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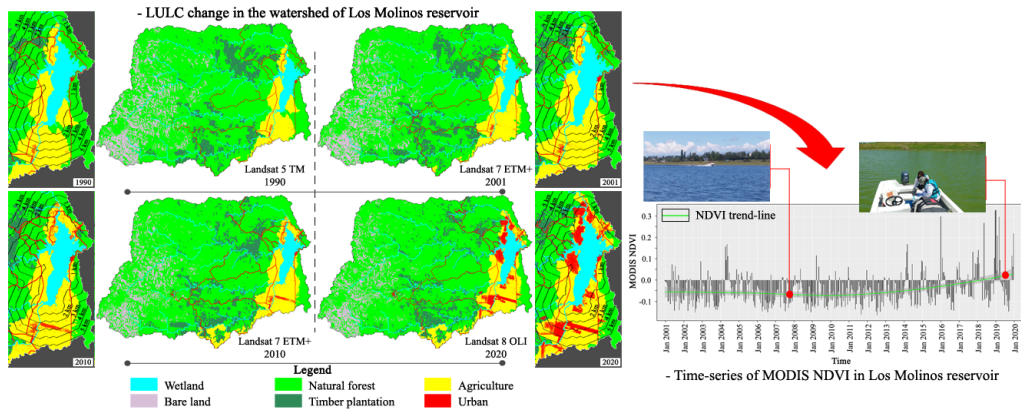
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1 **Assessing land use and land cover change in Los Molinos reservoir watershed and the**
2 **effect on the reservoir water quality**

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20 **ABSTRACT**

21 Understanding and modelling land use and land cover (LULC) change have become one of
22 the major subjects of interest for environmental management due to the negative effects that
23 human activities generate on the normal functioning and dynamics of freshwater resources.
24 Remote sensing and geographic information systems (GIS) are essential tools for assessing
25 the drivers that cause LULC change and its relationship with lake and reservoir water quality.

26 The objective of this study was to assess the spatial and temporal dynamics of LULC change
27 in the watershed of Los Molinos reservoir (Argentina), and to investigate its relationship with
28 the reservoir's water quality. Four Landsat imagery was used to analyse the LULC change in
29 the studied watershed and in different buffer zones from 1990 to 2020. Further, the
30 Normalized Difference Vegetation Index (NDVI) derived from a MODIS time-series dataset
31 (2001-2020) was used to explain the effects of LULC change on the status of the reservoir.
32 Results showed that the most significant LULC change started two decades ago and it has
33 intensified during the last ten years. This change is related to the intensification of agriculture
34 activities, and to the increasing conversion into urban areas, mainly on the shores of Los
35 Molinos reservoir. During the period 2010-2020, urbanization located in the 1 km buffer zone
36 defined from the shore of the reservoir increased at an annual rate of 18.02 %. The
37 degradation trend of LULC in Los Molinos watershed significantly contributed to the
38 degradation of water quality of the reservoir. This was corroborated by analysing the MODIS
39 NDVI time-series, which showed that since 2014 the NDVI trend-line presented an increasing
40 behaviour and extreme values of NDVI, related to algal blooms, were more frequently
41 observed.

42

43 **KEYWORDS:** Landsat satellites; LULC change; MODIS NDVI time-series; Reservoir;
44 Watershed.

45

46 1. INTRODUCTION

47 Lakes and reservoirs provide a variety of services to human society,
48 including drinking water, energy, fisheries, navigation, flood storage and recreation.
49 However, these ecosystems are vulnerable to water pollution, such as eutrophication, which is

50 one of the most urgent environmental problems in water resource management all over the
51 world (Smith and Schindler, 2009; Padedda et al., 2017; Bhagowati and Ahamad, 2019).
52 Eutrophication, which is primarily attributed to the continuous increase in the input of
53 essential nutrients into the aquatic systems, such as phosphorus and nitrogen (Guan et al.,
54 2020), results in adverse impacts on ecological functions provided by water resources (Amé et
55 al., 2003; Mendoza et al., 2015; Cai et al., 2020). Although eutrophication is a slow natural
56 process, it can be sharply accelerated by human activities posing potential risks to human
57 health and sustainable development (Ferral et al., 2018; Le Moal et al., 2019; Lin et al., 2020).
58 Different studies suggest that intensive land use practices, such as urbanization, agriculture,
59 mining, afforestation and deforestation, have greatly altered the normal functioning and
60 dynamic of freshwater resources (Ngoye and Machiwa, 2004; Seeboonruang, 2012; Bonansea
61 et al., 2016; Padedda et al., 2017; Le Moal et al., 2019; Rojas et al., 2019; Desta and Fetene,
62 2020; Guan et al., 2020). Therefore, understanding and modelling of land use and land cover
63 (LULC) change at a watershed scale has become one of the major subjects of interest for
64 environmental management and land use planning since it can be used as an indicator of the
65 impact of human activities over aquatic systems (Yu et al., 2011; Xu et al., 2020).

66 Remote sensing and geographic information systems (GIS) are essential tools for
67 assessing the drivers that cause LULC change and its relationship with lake and reservoir
68 water quality (Fukushima et al., 2007; Huang et al., 2014; Santos et al., 2017; Wei et al.,
69 2020). These techniques facilitate monitoring and analysis of LULC change and
70 eutrophication more efficiently than ground-based observations due to its time- and cost-
71 effectiveness over large areas as well as remote locations. In this sense, the Landsat Program,
72 which consists of a series of satellites missions, has the ability to monitor and quantitatively
73 characterize LULC change at a scale where natural and human-induced causes can be

74 detected (Loveland and Irons, 2016; Gómez et al., 2016). This can be achieved due to its long
75 history and relatively high spatial resolution (Loveland and Irons, 2016).

76 On the other hand, different authors suggest that spectral indices (e.g., NDVI, FAI,
77 NDAVI, WAVI, NDTI, MCI, NDPI, CCI, among others) are highly recommended for
78 mapping and monitoring water quality parameters in aquatic environments (Tebbs et al.,
79 2013; Villa et al., 2014; Zhou et al., 2018; Elhag et al., 2019; Kiage and Douglas, 2020).
80 These indices are generally calculated from the combination of visible and infrared bands of
81 remote sensing sensors (Qing et al., 2020). Among several satellites that have been
82 successfully used for evaluating water quality from spectral indices, the MODIS (Moderate
83 Resolution Imaging Spectroradiometer) sensor onboard NASA's Terra/Aqua satellites has
84 proven useful for deriving assessment of inland and coastal waters due to its shorter revisit
85 time and its ability to provide long-term and dense time-series data (Villar et al. 2013, Chen et
86 al., 2014; Germán et al., 2020; Li et al., 2020).

