out water with higher NIR reflectance to ensure that these pixels were free from
244 floating vegetation and did not include areas of exposed mud flats. Subsequently, we defined
245 Regions of Interest (ROIs) to extract reflectance values at or close to the in-situ sampling sites and
246 calculated an averaged reflectance per ROI based only on open water pixels. We investigated the
247 sensitivity of the spectral L8 and RE bands related to Chl-a, in particular the 'relative greenness', i.e.
248 the height of the green reflectance peak *versus* the absorption by phytoplankton pigments in the blue
249 and red wavelength region of the first and second Chl-a main absorption bands using
250 $[(2 * \text{Green}) / (\text{Blue} + \text{Red})]$.

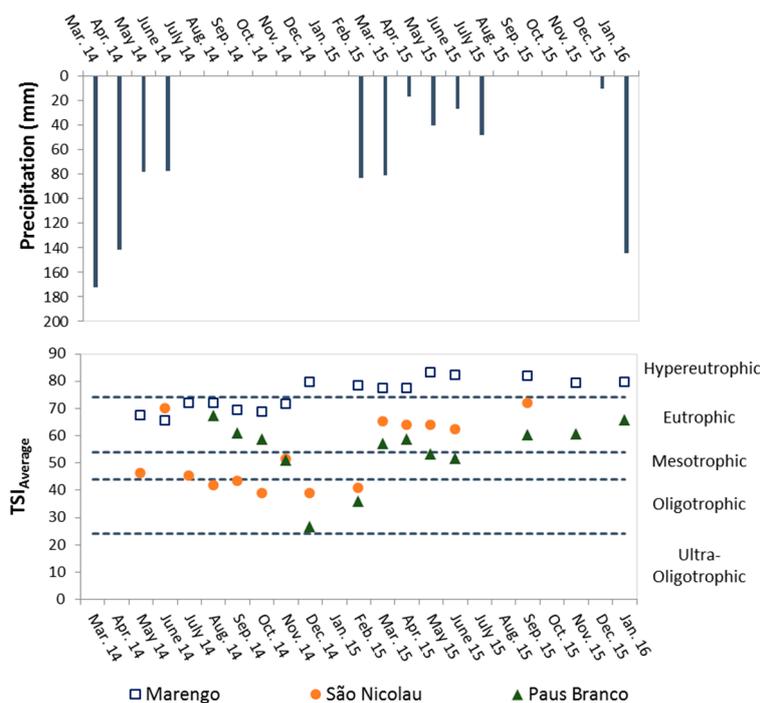
251 For CDOM the Green/Red ratio approach according to [8] was tested. However, high CDOM
252 concentration and therefore absorption in some of the surface waters spectrally flattened the surface
253 reflectance over a very broad wavelength range resulting in a too small range of the ratio values.
254 Therefore, the Blue instead of the Green wavelength region was chosen in order to increase the
255 value range utilizing the Blue/Red ratio, for both satellite products (L8 and RE) and keeping the
256 principle of Kutser's CDOM band ratio algorithm of using a shorter wavelength band with
257 exponentially higher CDOM absorption *versus* the longer wavelength red spectral band.

258 Regressions of the 'relative greenness' versus *in situ* Chl-a and also of the Blue/Red ratio values versus
259 *in situ* CDOM were established. Additionally several bands or bands combination for retrieving
260 these limnological attributes were tested. Nash-Sutcliffe coefficient – NSE [53] was adopted for
261 validation and calibration as a performance indicator for all models tested. RMSE-observations
262 standard deviation ratio (RSR) and Percent bias (PBIAS) were also carried out as model evaluation
263 statistics [54].
264

265 3. Results

266 3.1. Assessment of chlorophyll-a dynamics and trophic state level

267 The results related to trophic state show a high trophic status for Marengo reservoir during the
268 entire study period, reaching its highest value of hypertrophy (TSI = 83.0) at the end of the wet
269 season in May 2015. Paus Branco and São Nicolau reservoirs presented lower values for mean TSI
270 than Marengo and they varied between oligotrophic, mesotrophic and eutrophic level. Both
271 reservoirs showed changes in the trophic level significantly related to the first rains in March 2015
272 (Figure 3).



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Figure 3. Observed precipitation at Madalena rainfall gauge (Ceará) and mean Trophic State Index (TSI) at Marengo, São Nicolau and Paus Branco reservoirs from May 2014 until January 2016.

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SN remained eutrophic, later it completely dried preventing sampling in this reservoir after September 2015 when it was at the limit level of hypereutrophic status. PB showed a highly dynamic behavior. It was eutrophic from August to October 2014, showing better trophic conditions between November 2014 and February 2015 up to reach an oligotrophic level (TSI = 26.8 in December 2014 and TSI = 35.9 in February 2015), returning to a eutrophic level in March and April 2015 (rainy season) and mesotrophic level in May and June 2015 (end of rainy season); from September 2015 and subsequent months PB was classified as eutrophic lake.

The occurrence of submersed macrophytes is usually correlated to lower trophic levels. This trophic level estimate was also supported by Secchi disk depth > 1.0 m from August 2014 to February 2015 for SN reservoir and equal to 0.9 m (on average) from September 2014 to June 2015 for PB that are much higher than those observed for MAR. This parameter indicates the amount of light penetration into the water.

The third limnological attribute considered in TSI estimate is TP that stimulates the growth of algae and is a *proxy* indicator of water fertility. It was above 30 mg.m⁻³ for all reservoirs during the studied period. This value corresponds to the threshold for case 2 waters according to CONAMA 357 [55]. The orthophosphate (PO₄³⁻), organic fraction available to aquatic plants, was higher for PB and SN possibly due to proliferation of macrophytes. The attributes measured in our investigation are summarized in the Table 2.

295
296**Table 2.** Summary of attributes measured during seasonal analyses in Marengo, Paus Branco and São Nicolau reservoirs (Ceará, Brazil).

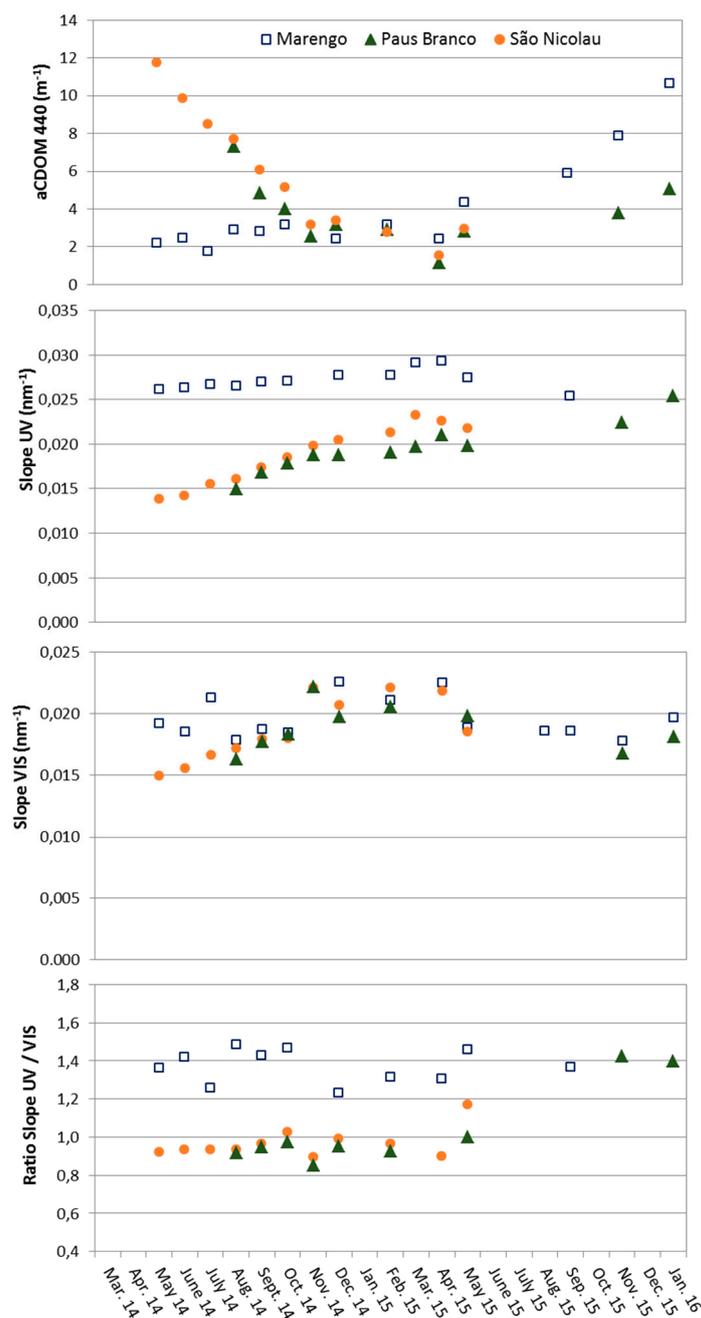
Attributes		Marengo		Paus Branco		São Nicolau	
		Rainy	Dry	Rainy	Dry	Rainy	Dry
Secchi Disk (m)	n	n= 22	n = 31	n=14	n=21	n =17	n = 23
	Min-Max	0.1-0.3	0.1-0.3	0.3-1.5	0.4-1.1	0.4-1.5	0.2-1.5
	Mean±SD	0.18±0.06	0.2±0.07	0.9±0.3	0.7±0.1	1.0±0.4	0.95±0.4
T.S.S. (mg.L ⁻¹)	n	n = 18	n = 12	n = 14	n = 12	n = 15	n = 8
	Min-Max	28.2-90.5	27.8-77.8	2.3-129.0	3.6-14	0.4-26	1.2-101.3
	Mean±SD	47.8±12.6	51.4±12.3	20.0±23.3	7.7±3.4	6.0±6.1	23.7±30.3
Chl-a (µg.L ⁻¹)	n	n = 22	n = 27	n = 14	n = 14	n = 15	n = 22
	Min-Max	28.2-90.5	20.2-263.0	<1-41.0	<1-34.7	<1-25.0	<1-23.2
	Mean±SD	158.2±79.0	80.2±42.1	8.9±10.1	10.6±10.5	5.7±7.2	2.5±3.8
TP (mg.m ⁻³)	n	n = 22	n = 31	n = 12	n = 27	n = 18	n = 23
	Min-Max	161.1-571.1	101-1130	28-236.4	25-501	218-666	204-2851
	Mean±SD	248.2±80.2	266±174	102.4±49	145±97	376±104	527±234
PO ₄ ³⁻ (mg.m ⁻³)	n	n = 22	n = 24	n = 12	n = 18	n = 18	n = 20
	Min-Max	0.0-0.1	0.0-0.2	0.0-0.1	0.0-0.2	0.1-0.4	0.2-2.8
	Mean±SD	0.0±0.0	0.1±0.1	0.0±0.0	0.1±0.1	0.2±0.1	0.4±0.2
aCDOM ₍₄₄₀₎ (m ⁻¹)	n	n = 19	n = 22	n = 8	n = 13	n = 15	n = 18
	Min-Max	2.1-13.0	0.4-8.7	1.1-5.1	2.4-7.3	0.9-12.2	3.1-9.9
	Mean±SD	4.1±2.2	3.3±1.5	3.0±0.6	4.1±1.0	4.5±2.9	6.7±1.9