87 Advances in remote sensing technology have considerably improved the assessment of
88 a long time-series of LULC change (Huang et al., 2002; Fukushima et al., 2007; Huang et al.,
89 2014; Loveland and Irons, 2016; Gómez et al., 2016; Wei et al., 2020), variations in water
90 quality (Tebbs et al., 2013; Chen et al., 2014; Villa et al., 2014; González-Márquez et al.,
91 2018; Zhou et al., 2018; Bonansea et al., 2019; Elhag et al., 2019; Germán et al., 2020; Kiage
92 and Douglas, 2020; Li et al., 2020, Torregroza-Espinosa, et al. 2020) and eutrophication
93 causes (Smith and Schindler, 2009; Mendoza et al., 2015; Padedda et al., 2017; Bhagowati
94 and Ahamad, 2019; Le Moal et al., 2019; Cai et al., 2020). However, to the best of our
95 knowledge, studies that relate LULC change and reservoir water quality are scarce and
96 dispersed, particularly in the Southern hemisphere. Therefore, the main objective of this study
97 was to characterize and assess the spatial and temporal dynamics of LULC change in the

98 watershed of Los Molinos reservoir (Argentina), and to investigate its relationship with the
99 water quality of the reservoir.

100 Los Molinos reservoir, located in the central region of Argentina, was built in 1953 for
101 multi-purposes, including water supply for towns located on its shores and for the 30% of
102 Córdoba city (1.4 million inhabitants), power generation, flood control, irrigation, habitat for
103 different species of animals and plants, tourism and recreational activities (Bazán et al., 2005).
104 The watershed of this reservoir is dominated by natural forest, characterized by dry woodland
105 and hard grasses (Cabrera, 1976). In the late 1970s, as a result of a tax deferral plan
106 implemented by the government, afforested areas for sawn timber production were
107 established in the watershed with exotic pines, mainly *Pinus elliottii*, followed by *P. taeda*, *P.*
108 *radiata* and *P. patula* (Dorado et al., 1997; Cibils-Martina et al., 2017). In the last decades,
109 climate change and high-intensity human activity pressures related to agricultural activities
110 and urban expansion have caused the degradation of the watershed. As a consequence, Los
111 Molinos reservoir has experienced an advance of the trophic state, a degradation of water
112 quality, and multiple algal blooms (Bazán et al., 2005; Bonansea et al., 2018). In the present
113 study, by using Landsat imagery we have focused on assessing the dynamics of spatial and
114 temporal LULC change in Los Molinos watershed from 1990 to 2020. Further, we promoted
115 the Normalized Difference Vegetation Index (NDVI) derived from a MODIS time-series
116 dataset (2001-2020) as an indicator that reflects the status of Los Molinos reservoir and to
117 explain the effects of LULC change on the trophic state of the reservoir. Hence, our study
118 represents a baseline that could provide tools for watershed management to evaluate variation
119 in LULC changes and to assess the interaction of these changes with the reservoir's water
120 quality.

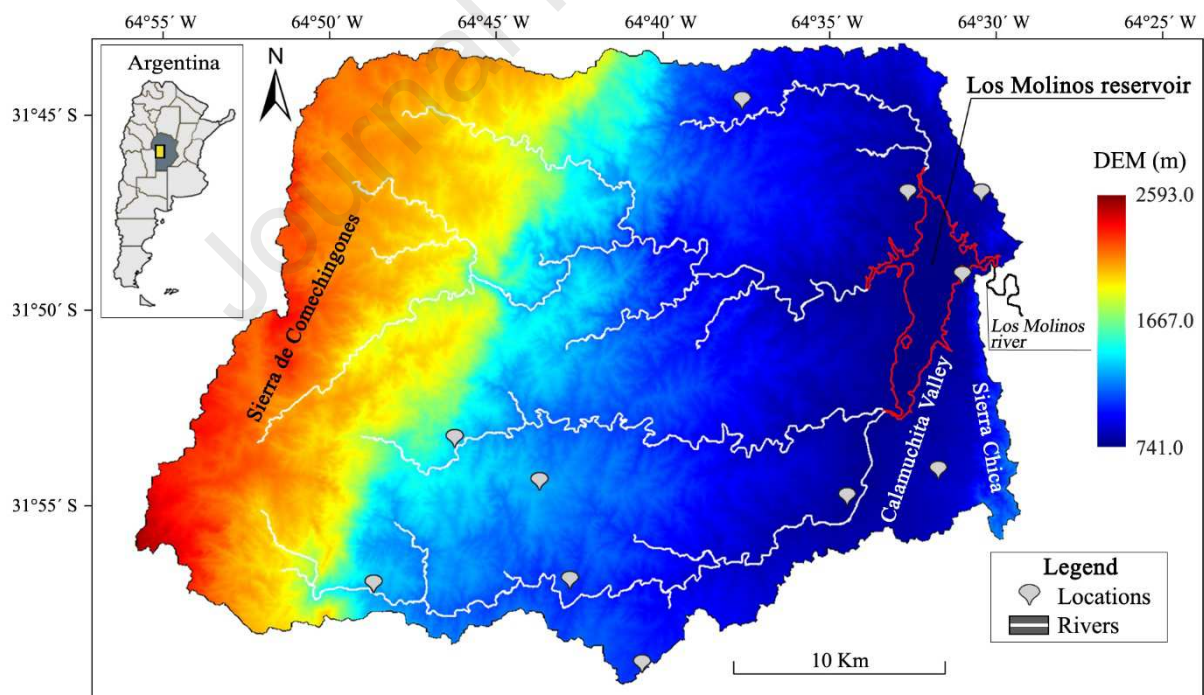
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122 2. METHODOLOGY

123 2.1 Study area

124 Los Molinos reservoir (31° 49' S, 64° 31' W) is located in Calamuchita Valley, in the
 125 central region of Argentina (Fig. 1). This reservoir has a maximum surface area of 2100 ha, a
 126 maximum volume of 399 hm³, and a maximum and mean depth of 53.0 and 14.0 m,
 127 respectively (Bonansea et al., 2018). Los Molinos reservoir has four main tributaries and one
 128 effluent called Los Molinos river. The watershed of this reservoir occupies an area of 904.89
 129 km². The watershed is characterized by a mountainous system called Sierra de
 130 Comechingones to the west and Sierra Chica to the east. Elevations of the area range from
 131 2593 to 741 m above sea level (masl), and lithology is dominated by granite with patches of
 132 metamorphic rocks (gneiss, schist, migmatite) (Cibils-Martina et al., 2017).

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Fig. 1. Location and digital elevation model (DEM) of the study area.

136

137 2.2 Analysis of LULC change

138 A Landsat image from the beginning of each decade was selected to retrospectively
139 characterize and assess the spatiotemporal variations in LULC change of Los Molinos
140 watershed between 1990 and 2020. Thus, four 30-m spatial resolution Landsat Level-2
141 images of the studied area (Path: 229, Row: 82) were downloaded from the USGS website
142 (<https://earthexplorer.usgs.gov/>), including one Landsat 5 Thematic mapper (TM) (February
143 28, 1990), two Landsat 7 Enhance Thematic mapper Plus (ETM+) (February 02, 2001;
144 February 11, 2010), and one Landsat 8 Operational Land Imager (OLI) (March 02, 2020). The
145 selected cloud-free images corresponded to the summer season, preferably between February
146 and March, when vegetation presents its highest productivity. Using images of the same
147 season, we have minimized discrepancies in reflectance of the evaluated land classes. The
148 ETM+ February 02, 2001 image was used instead of an image of the year 2000 because no
149 cloud-free images were available during the summer of that year.

150 The Landsat Level-2 scenes are on-demand products which are generated from the
151 corresponded Level-1 precision and terrain corrected product (L1TP). Level-2 data consists of
152 pixel values that estimate spectral reflectance as measured at the Earth's surface eliminating
153 the presence of atmospheric scattering or absorption effects (Teixeira Pinto et al., 2020).
154 Therefore, by accounting for atmospheric effects, Level-2 data products contain surface
155 reflectance values as measured *in-situ*, improving comparison between multiple images and
156 sensors and helping in the detection and characterization of LULC change.

157 Since May 31, 2003, ETM+ images have presented a failure called scan-line corrector
158 off (SLC-off) resulting in strips of data lost in the ETM+ images acquired afterwards (USGS,
159 2004). Fortunately, as the SLC-off has not affected the radiometric and geometric quality of
160 the ETM+ sensor, 80% of the pixels of each image are scanned in a perfect way (Chen et al.,

161 2011). To fill the un-scanned pixels of the February 11, 2010 ETM+ image, the SLC-off Gap-
162 Filled Products, Phase Two Methodology article was applied (USGS, 2004).

163 According to the characteristics of the watershed of Los Molinos reservoir, a support
164 vector machine classification algorithm provided by the ENVI software Feature Extraction
165 Module was used to classify and evaluate the LULC of the studied area. Different studies
166 suggest that a support vector machine algorithm could result in a higher classification
167 accuracy than the maximum likelihood classifier (Huang et al., 2002; Xu et al., 2013). A total
168 of six LULC classes were identified for the different periods, including wetland, bare land,
169 natural forest, timber plantation, agriculture and urban area. Wetland consisted of areas of
170 open water, including rivers, streams, and reservoirs. Bare land referred to sparse plant cover,
171 rocks, gravels, sandy land and bare soil. Natural forest consisted of areas dominated by
172 natural shrubs, thickets and herbs. Timber plantation corresponded to the exotic pines for
173 timber production. Agriculture consisted of agricultural and livestock development, and urban
174 included residential and industrial areas and roads. Between 60 and 120 training data (samples
175 of known identity) that were well distributed across the watershed were chosen for each land
176 use class of each image. *In-situ* investigations, previous author's knowledge of the studied
177 area, and high-resolution Google Earth Pro images were used to help in the identification and
178 validation of LULC classes. All processed images presented a kappa coefficient higher than
179 0.88, showing a good accuracy in the classification procedure (Chuvienco Salinero, 2008).
180 Thus, four LULC maps of Los Molinos reservoir watershed were obtained for the selected
181 years (1990, 2001, 2010 and 2020).

182 To describe the dynamic of LULC change in the studied watershed and in different
183 buffer zones defined from the shore of the reservoir, we generated a "from-to" change matrix
184 which was used to evaluate the transition from one LULC class to another between successive

185 land cover maps. Further, the quantification of LULC change was assessed by using the
186 LULC change rate described in the following Equation (Li et al., 2009):

$$187 \quad k = (S_2 - S_1)/S_1 * 1/T * 100\% \quad \text{Eq. (1)}$$

188 where k is the land-use change rate for each land-use type, S_1 and S_2 are the areas of the land-
189 use type at the beginning and at the end of a period, respectively, and T is the time interval
190 (years).

191

192 *2.3 NDVI time-series evaluation*

193 In the present study, the NDVI was used to estimate algal presence or algal blooms in
194 Los Molinos reservoir, due to its optical characteristics and because it is sensitive to pigment
195 absorption (Zhu et al., 2015). NDVI time-series of Los Molinos reservoir was derived from
196 MODIS sensor using the 250-m spatial resolution MOD13Q1 composition product. Each
197 value of MOD13Q1 is atmospherically corrected (Vermote and Vermeulen, 1999) and
198 indicates the best quality pixel value within the observed 16-days period (Zhang et al., 2019).
199 Dates that present pixels with clouds over the reservoir were removed using the quality flag
200 information in MOD09GQ. In order to avoid land effects, for each evaluated date an NDVI
201 average value was calculated among pixels located further than 250 m from the shore of the
202 reservoir (Germán et al., 2020).

203 NDVI ranges from -1 to +1. Negative values usually indicate pure water pixels and
204 positive values are strongly correlated with the greenness of the water due to the presence of
205 photosynthetic organisms such as algae (Germán et al., 2020). Values equal or greater than
206 0.1 can be interpreted as algal scums or blooms (Kiage and Douglas, 2020).

207 MODIS sensor is operational since 2000, so the NDVI time-series was created from
208 January 2001 to January 2020. The derived MODIS NDVI time-series was analysed using the

209 Local Regression model (LOESS), which is a non-parametric approach that uses local
210 weighted regression to fit a smooth curve (Cleveland et al., 1992). This method can reveal
211 trends and/or cycles presented in the data set that might be difficult to model with a
212 parametric curve (Cleveland et al., 1992). Using a seasonal-trend decomposition model based
213 on LOESS method (Cleveland et al., 1990), the time-series (Y_t) was decomposed into trend
214 (T_t), seasonal (S_t), and remainder or random (e_t) components at each period of time. Thus, the
215 NDVI time-series was simply described as the sum of the three components using the
216 following equation:

$$217 \quad Y_t = T_t + S_t + e_t \quad \text{Eq. (2)}$$

218 Finally, to compare the variable in different time periods, normalized annual
219 anomalies at each 5 years period boxplots were used.