297 *3.2. Characterization of Dissolved Organic Matter (DOM)*

298 When analyzing the CDOM absorption spectra high variation between the three studied water
 299 bodies and high seasonal dynamics were observed. The values ranged from very low aCDOM(440)
 300 = 0.8 m⁻¹ (for SN, in April 2015) up to high values of aCDOM(440)=12.9 m⁻¹ (for Marengo, in January
 301 2016), as shown in Fig. 4a and also described in Table 2. Marengo reservoir presented aCDOM(440)
 302 equal to 2.6 m⁻¹ on average from May 2014 to April 2015; from this period onwards, there was a
 303 gradual increase reaching its maximum value in the last month of the study period. Different from
 304 MAR reservoir, SN reservoir showed a CDOM decrease from the beginning of samplings until April
 305 2015, when a tendency towards an upward behavior developed. PB reservoir showed a similar
 306 behavior to SN starting at an aCDOM(440) average equal to 7.3 m⁻¹ in August 2014 and gradually
 307 decreasing until 1.1 m⁻¹ (in April 2015) and increasing again in the following months.

308 CDOM showed an inverse relationship with water transparency being remarkably observed
 309 for SN and PB reservoirs. SN had lower Secchi disk measurements in the initial months (from May
 310 to July 2014) with an average equal to 0.5 m and brownish water, showing higher transparency and
 311 clearest water between October and December 2014 with an average depth of 1.3 m. Simultaneously,
 312 aCDOM(440) showed higher values in rain season months in 2014 (10.1 m⁻¹, in average) and had in
 313 general much lower concentrations in the dry period of 2014 (equals to 3.8 m⁻¹, in average) for this
 314 reservoir. The same result is observed for PB where the lower transparency water observed was
 315 equal to 0.5 m in August 2014, whereas aCDOM(440) in this sampling was 7.3 m⁻¹, the highest
 316 aCDOM value for this reservoir. These findings indicate the interference of CDOM concentration in
 317 the transparency of water as also found in the relevant publications on water organic matter
 318 interactions [15,19].

319 Spectral Slope (S) gives information on CDOM structure and composition and also its
 320 susceptibility to photo- and biodegradation. SN and PB showed low S_{UV} values (within the spectral
 321 range between 275 and 295 nm; $S_{UV} = 0.02 \text{ nm}^{-1}$ in average), in contrast to Marengo that showed
 322 higher S_{UV} values. All reservoirs presented a heterogeneous behavior with regard to S_{VIS} (within the
 323 spectral range between 350 and 500 nm) where the values ranged in average from 0.015 nm^{-1} (for SN
 324 reservoir in May 2014) to 0.023 nm^{-1} (for Marengo in April 2015). Very close values between the three
 325 reservoirs studied ($\pm 0.002 \text{ nm}^{-1}$) were observed. Regarding the Slope ratio (Sr), the values ranged
 326 between 0.8 (for SN in June 2014) and 1.5 (for Marengo in August 2014). A similar behavior as for S_{UV}
 327 was observed with Marengo reservoir presenting the highest values in comparison to SN and PB.
 328 All results are shown in Figure 4.



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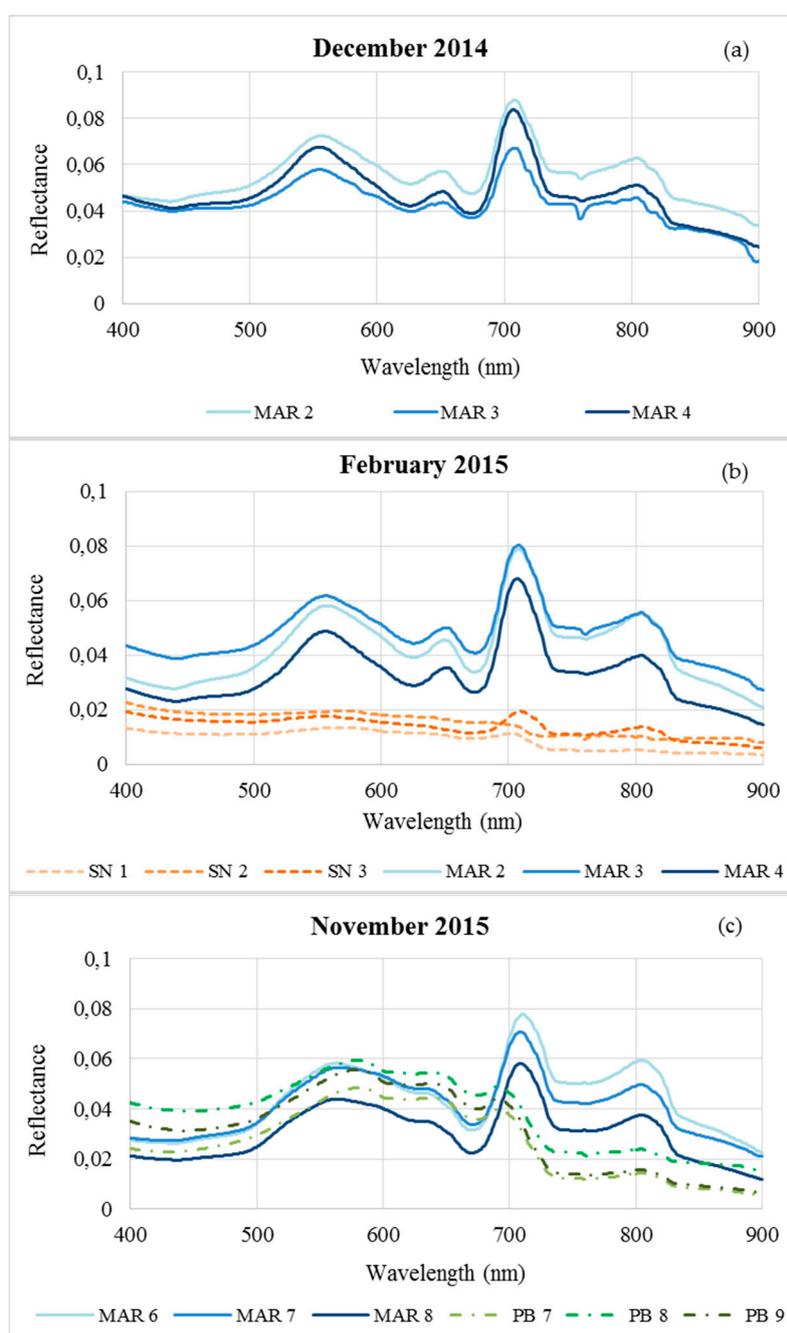
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Figure 4. (A) Mean values of specific absorption coefficient aCDOM at 440 nm; (B) Spectral slope of CDOM in the ultraviolet range (275 – 295 nm); (C) Spectral slope of CDOM in the visible range (350-500 nm) and; (C) Ratio Slope (S_{UV}/S_{VIS}) for Marengo, Paus Branco and São Nicolau reservoirs from May 2014 until January 2016.