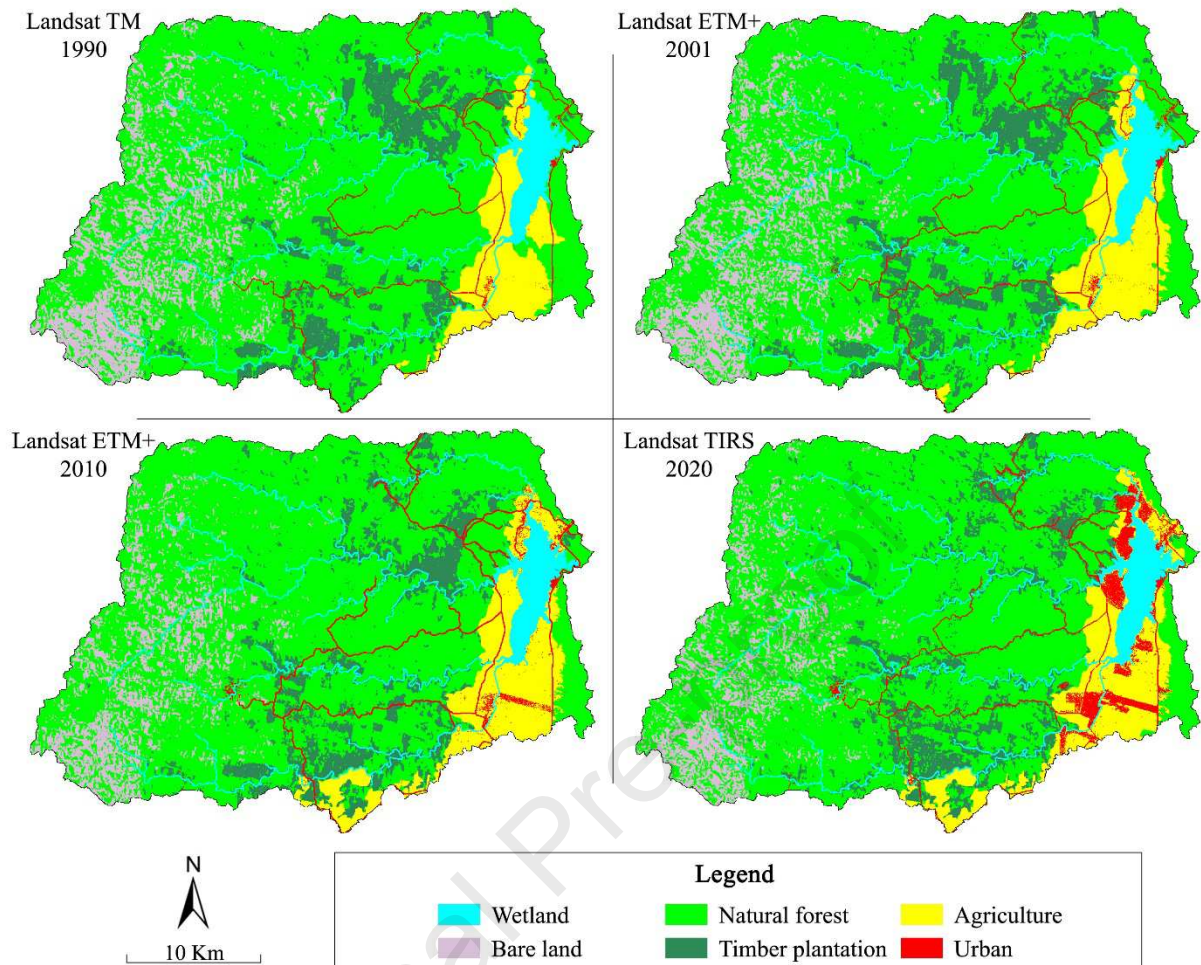
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221 **3. RESULTS AND DISCUSSION**

222 *3.1 Land use changes at watershed scale*

223 The spatio-temporal distribution of LULC in Los Molinos watershed was obtained
224 using Landsat imagery from 1990 to 2020 (Fig. 2). During the studied period, the watershed
225 was mainly dominated by natural forest, which slightly decreased at an annual rate of -0.04%,
226 from 67.02% in 1990, to 66.23% of the total watershed in 2020 (Fig. 3). The most important
227 transition of natural forest was to timber plantation, followed by bare land and agriculture
228 (Table 1).

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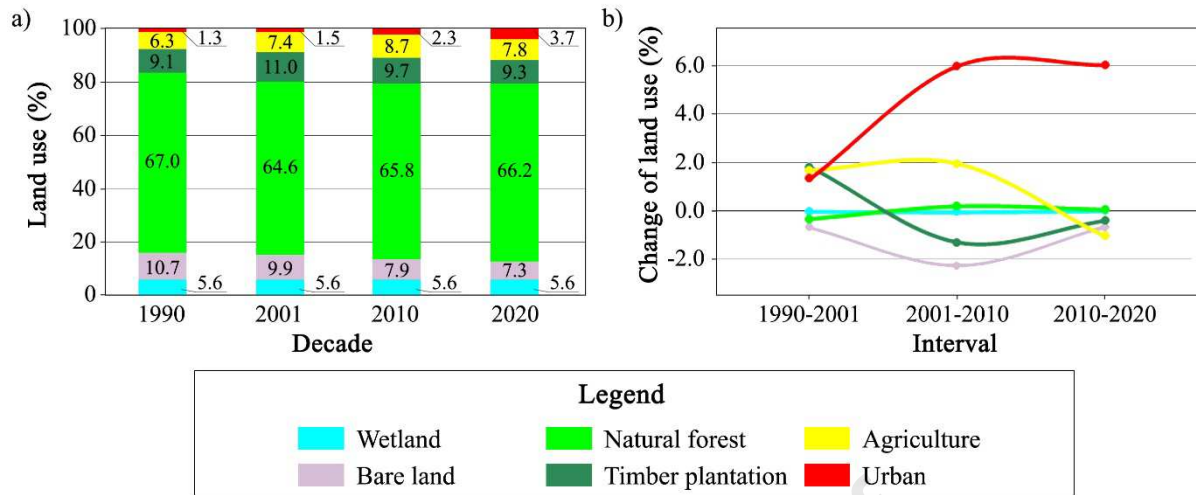
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Fig. 2. LULC classification maps in the watershed of Los Molinos reservoir for the studied

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period (1990-2020).

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Fig. 3. a) LULC proportions for each decade; b) Comparison of the annual rate of change for

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each class during the studied period.

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Table 1. Transition matrix of land use change in Los Molinos watershed during the studied

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period (km²).

Period	LULC	Wetland	Bare land	Natural forest	Timber plantation	Agriculture	Urban	Total
1990-2001	Wetland	50.85	0.00	0.00	0.00	0.00	0.01	50.86
	Bare land	0.00	70.60	25.90	0.01	0.00	0.04	96.55
	Natural forest	0.00	18.64	541.04	34.68	11.17	0.95	606.47
	Timber plantation	0.00	0.03	17.99	64.40	0.02	0.22	82.65
	Agriculture	0.00	0.00	0.06	0.00	55.95	0.58	56.59
	Urban	0.00	0.00	0.02	0.02	0.01	11.73	11.77
	Total	50.85	89.26	585.00	99.11	67.15	13.52	-
2001-2010	Wetland	50.77	0.00	0.00	0.00	0.00	0.08	50.85
	Bare land	0.00	54.87	34.32	0.01	0.00	0.06	89.26
	Natural forest	0.00	16.20	516.96	34.28	14.71	2.85	585.00
	Timber plantation	0.00	0.02	43.63	53.18	0.57	1.71	99.11
	Agriculture	0.00	0.00	0.67	0.00	63.68	2.80	67.15
	Urban	0.00	0.01	0.11	0.01	0.07	13.31	13.52
	Total	50.77	71.11	595.69	87.49	79.03	20.81	-
2010-2020	Wetland	50.77	0.00	0.00	0.00	0.00	0.00	50.77
	Bare land	0.00	47.23	23.83	0.03	0.00	0.01	71.11
	Natural forest	0.00	19.25	537.14	32.41	5.10	1.79	595.69
	Timber plantation	0.00	0.00	36.45	50.18	0.66	0.20	87.49