334 3.3. Exploratory analysis of *in situ* reflectance data

335 Radiometric *in situ* data are expressed as spectral reflectance (Figure 5). Each measurement was
 336 taken from an average of three readings resulting in a spectral data base of 45 spectral signatures
 337 showing distinct features between MAR, PB and SN reservoirs. Across the spectrum the reflectance
 338 from Marengo was higher than the reflectance from PB and SN reservoirs. On every measurement
 339 date (Fig.5a) December 2014; (Fig.5b) February 2015; and (Fig.5c) November 2015 the reflectance
 340 showed a distinct reflectance peak in the green region (~570 nm), an absorption feature in the red
 341 (~675 nm) and another high reflectance peak around 710 nm. The reflectance spectra from SN in
 342 February 2015 show lowest reflectance values and overall flatness. The reflectance spectra from PB
 343 in February 2015 show high reflectance in the visible but more absorption in the NIR wavelength
 344 region then the reflectance of Marengo.



345 **Figure 5.** *In situ* spectral reflectance of the Marengo (MAR), Paus Branco (PB) and São Nicolau (SN)
 346 reservoirs sampled on (a) December 2014; (b) February 2015; and (c) November 2015.

347 3.4. Satellite-based estimation of Chlorophyll-a and CDOM

348 The reservoirs were separated into two groups, perennial and non-perennial that have different
 349 CDOM and chlorophyll-a regimes to evaluate the capability of L8 and RE reflectance in retrieving
 350 chlorophyll-a and CDOM. The *proxy* for the Chl-a algorithm $[(2 * \text{Green}) / (\text{Blue} + \text{Red})]$ is based on the
 351 principle of combining maximum and minimum reflectance values, since the green wavelength
 352 range is related to a region of low absorption by chlorophyll-a, while the blue and red wavelength
 353 range are domains where energy absorption by chlorophyll-a is high. However, this spectral band
 354 combination (Index 1, I1) showed a low Pearson correlation ($r = 0.1$ for L8 and $r = -0.23$ for RE)
 355 whereas the Blue/NIR ratio (Index 2, I2) and the NIR reflectance (Index 3, I3) had the best Pearson
 356 correlation with values equal to -0.5 and 0.84 for L8 and RE, respectively.

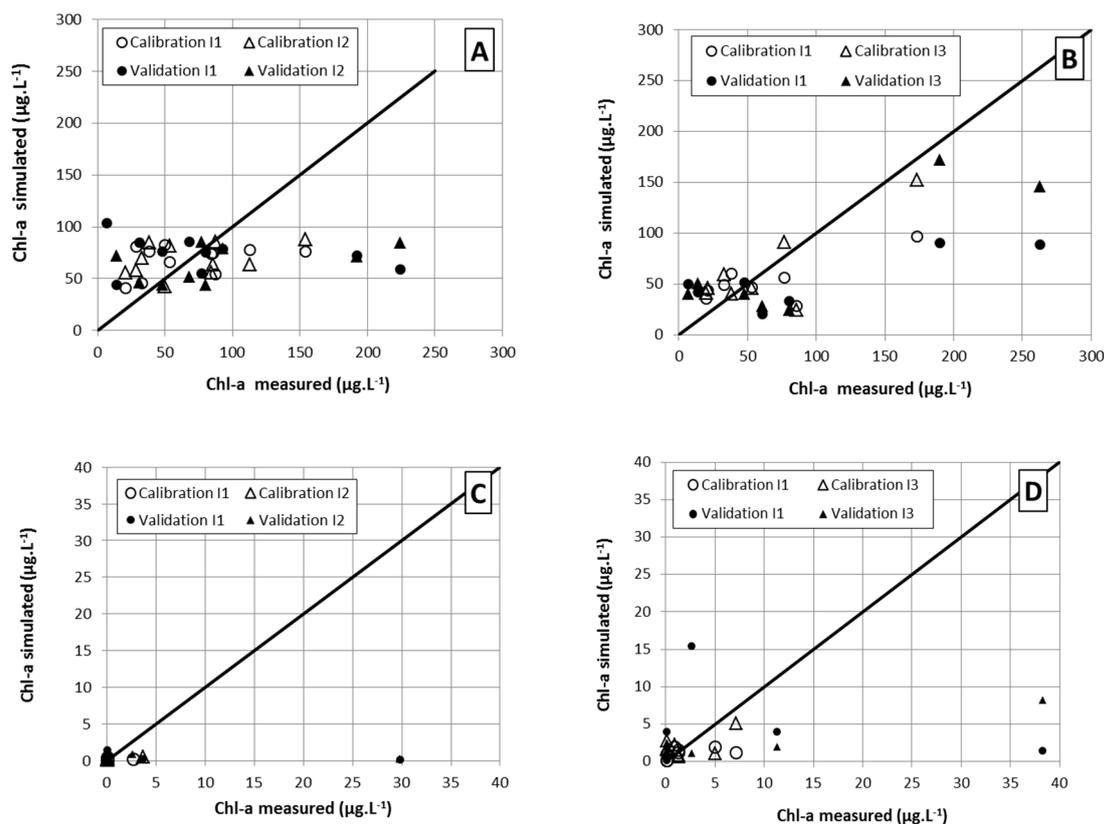
357 The Blue/Red ratio (Index 4, I4) applied for retrieving CDOM showed a Pearson correlation of
 358 0.23 and -0.22 for L8 and RE, respectively. Only the green reflectance (Index 5, I5) showed an
 359 acceptable Pearson correlation for CDOM retrieval this attribute with values of -0.47 and -0.57 for L8
 360 and RE, respectively. These are the highest correlations found and evaluated. Table 3 presents all
 361 models tested for each group of reservoirs (perennial, non-perennials and all reservoirs).

362 **Table 3.** Index and equations applied for estimation of Chlorophyll-a and CDOM from OLI/
 363 Landsat-8 and REIS/RapidEye datasets for Marengo reservoir (perennial reservoir), Paus Branco and
 364 São Nicolau (non-perennials reservoirs).

Attribute: Chlorophyll-a ($\mu\text{g}\cdot\text{L}^{-1}$)				
	OLI/Landsat-8	REIS/RapidEye	OLI/Landsat-8	REIS/RapidEye
	Index I1 = $[(2 * \text{Green}) / (\text{Blue} + \text{Red})]$		Index I2 = (Blue/NIR)	Index I3 = NIR
PR	$\text{Chl} = -74.33 * \text{I1} + 170.09$	$\text{Chl} = 96.52 * \text{I1}^{-1.66}$	$\text{Chl} = -90.12 * \text{I2} + 137.12$	$\text{Chl} = 848.9 * \text{I3} - 6.62$
NPR	$\text{Chl} = 0.07 * \text{I1}^{4.55}$	$\text{Chl} = 1.2 * \text{I1}^{1.15}$	$\text{Chl} = 0.02e^{3.02 * \text{I2}}$	$\text{Chl} = 74.07 * \text{I3} - 0.36$
All	$\text{Chl} = 0.27 * \text{I1}^{11.76}$	$\text{Chl} = 3.89 * \text{I1}^{1.0}$	$\text{Chl} = 14.70 * \text{I2}^{2.94}$	$\text{Chl} = 11951 * \text{I3}^{2.46}$
Attribute: CDOM (m^{-1})				
	OLI/Landsat-8	REIS/RapidEye	OLI/Landsat-8	REIS/RapidEye
	Index I4 = (Blue / Red)		Index I5 = Green	
PR	$\text{CDOM} = 8.04 * \text{I4}^{2.65}$	$\text{CDOM} = 2.25 * \text{I4}^{0.02}$	$\text{CDOM} = 89.82 * \text{I5} - 0.54$	$\text{CDOM} = 2.03 * \text{I5}^{-0.01}$
NPR	$\text{CDOM} = 11.03 * \text{I4}^{5.07}$	$\text{CDOM} = 2.86 * \text{I4}^{-0.30}$	$\text{CDOM} = 0.5437 * \text{I5}^{-0.56}$	$\text{CDOM} = 9.12e^{-12 * \text{I5}}$
All	$\text{CDOM} = 0.16e^{4.09 * \text{I4}}$	$\text{CDOM} = -0.06 * \text{I4} + 0.64$	$\text{CDOM} = 0.59 * \text{I5}^{-0.53}$	$\text{CDOM} = 0.80 * \text{I5}^{-0.24}$

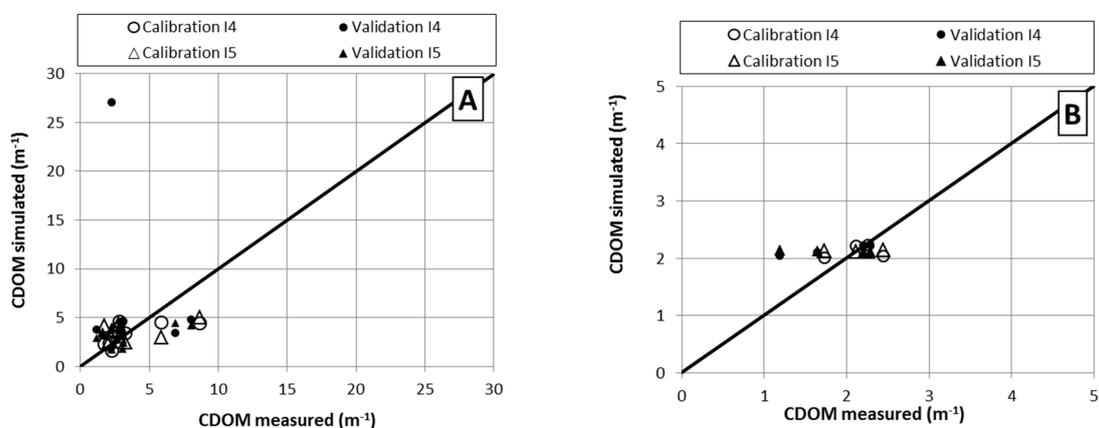
365 PR = Perennial reservoir; NPR = non perennial reservoirs; All = perennial and non-perennial reservoirs

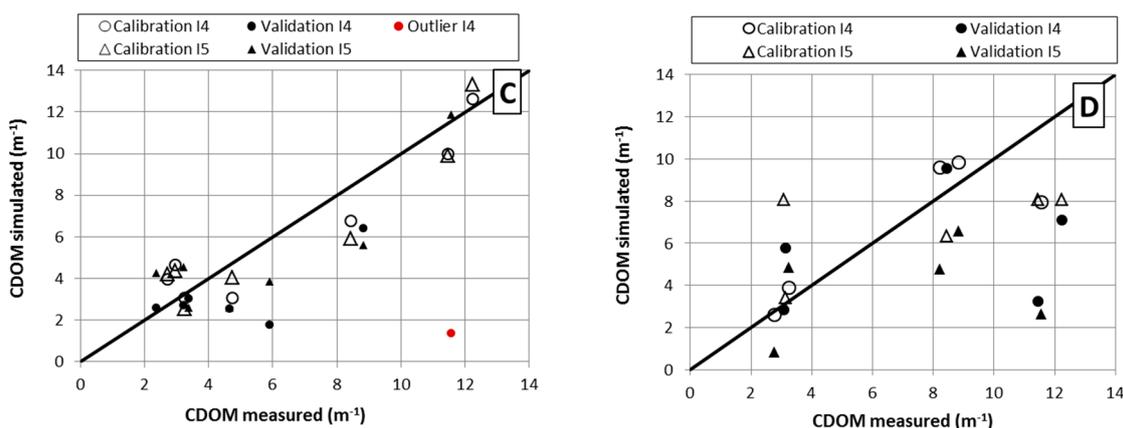
366 The closest relationship between simulated and measured Chl-a concentration was found for
 367 the perennial reservoir data type that is characterized by high phytoplankton concentration and is
 368 successful using the NIR reflectance of RE satellite (Figure 6 B) with reasonable degree of reliability
 369 (NSE = 0.66, RMR = 0.57, PBIAS = 0.24). The fit values smoothly improved when all reservoirs
 370 together were considered with a validation NSE = 0.71, RMR = 0.52 and PBIAS = 0.24. However, the
 371 NIR reflectance showed a low fit for non-perennials reservoirs characterized by low phytoplankton
 372 abundance (NSE = 0.15, RMR = 0.87, PBIAS = 0.31).



373 **Figure 6.** Chlorophyll-a. A. Perennial reservoir & Landsat-8; B. Perennial reservoir & RapidEye; C.
 374 Non-perennial reservoirs & Landsat-8; D. Non-perennial reservoirs & RapidEye.

375 The results of models tested for retrieving CDOM are displayed in Figure 7 and the closer
 376 relationship between simulated and measured CDOM concentration was found for non-perennial
 377 reservoirs data using the Green reflectance of L8 satellite (Figure 7 C) with high degree of reliability
 378 indicated by NSE = 0.89, RMR = 0.56 and PBIAS = 0.09. However this same band had a low fit for the
 379 perennial reservoir (NSE = 0.25, RMR = 0.81, PBIAS = 0.12) and they showed a reasonable value
 380 indicated in the calibration considering all reservoirs (perennial and non-perennial), but this
 381 presented a low fit for validation, demonstrating that the best performance is reservoir-specific.





382 **Figure 7.** CDOM - A. Perennial reservoir & Landsat-8; B. Perennial reservoir & RapidEye; C.
 383 Non-perennial reservoirs & Landsat-8; D. Non-perennial reservoirs & RapidEye.

384 The coefficient evaluation statistics (NSE, RMR and PBIAS) values found for all models tested
 385 are presented in Table 4.

386 **Table 4.** Pearson correlation (r) and coefficient of determination (R^2) values between indexes versus
 387 concentration of Chlorophyll-a or CDOM *in situ* measurements, Nash-Sutcliffe coefficient (NSE),
 388 RMSE-observations standard deviation ratio (RSR) and Percent bias (PBIAS) used for each equation.

Attribute	Chlorophyll-a				CDOM			
	L8		RE		L8		RE	
	I1	I1	I2	I3	I4	I4	I5	I5
r - all reservoirs	0,10	-0,23	-0,50	0,84	0,23	0,22	-0,47	-0,57
R^2 - PR reservoir	0,12	0,24	0,13	0,66	0,50	0,13	0,15	0,00
R^2 - NPR reservoirs	0,14	0,54	0,38	0,29	0,75	0,10	0,77	0,29
R^2 - all reservoirs	0,28	0,20	0,13	0,40	0,69	0,25	0,48	0,65
NSE - PR reservoir	-0,21	0,15	0,15	0,62	-13,54	-0,15	0,25	-0,47
NSE - NPR reservoirs	-0,18	-0,32	-0,16	0,15	0,17	-0,32	0,89	-0,79
NSE - all reservoirs	-3,89	-0,40	-0,45	0,71	-80,28	-1,18	0,38	0,28
RMR - PR reservoir	1,05	0,85	0,87	0,57	0,73	0,93	0,81	3,58
RMR - NPR reservoirs	1,02	1,06	1,01	0,85	1,29	1,03	0,56	1,20
RMR - all reservoirs	2,15	1,14	1,17	0,52	8,73	1,39	0,76	0,80
PBIAS - PR reservoir	0,12	0,43	0,31	0,24	0,80	-0,17	0,12	0,98
PBIAS - NPR reservoirs	0,93	0,51	0,94	0,66	0,49	0,26	0,09	0,43
PBIAS - all reservoirs	0,12	0,90	0,78	0,24	-1,24	0,88	0,12	0,28

389

390 4. Discussion

391 4.1. Limnological dynamics and trophic conditions

392 Limnological dynamics in the semiarid reservoirs are strongly influenced by seasonality and
 393 also by the prevailing hydro-climatic conditions [34,56]. Generally, it is expected that rainfall
 394 stimulates fluctuations in trophic conditions of water bodies due to an increase of sediment input
 395 from the drainage areas. This sediment can be rich in organic material and nutrients promoting the
 396 eutrophication process in the reservoirs.

397 Work comprising the trophic level of Marengo reservoir between the rainy season (April 2011
398 and March 2012) and the dry season (October 2012 and November 2013) reported the seasonality
399 related to tendency of water quality deterioration during dry season while the TSI estimated for wet
400 period showed a better condition [34]. However, our investigation was carried out at the beginning
401 of five consecutive years of drought in the State of Ceará [43], which explains the high mean TSI for
402 Marengo in the middle of 2014 worsening during 2015, indicating hypereutrophic conditions.
403 Overall, the reservoirs studied are characterized by high trophic levels between September 2015 and
404 January 2016, which can be explained by the prolonged drought. Hydrological drought is indeed an
405 important factor that controls water quality in aquatic ecosystems, mainly due to water level
406 reduction, increase in residence time and shifts in internal processes [40,56].

407 The temporal variation in the trophic status of the General Sampaio reservoir (located in the
408 center of the state of Ceará, Brazil) was also verified [32], showing lower values during the rainy
409 season and gradually increasing value throughout the dry period in 2012. Altogether, the
410 hydrological regime in semiarid Brazil acts on limnological dynamics in two ways: with the first
411 rains of the wet season a deterioration trend of water quality occurs due to the high input of organic
412 matter and nutrients carried from the drainage areas; in a second phase, the water tends to show an
413 improved quality resulting from the dilution of the material carried; the water quality can be
414 upgraded when there is an overflow event in the reservoirs. Usually there is a deterioration trend of
415 water quality during the dry period [34] and this condition is intensified in extreme hydrological
416 droughts [40].

417 The results showed intense trophic dynamics for SN and PB reservoirs in comparison to MAR
418 reservoir. This fact can be mainly explained by the aquatic ecosystem size, since smaller lakes have
419 more intense limnological dynamics [15]. In case of SN reservoir, the dense occurrence of
420 *Ceratophyllum demersum* can be directly linked to the low trophic status found from May 2014 to
421 February 2015. It is an exotic submerged species that exerts strong competition with nutrients, while
422 its distribution depends on transparency conditions of the water column [57,58].

423 Despite the fact that the drainage areas of MAR, PB and SN reservoirs are apparently subject to
424 similar practices and land uses (agricultural production with livestock accessing the water bodies,
425 precarious sewage infrastructure in the villages contributing to the supply of nutrients especially
426 through surface runoff), these reservoirs show distinctively different limnological conditions. It
427 shows that the limnology also depend on other factors such as residence time, size and storage
428 capacity of the water bodies, operating rules, morphology features and internal lake processes that
429 occur due to high water level fluctuations.