	Agriculture	0.00	0.00	0.46	0.01	64.18	14.38	79.03
	Urban	0.06	0.00	1.46	1.37	0.93	16.99	20.81
	Total	50.83	66.48	599.34	83.99	70.87	33.38	-

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Intensive land use classes, which include timber plantation, agriculture and urban area, increased during the studied period (1990-2020). However, the change of these land uses did not occur at equal rates. Between 1990 and 2001, timber plantation presented a strong increase at an annual rate of 1.81 %, mainly from natural forest (34.68 km²) (Fig. 3.b, and Table 1). Since 2001, due to timber production (Cibils-Martina et al., 2017) and wildland fires (Bonansea and Fernandez, 2013; Fernandez et al., 2014), the area occupied by sawn plantations has decreased at an annual rate of -1.30 % from 2001 to 2010, and -0.40 % from 2010 to 2020. During these time periods, 43.63 km² and 36.45 km² of timber plantation changed to natural forest, respectively. According to different authors, scarce information about forest management and wood processing in the region discourages the application of environmental and economical sound forest practices that increase the productivity of forest land (Dorado et al., 1997; Cibils-Martina et al., 2017).

Agriculture, mainly located in Calamuchita Valley and around Los Molinos reservoir, showed an annual growth rate of 0.84% during the 30-year studied period. During the periods 1990-2001 and 2001-2010, agriculture land increased at an annual rate of 1.70 and 1.97 %, respectively. The most important transition to agriculture was from natural forest (11.17 km², for the period 1990-2001, and 14.71 km² for the period 2001-2010). From 2010 to 2020, agriculture has decreased at an annual rate of -1.03 % due to conversion to urban area.

The most consistent upward trend of class change was the progressive gain of urban area. During the last 30 years, urban area has had an increasing annual rate of 6.12 %. In the beginnings of 1990s, urbanization occupied an area of 1.30 % (11.77 km²) of the total watershed. During this decade, this category has increased at an annual rate of 1.34 %.

263 However, since 2001 urban area has abruptly increased its annual rate to 6.00 and 6.03 %, for
264 the periods 2001-2010 and 2010-2020, respectively. Most of the urban area was obtained
265 from a conversion of agriculture, mainly between 2010 and 2020, where 14.38 km² of
266 agriculture land was converted to urbanization.

267 Results show that urbanization increased mainly on the shore of Los Molinos reservoir
268 (Fig. 2). Therefore, an exhaustive LULC change analysis was done in different buffer zones
269 defined from the shore of the reservoir (Fig. 4). By analysing the different buffer zones, we
270 observed that from 1990 to 2001 the rate of change was similar for each land use class with
271 the exception of timber plantation, which presented the highest land-use change rate in the 1
272 km buffer zone (14.59 %) due to the expansion of a timber plantation located in the northwest
273 area of this buffer zone (Fig. 5.a). Figure 5 also shows that the increase in urban area until
274 2001 was unremarkable in the different buffer zones. However, since 2001 different urban
275 investment projects, primarily located close to the shores of Los Molinos reservoir, have
276 generated a drastic change in LULC of the watershed. This was more evident in the 1 km
277 buffer zone of the period 2010-2020, where urban area presented a drastic increasing annual
278 rate of 18.02 %, and the proportion of land occupied by urbanization presented an exponential
279 increase (6.06 % in 1990, 6.97 % in 2001, 10.96 % in 2010, and 30.71% in 2020). In addition,
280 Figure 5.b shows that during the last decades the most consistent trend of inter-class changes
281 in the buffers zone has been a progressive loss of natural forest, which was replaced by the
282 increase of intensive land use classes. For example, at present, more than the 80 % of the 1
283 km buffer zone corresponds to intensive land use classes, mainly agriculture (46.40 %) and
284 urban area (30.71 %).

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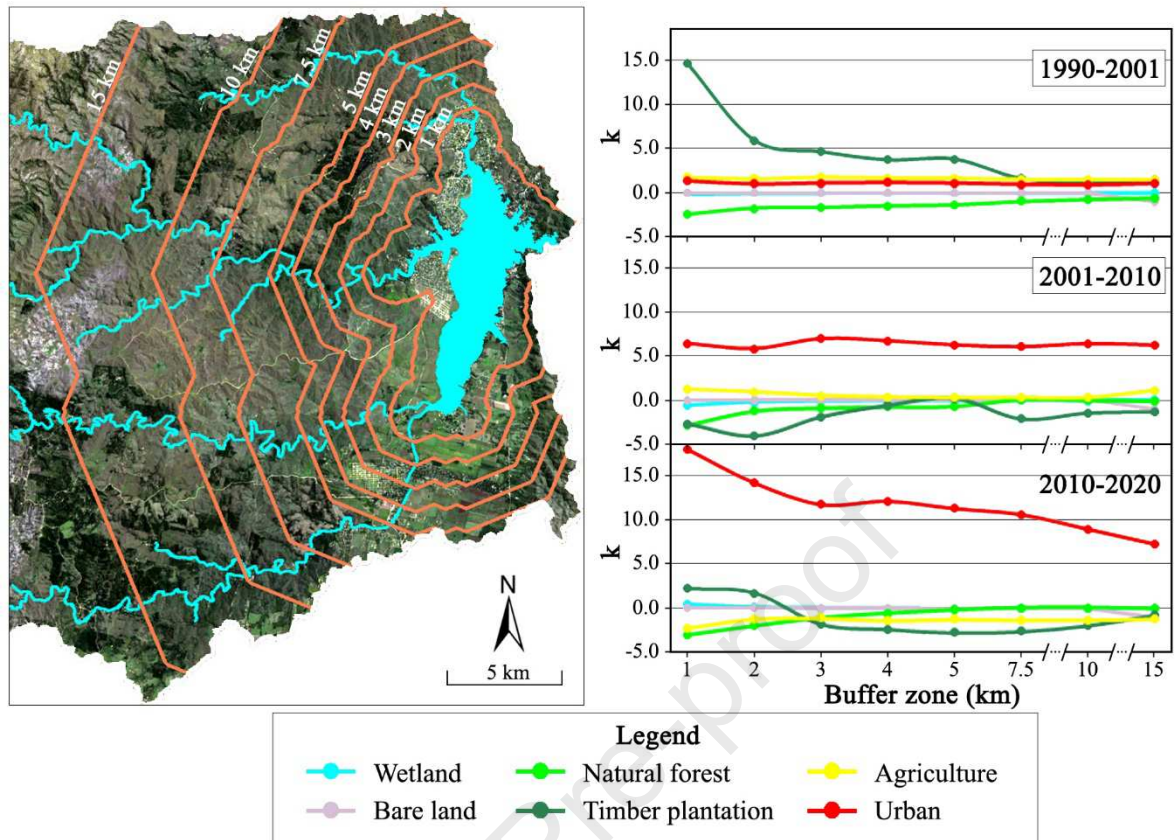
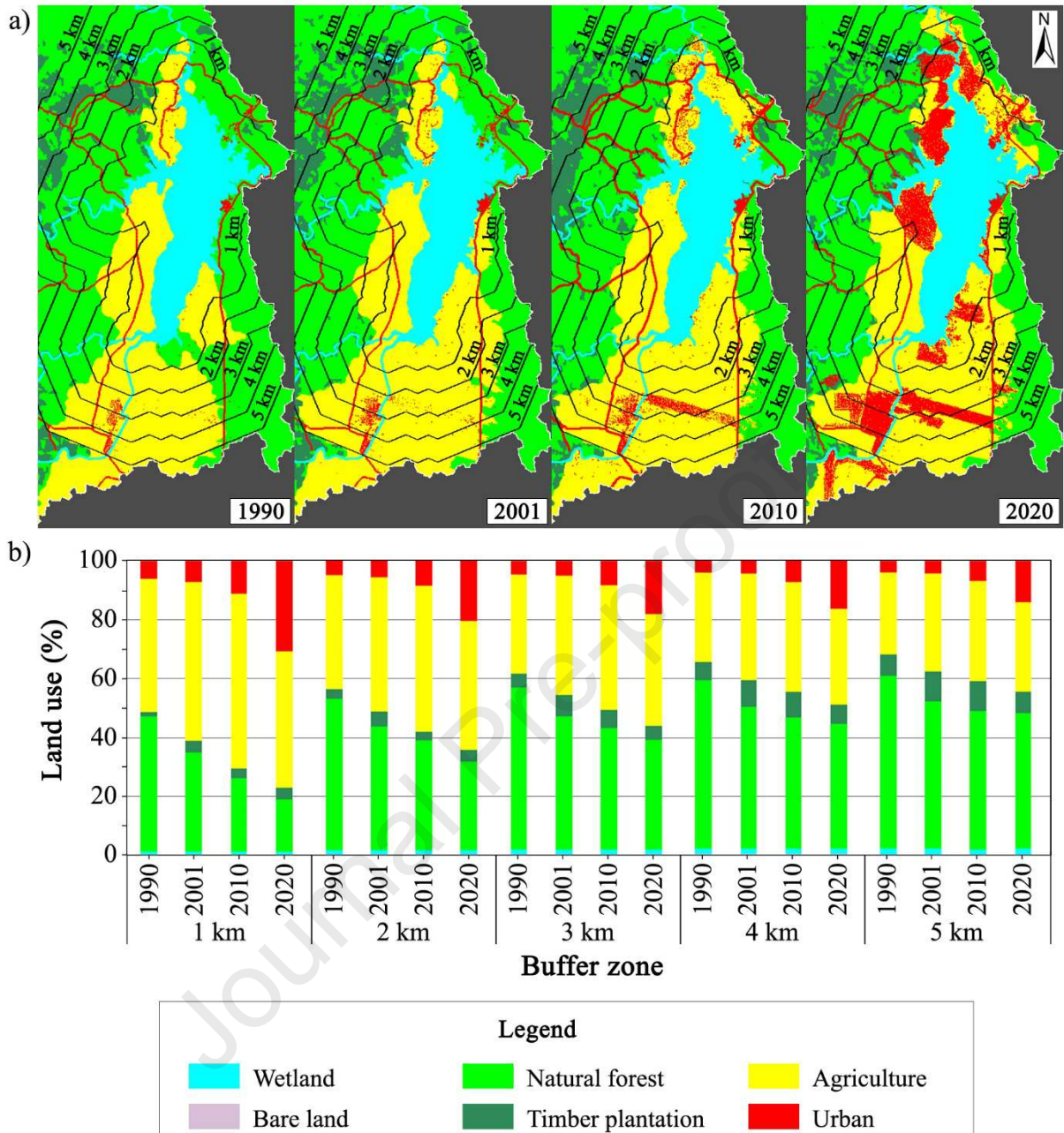


Fig. 4. Comparison of the rate of change for each class in different buffer zones.

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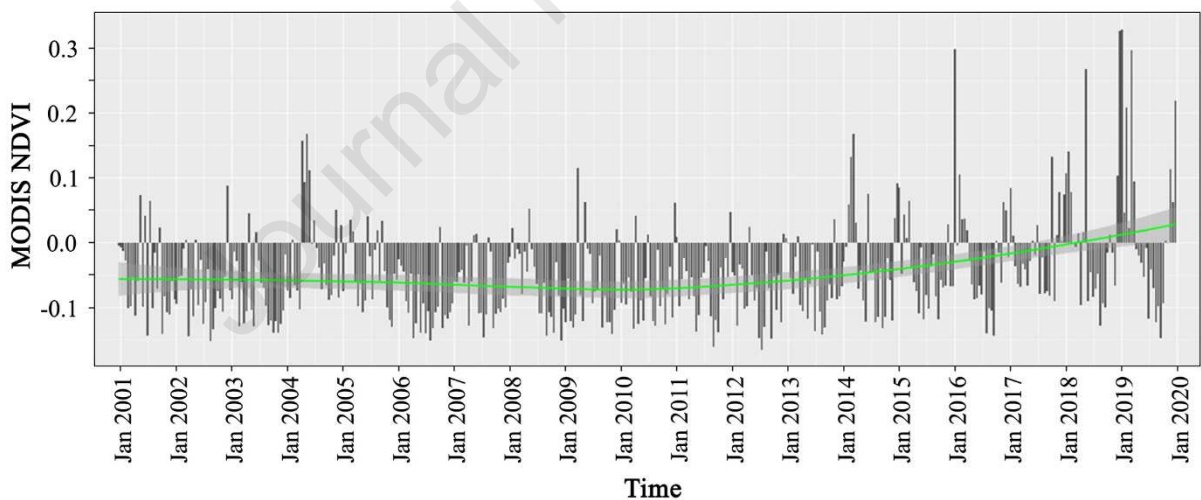
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290 **Fig. 5.** a) LULC classification maps of the buffer zones defined from the shore of the
291 reservoir (1990-2020); b) LULC proportions for each buffer zone during the studied period.