430 The different limnological conditions between MAR, PB and SN reservoirs can mainly be
431 linked to characteristics of intermittence, since Marengo is perennial, already built in 1934 (and
432 enlarged in 1956), while SN and PB were built more recently and are non-perennial, highly
433 disturbed limnological systems. There are records of emptying during 2013 in case of PB and from
434 October 2015 (until the end of this research) in case of SN. Furthermore, small local communities
435 take advantage of the water level decrease and use the (still) moist area of reservoir floor for
436 cultivating (floodplains, popularly called “vazantes” in the semiarid Brazil). Therefore, when these
437 reservoirs fill up again, a distinct condition related to aquatic metabolic processes arises and leads to
438 different physiochemical characteristics in waters. This factor may also affect the later pattern of
439 spatial and temporal distribution of macrophytes species.

440 4.2. Dynamics of CDOM

441 Considering that an optical characterization approach using CDOM is commonly performed in
442 temperate and boreal lakes [13,22,59], it is emphasized that there is – to our best knowledge - no
443 reference for small surface reservoirs of in semiarid Brazil that includes different limnic ecosystems,
444 different seasons and years. A study carried out in the hypertrophic Barra Bonita hydroelectric
445 reservoir located in São Paulo (subtropical Brazilian region), which has a storage capacity of 2.600
446 hm³ measured aCDOM(440) equal to 1.6 m⁻¹ in May 2014 (end of the wet season) and around 2.8 m⁻¹
447 in October 2014 (end of the dry season) [60]. In the same study area, Alcântara *et al.* [61] described

448 aCDOM(440) values between 0.6 m^{-1} (in January 2014) and 1.7 m^{-1} (in September 2014) for this
449 eutrophic system. Both works related the results to the seasonality with lower CDOM values for the
450 wet season and higher concentration in the dry season. In comparison, we observe a high dynamic
451 of CDOM that seems to be related to wet and dry seasonality and to the different sizes of the
452 reservoirs: the short water retention time in the reservoirs of Madalena basin and a close
453 hydrological connection to the drainage system, as well as high water level fluctuations influencing
454 the thermal stratification are also important factors to explain the high CDOM dynamic and also
455 dynamics of the trophic level found in this study.

456 Theoretically, rainfall has an impact on the CDOM concentration mainly due to material and
457 detritus transported from the drainage area to the water bodies. Taking into account that humic-rich
458 discharges follow the hysteresis process [19], the rainfall effects in the wet and dry seasons did not
459 seem so evident in our investigation throughout the entire study period. In the first wet season 2014,
460 rainfall effects were very well visible. CDOM regimes in the smaller SN and PB reservoirs were high
461 decreasing at the end of the wet season. The moderate wet season of 2015 seemed not to have a
462 considerable impact on the CDOM regimes in the smaller reservoirs because no changes in CDOM
463 related to precipitation could be observed. In contrast, the prolonged drought period may be a
464 stronger interfering factor on the temporally variable CDOM in reservoirs of Madalena basin.
465 CDOM in MAR and PB reservoirs considerably increased in December 2015 and January 2016 after
466 the end of the dry season. The availability of light in the water column regulates the photosynthetic
467 activity driving low phytoplankton concentrations in case of high CDOM as it occurs in the SN and
468 PB reservoirs during the wet season of 2014 to 2015.

469 Optical measurements of CDOM absorbance is increasingly used to track the composition and
470 the source of organic matter in aquatic ecosystems. No consensus exists on strictly formulated value
471 ranges for the interpretation of CDOM slope values across specific wavelength regions and the slope
472 ratio. Literature findings emphasize that typically lower slope values indicate high molecular
473 weight DOM and/or increasing aromaticity as it is typical for the humic-rich DOM type from
474 terrestrial sources [26,28]. Previous studies have reported S_{UV} values in the range of 0.02 to 0.03 nm^{-1}
475 for pelagic ocean waters dominated by phytoplankton [62], 0.014 - 0.018 nm^{-1} for wetlands [26] and a
476 wide range of 0.012 - 0.023 nm^{-1} for terrestrial aquatic systems [63]. The spectral slope ratio S_r is
477 shown to be negatively correlated to DOM molecular weight and to generally increase with
478 irradiation due to photodegradation [26,28]. Previous studies have reported S_r values with a wide
479 range of 0.76 - 1.79 in wetlands [26] and 0.7 - 2.4 in eutrophic shallow lake waters [25]. Experiments
480 carried out on CDOM photodegradation from algae leachates observed unusual initial S_{UV} values $>$
481 0.055 nm^{-1} and very low S_{VIS} values around 0.001 nm^{-1} resulting in S_r values around 50 that may not
482 represent a true exponential fit during the first days of the leaching experiment [28]. S_{UV} values and
483 S_r from algae leachates returned to commonly reported values within few days.

484 The CDOM type of the Marengo reservoir is consistently characterized by high S_{UV} values of
485 0.025 nm^{-1} , indicating a dominance of phytoplankton-derived CDOM. Phytoplankton degradation
486 in eutrophic systems is an important source of DOM with an autochthonous characterization which
487 has already been demonstrated for other eutrophic systems [25,64]. In contrast, the much lower S_{UV}
488 values from SN and PB reservoirs indicate a more humic-rich CDOM type. aCDOM(440) values
489 from SN and PB reservoirs were high in 2014 with decreasing values after the wet season 2014
490 accompanied by decreasing S_{UV} that may indicate the loss of aromaticity.

491 The S_{VIS} values showed a lower range than S_{UV} but followed in general the temporal dynamics
492 of S_{UV} . A close result was reached for lakes in Florida, with lower variability of S_{VIS} in comparison to
493 S_{UV} [10]. Work on optical conditions of two tropical freshwater systems (located in Minas Gerais,
494 Brazil) in 2013, both characterized by low CDOM but with different phytoplankton regimes
495 measured mean values of S_{UV} and S_{VIS} around 0.02 nm^{-1} and S_r equals to 1.2 for the eutrophic water
496 body, the Pampulha reservoir, which indicates CDOM derived from phytoplankton [64]. These
497 findings reinforce the suggestion that S and S_r values in the lower range characterize PB and SN
498 reservoirs likely much more influenced by allochthonous sources due to their much smaller water
499 volume. The CDOM source in the higher trophic Marengo reservoir tends more to *in situ* production

500 of CDOM from phytoplankton. Hypertrophic systems like MAR experience massive algal blooms
501 with several hundred $\mu\text{g}\cdot\text{L}^{-1}$ Chl-a. The occurrence of cyanobacterial harmful algal blooms
502 (CyanoHAB) may lead to Microcystis-DOM that is highly biolabile which implies high biological
503 alteration and turnover rates of DOM in the environment [27].

504 To understand the origin and the fate of DOM in the investigated very contrasting systems of
505 MAR, PB and SN reservoirs, further samplings and investigations are needed. It is recognized in the
506 literature that macrophytes and periphery act on the metabolism of DOM [22–25] and are also
507 introducing DOM in aquatic systems.

508 4.3. Bio-Optics and Remote Sensing application

509 The spectral responses of water can present a diverse behavior and magnitude of its
510 constituents, especially in complex waters. Brezonik *et al.* [10] undertook field spectrometry on a
511 large number of lakes in Minnesota (US) of different trophic levels and CDOM [10]. This work
512 collected relevant spectral groups in terms of CDOM regime and trophic level into: A) moderate
513 colour (from CDOM) in combination with highly eutrophic (from SD and Chl-a); B) high colour and
514 mesotrophic to eutrophic; C) high colour, oligotrophic (dystrophic system); D) low colour,
515 oligotrophic; and E) low colour, moderately eutrophic. Class B showed the highest variability of the
516 reflectance spectra within one group that could not further be separated related to suspended
517 particulate matter that was not an input option in this study. The authors describe for group B)
518 either nearly flat and lowest reflectance in case of CDOM being the dominant factor, or an overall
519 lower reflectance < 600 nm compared to the other groups in case of other water colour influencing
520 factors such as phytoplankton.

521 Researchers from Brazilian semiarid applied field spectroscopy in their remote sensing study of
522 the Orós reservoir (State of Ceará, Brazil) and they verified that this is an eutrophic reservoir with a
523 reflectance peak in the green and a second (lower) one in the NIR and the pigment absorption bands
524 in the blue and red wavelength region [12] such as it is described for group A) in [10]. Reflectance
525 spectra from Marengo also show a distinct reflectance peak in the green region (~570 nm), an
526 absorption feature in the blue and the red (~675 nm) and another reflectance peak in the NIR domain
527 around 710 nm. This is a similar spectral shape for Marengo in case of all measurements, being a
528 behaviour associated to high concentration of photosynthetic pigments. The reflectance spectral
529 type for Marengo can also be classified as group A) proposed by [10] as moderate CDOM in
530 combination with highly eutrophic.

531 The presence of CDOM can be observed from high absorbance of light in the shorter
532 wavelength regions with exponential reduction of values towards the longer wavelengths [11,61].
533 Because of lowest phytoplankton concentration in PB and SN reservoirs throughout most of the
534 observation period we assign PB and SN to the spectral group C) high colour, oligotrophic,
535 dystrophic system according to [10]. In case of high CDOM and low concentration of particles for
536 scattering the reflectance spectra show overall a flat reflectance. In case of high phytoplankton
537 concentration, phytoplankton cells and colonies highly scatter rising the reflectance across the
538 visible spectrum. PB shows higher phytoplankton development and changes to the spectral group
539 B) high colour and mesotrophic to eutrophic.

540 The satellite data analysis showed the limitations one has to deal with when using satellite data
541 for such complex and small inland water bodies of varying character. Differentiation of absorption
542 from phytoplankton and non-algal organic particles *versus* CDOM in surface waters is complex and
543 challenging, resulting from the variation of the shape and magnitude of the water surface
544 reflectance [10,11]. The Chl-a model that best performed used the NIR reflectance from the
545 RapidEye dataset (Fig. 6B). This was clearly perceived for Marengo reservoir, due to the high Chl-a
546 concentration. Landsat-8 imagery was applied for mapping Chl-a concentration in a hypertrophic
547 tropical lake (Barra Bonita hydroelectric reservoir - BBHR, São Paulo State, Brazil) with satisfactory
548 results [60]. This study found the best fit by applying a NIR-Red ratio with $R^2 > 0.7$ for
549 hypereutrophic data. However, no index from L8 showed reasonable results for our data. The best
550 fit obtained for CDOM simulation was based on I5 (green reflectance derived from L8 data), as

551 showed in Figure 7C for the non-perennials reservoirs (PB and SN). This result can be explained by
552 the fact that CDOM spectral influence is most evident in lakes with low abundance of
553 phytoplankton corroborating the assumptions of Brezonik *et al.* [6,10].

554 Similarly to our investigation, Alcântara *et al.* [61] tested the potential of red/blue-green ratio
555 for predicting aCDOM(440) from Landsat-8 images. They also worked in the same hypertrophic
556 reservoir studied by [60] and the best fit was yielded through the red-blue ratio band, derived by *in*
557 *situ* hyperspectral remote sensing reflectance, at 650 nm (L8 band 4) and 480 nm (L8 band 2), i.e.
558 [Rrs(650) / Rrs(480)]. This performance was validated by good coefficient of determination ($R^2 = 0.7$).
559 However, in our investigation, the red-blue ratio band did not show a good correlation against *in*
560 *situ* aCDOM(440). Other studies also show how satellite-derived CDOM estimation deteriorates if
561 lakes are eutrophic with high phytoplankton [10,65].

562 With regard to performance of Chl-a and CDOM algorithms, it was not possible to collect *in situ*
563 data in all reservoirs at the exact same day as the satellite overpasses. Nevertheless, in the same way
564 as discussed by Hansen *et al.* [66], the time interval between samplings and image acquisition does
565 not seem to be the main factor interfering on the results. For the small reservoirs SN and PB we
566 experienced difficulties in extracting 'pure water pixels' at or close to the sampling sites that are not
567 influenced by macrophytes. In case of PB we estimate that there will be always a background signal
568 of floating and emergent macrophytes specifically in the coarser spatial resolution Landsat pixels.
569 The time shift between image acquisition and *in situ* data collection may be a problem in case of
570 heavy precipitation events flushing into reservoirs that could impact thermal stratification and by
571 this the phytoplankton biomass and dissolved organic matter in the reservoirs, which was not
572 perceived during our study, however.

573 Notwithstanding, future research is required to monitor these highly dynamic systems
574 throughout the dry and wet seasons and throughout the years to better understand the
575 spatiotemporal pattern of Chl-a and CDOM and also to improve remote sensing indices and
576 understand which indices can be transferred to other water bodies for upscaling for larger area
577 mapping. It could be advisable to undertake a pre-classification for spectral types such as presented
578 in [10] and subsequently apply optimized indices for these different trophic levels and CDOM
579 regimes. It is emphasized that also global CDOM algorithms may be applied if the spectral
580 responses of the lake systems are better understood [10,14]. For mapping of Chl-a, a
581 pre-classification distinguishing lower trophic from higher trophic systems may improve the
582 prediction accuracies due to the distinctly different spectral signatures of the water bodies resulting
583 from the phytoplankton with high phytoplankton leading to high green and NIR reflectance in
584 contrast to oligotrophic systems. Additionally, a pre-classification of emergent and floating
585 macrophytes is needed excluded non-pure water pixels.

586 5. Conclusions

587 This study aimed at analyzing the eutrophication dynamics in three small water surface
588 reservoirs in semiarid Brazil. The study is based on an extensive *in situ* sampling for Chlorophyll-a
589 and CDOM data including other attributes to characterize the trophic state of the reservoirs for a
590 period of 20 months from May 2014 until January 2016. The studied reservoirs include the perennial
591 highly dendritic reservoir Marengo (MAR, 15 hm³), and the smaller non-perennial reservoirs Paus
592 Branco (PB, 5 hm³) and São Nicolau (SN, 0.9 hm³) in the state of Ceará, Brazil. The large MAR is
593 eutrophic, whereas the smaller reservoirs experienced extreme water level fluctuations (from
594 completely full to completely empty), extensive occurrence of macrophytes and low phytoplankton
595 concentrations. The investigated reservoirs showed higher eutrophication level towards the end of
596 the monitoring period, when compared with the initial situation, because the monitoring period fell
597 into a multi-year hydrological drought (2012 - 2016), which contributed to worsen the water quality.

598 The absorption and spectral slopes of CDOM were used for the characterization of DOM in the
599 reservoirs. CDOM Slopes (S) and Slope ratio (Sr) suggest that the perennial eutrophic MAR
600 reservoir is mostly influenced by autochthonous CDOM that fits to the dominance of phytoplankton
601 with moderate CDOM in this system. Contrarily, the small and highly disturbed limnic ones (PB

602 and SN) presented extreme water level fluctuations and occasional complete drying-up. CDOM S
603 and Sr values in the PB and SN reservoirs indicate a dominance of allochthonous-derived CDOM.
604 This is the first approach using absorption and slope spectral of CDOM for Brazilian semiarid
605 reservoirs providing additional information for the traditional parameters of water quality.

606 In addition, *in situ* spectral reflectance measurements were performed and the reservoirs
607 classified as spectral group type, according to Brezonik *et al.* [10]: the perennial Marengo reservoir
608 could be assigned to the highly variable spectral group A (moderate CDOM in combination with
609 highly eutrophic); whereas the non-perennial reservoirs could be assigned to the spectral group C
610 (high CDOM, oligotrophic-dystrophic system). The spectral group classification of PB shifted
611 towards the end of the study period to B (high CDOM and mesotrophic to eutrophic status).

612 This research shows the challenges for optical remote sensing applications for different spectral
613 types with widely variable trophic state. The Chl-a model that performed best used the NIR band
614 from RapidEye and was successful only for the eutrophic MAR reservoir, due to the high Chl-a
615 concentration. The best fit for CDOM retrievals used the green band from Landsat-8 imagery
616 performing best for non-perennial reservoirs with high CDOM without optical interference by
617 phytoplankton as in the perennial reservoir. Further collection of *in situ* data throughout the dry and
618 wet season is required to improve the indices and to apply them to a larger number of water bodies.

619 The land cover and land use changes in the watershed also contributed to an increase in the
620 eutrophication of the reservoirs, especially during the rainy season. Therefore, we strongly
621 recommend the land use classification for linking it to trophic conditions of the studied reservoir.
622 Considering the water scarcity of the Brazilian semiarid region, our investigation encourages an
623 important path for water resources management, showing a meaningful potential for retrieving the
624 quality of water bodies in areas where small and medium sized reservoirs are the main source of
625 water supply for the dense rural population. Due to the large number of reservoirs in the semiarid,
626 the governmental and State agencies prioritize on monitoring only large strategic reservoirs even
627 though not on a regular basis, whereas there is a complete absence for monitoring water quality of
628 small semiarid reservoirs.

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636 References

- 637
638 1. Peter, S.; de Araújo, J. C.; Araújo, N.; Herrmann, H.J. Flood avalanches in a semiarid basin with a dense
639 reservoir network. *J. Hydrol.* **2014**, *512*, 408-420, doi: 10.1016/j.jhydrol.2014.03.001.
- 640 2. De Araújo, J.C. Gestão das águas de pequenos açudes na região semiárida. In *Recursos hídricos em regiões*
641 *áridas e semiáridas*, 1st. ed.; Medeiros S.S.; Gheyi, H.S.; Galvão, C.O.; Paz, V.P.S. INSA: Campina Grande,
642 Brazil, 2011; pp. 334-351.
- 643 3. Heine, I.; Francke, T.; Rogass, C.; Medeiros, P. H. A.; Bronstert, A.; Foerster, S. Monitoring seasonal
644 changes in the water surface areas of reservoirs using TerraSAR-X time series data in semiarid
645 Northeastern Brazil. *IEEE* **2014**, *7*, 3190-3199, doi: 10.1109/JSTARS.2014.2323819.
- 646 4. Toledo, C.E., De Araújo, J.C., Almeida, C.L. The use of remote-sensing techniques to monitor dense
647 reservoir networks in the Brazilian semiarid region. *Intern. J. Remote Sens.* **2014**, *35*, 3683-3699, doi:
648 10.1080/01431161.2014.915593.
- 649 5. Zhang, S.; Foerster, S.; Medeiros, P., de Araújo, J. C., Motagh, M., Waske, B. Bathymetric survey of water
650 reservoirs in north-eastern Brazil based on TanDEM-X satellite data. *Science of the Total Environ.* **2016**, *571*,
651 575-593, doi: 10.1016/j.scitotenv.2016.07.024.