292 293 3.2 NDVI time-series analysis

294 Figure 6 shows the 16-days averaged NDVI time-series generated from MODIS
295 sensor for Los Molinos reservoir. The calculated NDVI values range from -0.17 (on July 11,
296 2012) to 0.33 (on January 01, 2019). It is possible to observe that from 2001 to 2013 the

297 NDVI trend-line remained practically constant and retrieved NDVI values were similar to -
298 0.05. However, since 2014 the trend-line has had an increasing behaviour, passing from
299 negative NDVI values, typical of pure water, to positive values which could be related to the
300 presence of algae. Further, since 2014 there has been an increase in the amount of positive
301 retrieved NDVI values compared with previous years. This is in accordance with Figure 7.a
302 which shows that positive annual anomalies of NDVI had been found since 2014.
303 Additionally, Figure 7.b shows that since 2014 there has been an increase in the number of
304 records that exceed the NDVI threshold of 0.1, which according to Germán et al. (2020), and
305 Kiage and Douglas (2020), it could be related with the presence of algal blooms in the
306 reservoir. These events have been more evident during the last two years (2018 and 2019),
307 followed by 2004.

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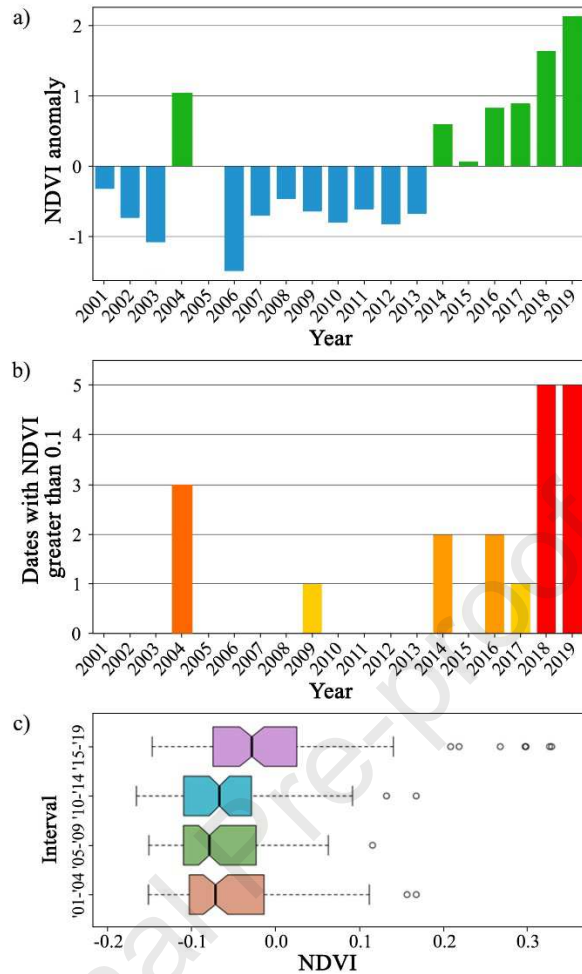
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Fig. 6. Time-series of MODIS NDVI. Green line represents the NDVI trend-line generated

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with the LOESS smoothing model.

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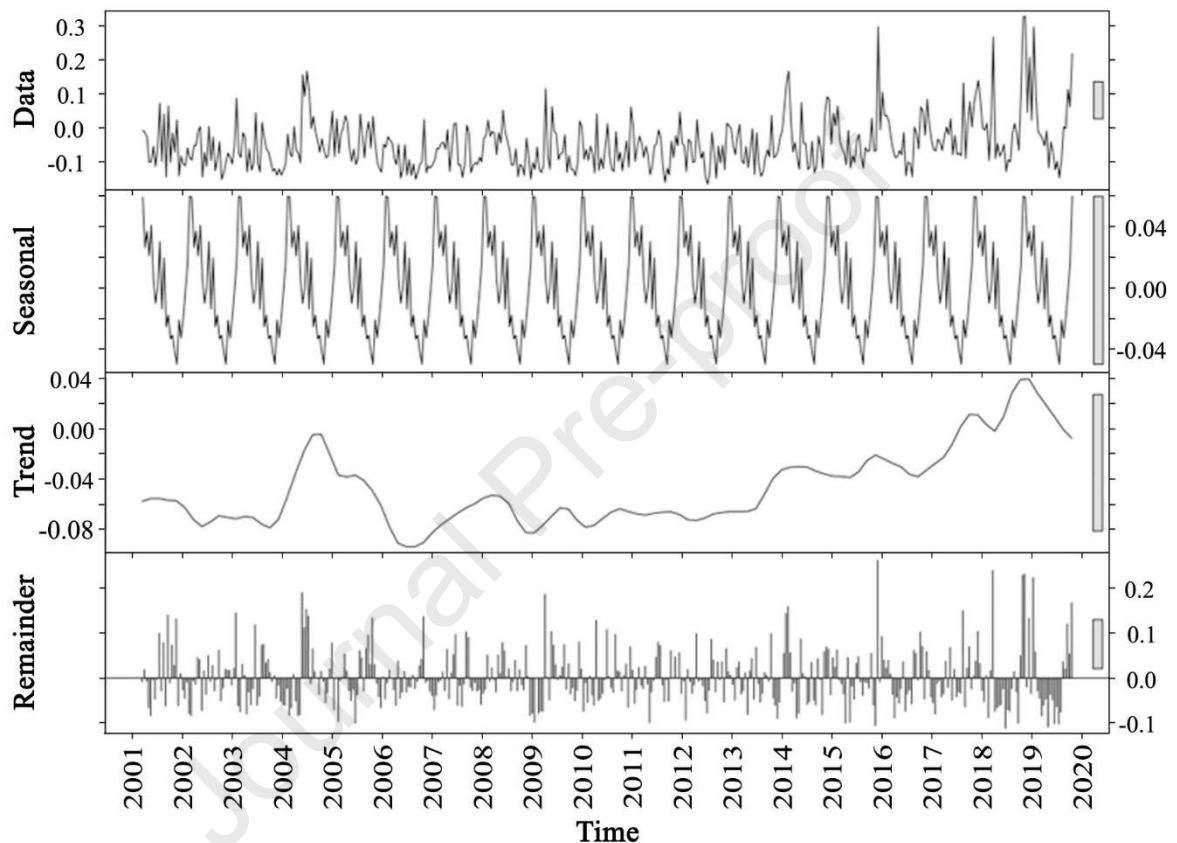
322

323

Fig. 7. a) Inter-annual variations in NDVI anomaly; b) Number of dates whose NDVI were higher than 0.1; c) Box-plots of NDVI classified by time interval.