- 652 6. Brezonik, P., Menken, K.D.; Bauer, M. Landsat-based remote sensing of lake water quality characteristics,
653 including chlorophyll and colored dissolved organic matter (CDOM). *Lake and Reservoir Management*
654 **2009**, 21, 373–382, doi: 10.1080/07438140509354442
- 655 7. Dall’Olmo G.; Gitelson, A. A. Effect of bio-optical parameter variability on the remote estimation of
656 chlorophyll-a concentration in turbid productive waters: experimental results. *Applied Optics* **2005**, 44,
657 412-422.
- 658 8. Kutser, T.; Pierson, D. C.; Kallio, K.Y.; Reinart, A.; Sobek, S. Mapping lake CDOM by satellite remote
659 sensing. *Remote Sens. Environ.* **2005**, 94, 535–540.
- 660 9. Zhu, W., Yu, Q., Tian, Y.Q., Becker, B.L., Zheng, T., Carrick, H. J. An assessment of remote sensing
661 algorithms for colored dissolved organic matter in complex freshwater environments. *Remote Sens.*
662 *Environ.* **2014** 140, 766–778, doi: 10.1016/j.rse.2013.10.015.
- 663 10. Brezonik, P.L. Olmanson, L.G.; Finlay, J.C.; Bauer, M.E. Factors affecting the measurement of CDOM by
664 remote sensing of optically complex inland waters. *Remote Sens. Environ.* **2015**, 157, 199–215, doi:
665 10.1016/j.rse.2014.04.033.
- 666 11. Shao, T.; Song, K.; Du, J.; Zhao, Y.; Lui, Z.; Zhang, B. Retrieval of CDOM and DOC using *in situ*
667 hyperspectral data: A case study for potable water in Northeast China. *J. Indian Society Remote Sens.*
668 **2015**, 44, 77-89, doi: 10.1007/s12524-015-0464-2.
- 669 12. Lopes, F. B.; Novo, E. M. L. M.; Barbosa, C. C. F.; Andrade, E. M.; Ferreira, R. D. Simulation of spectral
670 bands of the MERIS sensor to estimate chlorophyll-a concentrations in a reservoir of the semi-arid region.
671 *Rev. Agro@ambiente* **2016**, 10, 96-106, doi: 10.18227/1982-8470ragro.v10i2.3482.
- 672 13. Olmanson, L.G.; Brezonik, P.L.; Finlay, J.C.; Bauer, M.E. Comparison of Landsat 8 and Landsat 7 for
673 regional measurements of CDOM and water clarity in lakes. *Remote Sens. Environ.* **2016**, 185, 119–128, doi:
674 10.1016/j.rse.2016.01.007.
- 675 14. Dash, P.; Silwal, S.; Ikenga, J. O.; Pinckney, J. L.; Arslan, Z.; Lizotte, R. E. Water quality of four major lakes
676 in Mississippi, USA: Impacts on human and aquatic ecosystem health. *Water*. **2015**, 7, 4999-5030, doi:
677 10.3390/w7094999.
- 678 15. Wetzel, R. G. Limnology - lake and river ecosystems. 3rd ed.; Academic Press, San Diego, USA, 2001;
679 1006p, ISBN 978-0-12-744760-5.
- 680 16. Gholizadeh, M. H.; Melesse, A. M.; Reddi, L. A Comprehensive review on water quality parameters
681 estimation using remote sensing techniques. *Sensors*. **2016**, 1298, 2-43, doi: 10.3390/s16081298.
- 682 17. Dekker, A.; Peters, S. The use of the Thematic Mapper for the analysis of eutrophic lakes: a case study in
683 the Netherlands. *Int. J. Remote Sens.* **1993**, 14, 799-821, doi: 10.1080/01431169308904379.
- 684 18. Tranvik, L.J. Allochthonous dissolved organic matter as an energy source for pelagic bacteria and the
685 concept of the microbial loop. *Hydrobiol.* **1992**, 229, 107-114.
- 686 19. Steinberg, E. W. Regulatory impacts of humic substances in lakes. In *The Lakes Handbook – Limnology and*
687 *Limnetic Ecology*, 1st ed.; O’Sullivan, P. E.; Reynolds, C. S. (Eds.); Blackwell Science Ltd: Malden, USA, 2004;
688 pp. 153-196, ISBN: 0632047976.
- 689 20. Steinberg, C.E.W.; Meinelt, T.; Timofeyev, M.A.; Bittner, M.; Menzel, R. Humic Substances (review series).
690 Part 2: Interactions with Organisms. *Env Sci Pollut Res.* **2008**, 15, 128-135, doi: 10.1065/espr2007.07.434.
- 691 21. Loiselle, S. A.; Bracchini, L.; Dattilo, A. M.; Ricci, M.; Tognazzi, A.; Cózar, A.; Rossi, C. Optical
692 characterization of chromophoric dissolved organic matter using wavelength distribution of absorption
693 spectral slopes. *Limnol.Oceanogr.* **2009**, 54, 590-597.
- 694 22. Wetzel, R.G. Gradient-dominated ecosystems: source and regulatory functions of dissolved organic
695 matter in freshwater ecosystems. *Hydrobiol.* **1992**, 229, 181-198.
- 696 23. Pflugmacher, S.; Spangenberg, M.; Steinberg, C. E. W. Dissolved Organic Matter (DOM) and the effects on
697 the aquatic macrophyte *Ceratophyllum demersum* in relation to photosynthesis, pigment pattern and
698 activity of detoxication enzymes. *J. Applied Botany*, **1999**, 73, 184-190.
- 699 24. Farjalla, V.F.; Amado, A.M.; Suhett, A. L.; Meirelles-Pereira, F. DOC removal paradigms in highly humic
700 aquatic ecosystems. *Env Sci Pollut Res.* **2009**, 16, 531–53, doi: 10.1007/s11356-009-0165-x.
- 701 25. Zhang, Y.; van Dijk, M.A.; Lui, M.; Zhu, G.; Qin, B. The contribution of phytoplankton degradation to
702 chromophoric dissolved organic matter (CDOM) in eutrophic shallow lakes: Field and experimental
703 evidence. *Water Research*, **2009**, 43, 4685-4697, doi: 10.1016/j.watres.2009.07.024.

- 704 26. Helms, J. R.; Stubbins, A.; Ritchie, J.D.; Minor, E.C.; Kieber, D.J.; Mopper, K. Absorption spectral slopes
705 and slope ratios as indicators of molecular weight, source and photobleaching of chromophoric dissolved
706 organic matter. *Limnology and Ocean*. **2008**, *53*, 955–969, doi: 10.4319/lo.2008.53.3.0955.
- 707 27. Bittar, T.; Vieira, A.A.H.; Stubbins, A.; Mopper, K. Competition between photochemical and biological
708 degradation of dissolved organic matter from the cyanobacteria *Microcystis aeruginosa*. *Limnology and*
709 *Ocean*. **2015**, *60*, 1172–1194, doi: 10.1002/lno.10090.
- 710 28. Hansen, A.M.; Kraus, T.E.C.; Pellerin, B.A.; Fleck, J.A.; Downing, B.D.; Bergamaschi, B.A. Optical
711 properties of dissolved organic matter (DOM): Effects of biological and photolytic degradation. *Limnology*
712 *and Ocean*. **2016**, *61*, 1015–1032, doi: 10.1002/lno.10270.
- 713 29. Kutser, T.; Verpoorter, C.; Paavel, B.; Tranvik, L.J. Estimating lake carbon fractions from remote sensing
714 data. *Remote Sens. Environ.* **2015**, *157*, 138–146, doi: 10.1016/j.rse.2014.05.020.
- 715 30. Shi, L.; Mao, Z.; Wu, J.; Liu, M.; Zhang, Y.; Wang, Z. Variations in spectral absorption properties of
716 phytoplankton, non-algal particles and chromophoric dissolved organic matter in Lake Qiandaohu. *Water*
717 **2017**, *9*, 352–372; doi: 10.3390/w9050352.
- 718 31. Palmer, S. C. J., Kutser, T., Hunter, P. D. Remote sensing of inland waters: challenges, progress and future
719 directions. *Remote Sens. Environ.* **2015**, *157*, 1–8, doi: 10.1016/j.rse.2014.09.021.
- 720 32. Chaves, F.I.B.; Lima, P.F.; Leitão, R.C.; Paulino, W.D. Santaella, S.T. Influence of rainfall on the trophic
721 status of a Brazilian semiarid reservoir. *Acta Scient.* **2013**, *35*, 505–511, doi:
722 10.4025/actasciobiolsci.v35i4.18261.
- 723 33. De Freitas Lima, P.; Sousa, M.S.R.; Porfírio, A.F.; Almeida, B.S.; Freire, R.H.F.; Santaella, S.T. Preliminary
724 analysis on the use of Trophic State Indexes in a Brazilian semiarid reservoir. *Acta Scient.* **2015**, *37*, 309–318,
725 doi: /10.4025/actasciobiolsci.v37i3.27160.
- 726 34. Wiegand, M.C.; Piedra, J. I. G.; De Araújo, J.C. Vulnerability towards eutrophication of two tropical lakes
727 in both humid (Cuba) and semiarid (Brazil) climates. *Eng. Sanit. Ambient.* **2016**, *21*, 415–424, doi:
728 10.1590/s1413-41522016139527.
- 729 35. Figueirêdo, M.C.B.; Teixeira, A.S.; Araújo, L.F.P.; Rosa, M.F.; Paulino, W.D.; Mota, S.; de Araújo, J.C.
730 Avaliação da vulnerabilidade ambiental de reservatórios à eutrofização. *Eng. Sanit. Ambient.* **2007**, *12*,
731 399–409, doi: 10.1590/S1413-41522007000400006.
- 732 36. Santos, J.C.N.; Andrade, E.M.; Neto, J.R.A.; Meireles, A.C.M.; Palácio, H.A.Q. Land use and trophic state
733 dynamics in a tropical semiarid reservoir. *Rev. Ciênc. Agron.* **2014**, *45*, 35–44, doi:
734 10.1590/S1806-66902014000100005.
- 735 37. Carlson, R.E. A trophic state index for lakes. *Limnology and Ocean*. **1977**, *22*, 361–380, doi:
736 10.4319/lo.1977.22.2.0361.
- 737 38. Toledo Jr., A.P. Informe preliminar sobre os estudos para obtenção de um índice para avaliação
738 simplificada do estado trófico de reservatórios de regiões quentes tropicais. CETESB- Companhia
739 Ambiental do Estado de São Paulo, Brazil (Report in Portuguese). **1990**.
- 740 39. Salas, H.J.; Martino, P. A simplified phosphorus trophic state model for warm- water tropical lakes. *Water*
741 *Research* **1991**, *25*, 341–350, doi: 10.1016/0043-1354(91)90015-I.
- 742 40. Li, S.; Bush, R.T.; Mao, R.; Xiong, L.; Ye, C. Extreme drought causes distinct water acidification and
743 eutrophication in the Lower Lakes (Lakes Alexandrina and Albert), Australia. *J. Hydrol.* **2017**, *544*, 133–
744 146, doi: 10.1016/j.jhydrol.2016.11.015.
- 745 41. Münster, U. Studies on phosphatase activities in humic lakes. *Environ. International* **1994**, *20*, 49–59, doi:
746 10.1016/0160-4120(94)90066-3.
- 747 42. Carpenter, S.R.; Pace, M.L. Dystrophy and eutrophy in lake ecosystems: Implications of fluctuating
748 inputs. *Oikos* **1997**, *78*, 3–14, doi: 10.2307/3545794.
- 749 43. De Araújo, J.C.; Bronstert, A. A method to assess hydrological drought in semiarid environments and its
750 application to the Jaguaribe River basin, Brazil. *Water Intern.* **2015**, *41*, 213–230, doi:
751 10.1080/02508060.2015.1113077.
- 752 44. Milton, E.J. Principles of field spectroscopy. *I. J. Remote Sensing*, **1987**, *8*, 1807–1827.
- 753 45. APHA - American Public Health Association. *Standard Methods for the Examination of Water and Wastewater*.
754 21th ed. Washington DC, 2005.
- 755 46. Bricaud, A.; Morel, A.; Prieur, L. Absorption by dissolved organic matter of the sea (yellow substance) in
756 the UV and visible domains. *Limnology and Ocean*. **1981**, *26*, 43–53.

- 757 47. FUNCEME. Fundação Cearense de Meteorologia e Recursos Hídricos. Dados de Pluviometria do Estado
758 do Ceará. Available on line: <http://www.funceme.br/app/calendario/produto/municipios/media/mensal>
759 (accessed on 7 February 2017).
- 760 48. Roy, D.P.; Wulder, M.A. Loveland, T.R., Woodcock, C.E.; Allen, R.G.; Anderson, M.C.; Helder, D.; Irons, J.
761 R.; Johnson, D.M.; Kennedy, R. Landsat-8: Science and product vision for terrestrial global change
762 research. *Remote Sens. Environ.* **2014**, *145*, 154–172, doi: 10.1016/j.rse.2014.02.001.
- 763 49. RapidEye—Satellite Imagery Product Specifications, Version 6.1, April 2015. Available online:
764 http://blackbridge.com/rapideye/upload/RE_Product_Specifications_ENG.pdf (accessed on 18 August
765 2015).
- 766 50. Weichelt, H.; Rosso, P.; Marx, A. Reigber, S.; Douglas, K.; Heynen, M. The RapidEye Red Edge Band.
767 Available on line: https://resa.blackbridge.com/files/2014-06/Red_Edge_White_Paper.pdf (accessed on 30
768 August 2015).
- 769 51. USGS. Provisional Landsat 8 Surface Reflectance Data, U.S. Geological Survey: Sioux Falls, SD, USA, 2015.
770 Available on line: http://landsat.usgs.gov/about_LU_Special_Issue_3.php (accessed on 1st August 2016).
- 771 52. Richter, R.; Schläpfer, D. Atmospheric / Topographic correction for Satellite Imagery. ATCOR 2/3 User
772 guide. Available on line: http://www.rese.ch/pdf/atcor3_manual.pdf (accessed on 9 January 2016).
- 773 53. Nash, J.E.; Sutcliffe, J.V. River flow forecasting through conceptual models part I – A discussion of
774 principles, *J. Hydrol.*, **1970**, *10*, 282–290, doi: 10.1016/0022-1694(70)90255-6.
- 775 54. Moriasi, D.N.; Arnold, J.G.; Van Liew, M.W.; Bingner, R.L.; Harmel, R.D.; Veith, T.L. Model evaluation
776 guidelines for systematic quantification of accuracy in watershed simulations, *American Society of Agric.
777 and Biol. Engineers*, **2007**, *50*, 885-900.
- 778 55. CONAMA - Conselho Nacional do Meio Ambiente. Available on line:
779 <http://www.mma.gov.br/port/conama/res/res05/res35705.pdf> (accessed on 06 June 2014).
- 780 56. Braga, G.G.; Becker, V.; Oliveira, J.N.P.; Mendonça Júnior, J.R.; Bezerra, A.F.M.; Torres, L.M.; Galvão,
781 A.M.F.; Mattos, A. Influence of extended drought on water quality in tropical reservoirs in a semiarid
782 region. *Acta Limnol. Brasiliensia* **2015**, *27*, 15-23, doi: 10.1590/S2179-975X2214.
- 783 57. Bornette, G.; Puijalóné, S. Response of aquatic plants to abiotic factors: a review. *AquatSci*, **2011**, *73*, 1–14,
784 doi: 10.1007/s00027-010-0162-7.
- 785 58. Pelechaty, M.; Pronin, E.; Pukacz, A. Charophyte occurrence in *Ceratophyllum demersum* stands.
786 *Hydrobiol.* **2014**, *737*, 111–120, doi: 10.1007/s10750-013-1622-6.
- 787 59. Reche, I., Pace, M.L., Cole, J.J. Relationship of trophic and chemical conditions to photobleaching of
788 dissolved organic matter in lake ecosystems. *Biogeochemistry* **1999**, *44*, 258-289, doi: 10.1023/A:100609032.
- 789 60. Watanabe, F.S.Y.; Alcântara, E.; Rodrigues, T.W.P.; Imai, N.N.; Barbosa, C.C.F.; Rotta, L.H.S. Estimation of
790 Chlorophyll-a concentration and the trophic state of the Barra Bonita Hydroelectric Reservoir using
791 OLI/Landsat-8 images. *Int. J. Environ. Research Public Health* **2015**, *12*, 10391-10417,
792 doi:10.3390/ijerph120910391.
- 793 61. Alcântara, E., Bernardo, N., Watanabe, F., Rodrigues, T., Rotta, L., Carmo, A., Shimabukuro, M.,
794 Gonçalves, S, Imai, N. Estimating the CDOM absorption coefficient in tropical inland waters using
795 OLI/Landsat-8 images. *Remote Sens. Letters* **2016**, *7*, 661-670, doi: 10.1080/2150704X.2016.1177242.
- 796 62. Del Vecchio, R.; Blough, N. Photobleaching of chromophoric dissolved organic matter in natural waters:
797 Kinetics and modeling. *Marine Chemistry*, **2002**, *78*, 231-253, doi: 10.1016/S0304-4203(02)00036-1.
- 798 63. Spencer, R.G.M.; Butler, K.D.; Aiken, G.R. Dissolved organic carbon and chromophoric dissolved organic
799 matter properties of rivers in the USA, *J. Geophysical Research* **2012**, *117*, 1-14, doi:10.1029/2011JG001928.
- 800 64. Brandão, L.P.M.; Staehr, P.A.; Bezerra-Neto, J.F. Seasonal changes in the optical properties of two
801 contrasting tropical freshwater systems. *J. Limnology* **2016**, *75*, 508-51, doi: 10.4081/jlimnol.2016.1359.
- 802 65. Menken, K.D., Brezonik, P.L., Bauer, M.E. Influence of chlorophyll and colored dissolved organic matter
803 (CDOM) on lake reflectance spectra: implications for measuring lake properties by remote sensing. *Lake
804 Reserv. Manage* **2006**, *22*, 179–190, [http:// dx.doi.org/10.1080/07438140609353895](http://dx.doi.org/10.1080/07438140609353895).
- 805 66. Hansen, C.H., Burian, S.J., Dennison, P.E., Williams, G.P. Spatiotemporal variability of lake water quality
806 in the context of remote Sensing models. *Remote Sens.* **2017**, *9*, 409; doi:10.3390/rs9050409.