Figure 8 shows the components of the decomposing time-series dataset. This Figure, which is in agreement with our results, highlights that the NDVI trend component presented a peak in 2004. Afterward, the NDVI trend-line went down, and then it kept fluctuating until 2014. Since 2014 the trend-line has presented an increasing trend curve, and higher values have been registered during the last five years. This is also proven in Figure 7.c which shows that the box-plot created for the interval 2014-2019 presented the highest median value. In addition, the outliers or extreme values observed in the 2014-2019 interval, which can be

324 related with algal scums or blooms, were higher than the rest. Finally, the seasonal component
 325 of the decomposing time-series dataset (Fig. 8) shows two main periodic peaks that were
 326 related to warmer water registered in summer (January to March) and spring (September to
 327 December) seasons which could lead to algal growth (Bonansea et al., 2019).
 328



329 **Fig. 8.** Decomposed seasonal, trend and remainder components for MODIS NDVI time-
 330 series.
 331

332

333 3.3 Relationship between LULC change and reservoir water quality

334 During the last decades, the watershed of Los Molinos reservoir has experienced an
 335 intensive LULC change due to the impact of human activities. Although the watershed is
 336 dominated by natural forest, intensive land use classes have principally increased in the region
 337 of Calamuchita Valley and around Los Molinos reservoir. The most significant LULC change

338 started two decades ago and it has intensified during the last ten years. This change was
339 related with the intensification of agriculture activities, and with the increasing conversion of
340 agricultural land to urban area, primarily related with urban investment projects located close
341 to Los Molinos reservoir.

342 Results of this study also indicate that the degradation trend of LULC in the watershed
343 significantly contributed to the degradation of water quality of Los Molinos reservoir. This
344 was corroborated by analysing the MODIS NDVI time-series. As results showed, from 2001
345 to 2013, the NDVI trend-line remained practically constant and low values related with pure
346 water were retrieved. However, since 2014 and as a consequence of urban expansion, the
347 NDVI trend-line has presented an increasing behaviour, and extreme values, which could be
348 related to algal blooms, have been more frequently observed, mainly during the last two years.

349 Although urbanization and agriculture activities have generated an economic
350 development for the region, different studies suggest that watersheds are extremely sensitive
351 to an increase in intensive land use classes with significant impacts on eutrophication and
352 water quality of aquatic environments, including rivers, lakes and reservoirs (Welde and
353 Gebremariam, 2017; Desta and Fetene, 2020; Kiage and Douglas, 2020; Torregroza-Espinosa
354 et al., 2020; Wei et al., 2020). During the process of urbanization, a permeable vegetated land
355 surface is replaced by concrete and other impervious materials which can drastically
356 transform surface hydrological processes altering the balance between infiltration and runoff.
357 This is because the rain retention capacity is reduced, and runoff and erosion are sharply
358 increased (Shi et al., 2007; Rojas et al., 2019; Kiage and Douglas, 2020). According to
359 Huang et al. (2014), the increase in urban area also generates an increase of domestic sewage
360 discharge, which may significantly increase the amount of nutrients in waters causing
361 negatively impacts on water quality of water bodies. Furthermore, the adverse impacts of

362 agriculture expansion on water quality related with point and non-point pollution sources have
363 been assessed by different authors. According to Welde and Gebremariam (2017), large
364 amounts of chemical fertilizers used in agriculture activities flow to rivers and into the
365 reservoir after rains, being an important factor increasing eutrophication process and algal
366 blooms. Further, livestock manure, heavy metals and other organic and inorganic materials
367 produced by agricultural activities can easily enter into water bodies through surface runoff,
368 leakage, drainage, evaporation, and other processes (Wei et al., 2020).

369 Therefore, due to the increasing behaviour of the NDVI trend-line registered in the
370 reservoir which can be related with LULC change at a watershed scale and with rapid
371 urbanization along the lakeshore, we could expect that if water environmental policies do not
372 focus on controlling the discharge of urban sewage and agricultural pollution sources, Los
373 Molinos reservoir will be entering into an accelerated eutrophication process with negative
374 consequences on water quality and human health .

375 Finally, the extreme values of NDVI registered during 2004 could be related with a
376 wildland fire that occurred in the watershed of Los Molinos reservoir during September, 2003,
377 which was considered the most serious of the decade (Fernandez et al., 2014). Wildland fires
378 lead to changes in vegetation structure and soil properties increasing the risk of runoff,
379 biomass loss and severe erosion (Miller and Yool, 2002). Wildland fires can also have a
380 significant impact on water quality (Germán et al., 2018). During the fire event registered in
381 Los Molinos watershed, a total of 62.6 km² was burned (Bonansea and Fernandez, 2013).
382 This fire was followed by a severe drought and by December, 2003 a rainstorm caused great
383 floods on the reservoir's tributaries with high concentration of ashes. As a consequence, the
384 input of flows with high concentration of suspended sediments and nutrients into the water
385 body were registered, causing the occurrence of an algal bloom event which could have led to

386 an increase in retrieved NDVI values registered in 2004 (Bazán et al., 2005; Bonansea and
387 Fernandez, 2013; Fernandez et al., 2014).

388

389 **4. CONCLUSIONS**

390 The present study was carried out to assess the spatial and temporal variations in LULC in
391 Los Molinos watershed. Further, the relationship between LULC change and water quality of
392 the reservoir was analysed.

393 Satellites images, acquired by Landsat sensors provided evidences that Los Molinos
394 watershed has undergone significant LULC alterations and transformations, mainly in the
395 region of the Valley and close to the shore of the reservoir. Although natural forest is the
396 dominated land use class, intensive land use classes, such as agriculture and urbanization,
397 show an increasing trend, confirming that LULC change has been principally influenced by
398 human activity.

399 The most consistent upward trend of class change at a watershed scale was the
400 progressive gain of urban area, which has shown an annual increasing rate of 6.12 % during
401 the last 30 years. The maximum LULC change has occurred during the last decades due to
402 investment projects located close to the shores of Los Molinos reservoir that have replaced
403 agriculture land by urbanization. During the period 2010-2020, urban area located in the 1 km
404 buffer zone of the reservoir increased at an annual rate of 18.02 %. It is important to note that
405 at present, more than the 80 % of the 1 km buffer zone corresponds to intensive land use
406 classes, mainly agriculture (46.40 %) and urban area (30.71 %). Although urbanization and
407 agriculture activities have generated an economic development for the region, an increase in
408 eutrophication and a decrease in water quality of Los Molinos reservoir have been observed.
409 This was evidenced by analysing the MODIS NDVI time-series which showed that since

410 2014 as a consequence of LULC change, the NDVI trend-line has presented an increasing
411 behaviour and extreme values of NDVI related to algal blooms have been more frequently
412 observed, primarily during the last two years.

413

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419

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- 586

- 1 • LULC of the watershed was obtained using Landsat imagery from 1990 to 2020.
- 2 • MODIS NDVI time-series was used to reflect the status of Los Molinos reservoir.
- 3 • The relationship between LULC change and water quality was assessed.
- 4 • Intensive land use classes lead to water quality deterioration and eutrophication.

Journal Pre-proof

Declaration of interests

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

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