

Tracking Nature's Footprint: Uncovering NDVI Time Trends in Spanish High Mountain Biosphere Reserves

Patricia Arrogante-Funes^{1,3}, Dina Osuna¹, Fátima Arrogante-Funes^{2*}, Ariadna Álvarez-Ripado¹, Adrián G. Bruzón¹.

¹Department of Chemical and Environmental Technology, ESCET, Rey Juan Carlos University, C/Tulipán s/n, Móstoles, 28933 Madrid, Spain

² Universidad de Alcalá, Environmental Remote Sensing Research Group, Department of Geology, Geography and the Environment, Calle Colegios 2, 28801 Alcalá de Henares, Spain

³ Research Group on Technologies for Landscape Analysis and Diagnosis (TADAT), Rey Juan Carlos University, C/Tulipán s/n, Móstoles, 28933 Madrid, Spain

*Correspondence: fatima.arrogante@uha.es

1 Tracking Nature's Footprint: Uncovering NDVI Time Trends in 2 Spanish High Mountain Biosphere Reserves

3
4
5 Protecting Spain's mountain ecosystems is of great importance to preserve ecosystem services.
6 Global warming is increasing the vulnerability of these habitats, especially in peninsular Spain.
7 Biosphere Reserves are internationally protected areas that seek to protect biodiversity and, at the
8 same time, promote sustainable development. Evaluating these protected areas is essential to
9 verify environmental changes and establish priorities in their management. In this work, we have
10 studied the time trends of NDVI in the high mountain Biosphere Reserves of Spain, in the period
11 from 2001 to 2016, to check if the trend patterns are associated with: the Biosphere Reserves
12 studied, surface temperature, water stress, altitude, slope, habitat type, biogeographic region,
13 zonation, and distance to population centres. Significant differences were found between NDVI
14 trends and high mountain Biosphere Reserves. First, significant positive trends in NDVI were
15 observed when analyzing both reserves together. However, significant differences were found
16 between the two reserves. The Ordesa-Viñamala Reserve shows higher positive NDVI trends and
17 lower negative trends, while this pattern is reversed in the Sierra Nevada. Temperature and water
18 stress affect the Sierra Nevada Reserve to a greater extent, increasing the number of negative
19 NDVI trends. Habitat types show different patterns depending on the Reserve, with less resistance
20 and, therefore, more vulnerable habitats. It has been observed that the zoning of the reserves
21 significantly affects NDVI trends, following a pattern of increasing negative trends as we move
22 away from the core. The height variable appears to substantially affect NDVI trends, with an
23 increase in positive trends as we move in altitude. Finally, the distance to population centers has
24 significantly influenced NDVI trends; generally, greater negative trends have been observed in
25 areas closer to municipalities.

26
27 **Keywords:** NDVI, time series, remote sensing, Biosphere Reserve, high mountain, protected
28 areas.

29 1. Introduction

30
31 Spain is one of the European countries with the greatest ecosystems, habitats and
32 species diversity. It is home to more than half of the vertebrate and vascular plant species,
33 a high number of endemic species and 65% of priority habitats in the European Union
34 (OSE, 2010).

35
36 From the point of view of goods and services, these ecosystems fulfil three main
37 functions: productive, environmental and social (Rodà et al., 2003). In their productive
38 function, ecosystems provide renewable natural resources. Environmental services
39 include the maintenance of biodiversity, climate regulation, water cycle regulation,
40 biogeochemical cycles and soil conservation, among others. Finally, social services are
41 related to the recreational, leisure, educational, or research uses that these ecosystems can
42 provide and which, in some areas, are economic drivers for their development (Valladares
43 et al., 2005).

44 Vegetation is conditioned by climate, specifically by radiation, temperature,
45 precipitation and atmospheric humidity (Nemani et al., 2003). The great natural climatic
46 variability present in the Iberian Peninsula makes some ecosystems particularly
47 vulnerable to the effects of climate change. Mountainous areas are particularly relevant
48 for water resource generation, particularly in temperate and semi-arid zones, including
49 the Mediterranean zone. These mountainous areas are experiencing increased water stress
50 caused by increased temperatures and reduced precipitation (García-Ruiz et al., 2011). In
51 addition, these mountainous areas are one of the most vulnerable regions in the world,
52 where temperature changes are more pronounced than in flatter areas (Pepin et al., 2015).

53

54 It has been shown that Mediterranean ecosystems are more vulnerable to climate
55 change than other ecosystems in Europe (Schröter et al., 2005). These variations in
56 climate have direct consequences on vegetation development and, thus, ecosystems and
57 their services, resulting in ecological and economic losses (Field et al., 2012). In addition
58 to climatic factors, these differences in vegetation growth can also be explained by
59 anthropogenic (Fensholt et al., 2012) and topographic factors (Allen et al., 2010).

60

61 In order to protect natural diversity and ensure sustainable development, many
62 countries worldwide have declared different types of protection (Dudley, 2008).
63 Biosphere Reserves are areas designated by the United Nations Educational, Scientific
64 and Cultural Organisation (UNESCO) in the People and Biosphere Programme (MaB
65 Programme) framework. They are included in Law 33/2015 (BOE, 2015), of 21
66 September, which amends Law 42/2007 (BOE, 2007), of 13 December, on Natural
67 Heritage and Biodiversity, in the section on Areas protected by international instruments.
68 These areas of high natural value aim to reconcile nature conservation with sustainable
69 development. The participation of the local population and other social sectors, such as
70 economic agents and the scientific sector, is an essential part of this integrated approach.

71

72 Spain currently has 52 Biosphere Reserves, encompassing areas in both terrestrial
73 and coastal/marine ecosystems, 4 of which are transboundary (three with Portugal and
74 one with Morocco), making it the country with the highest number of Biosphere Reserves
75 in the world ([https:// miteco.gob.es](https://miteco.gob.es)).

76

77 The Biosphere Reserves consist of three areas, each with different functions but
78 complementing each other and contributing jointly to achieve the objectives of the
79 Reserve. These zones are:

80 - Core zone: this is the legally protected area in which conservation is prioritized.

81 - Buffer zone: the area surrounding or bordering the core zone, where activities
82 compatible with conservation are allowed.

83 -Outer transition zone is dedicated to sustainable economic and human
84 development.

85 Each country's government proposes the declaration of a new Biosphere Reserve
86 and is responsible for proper functioning with the designated territories' competent
87 entities. For example, in Spain, it depends on the Ministry for Ecological Transition and

88 the Demographic Challenge (MITECO) and, in turn, on the Autonomous National Parks
89 Organisation (OAPN) ([https:// miteco.gob.es](https://miteco.gob.es)).

90

91 For the correct management of these areas, it is essential to know how the effects
92 of climate change influence ecosystems. The Normalized Vegetation Index (NDVI) is
93 useful for studying ecological responses to environmental changes (Pettoirelli et al.,
94 2005). In addition, the study of NDVI time series has proven effective in finding
95 significant patterns or trends in vegetation (Bradley & Mustard, 2008).

96

97 NDVI is a vegetation index representing the fraction of photosynthetically active
98 radiation intercepted by vegetation (fPAR) and primary productivity (Tucker & Sellers,
99 1986). It is calculated from the reflectance in the red (R) and near-infrared (NIR)
100 according to the following formula: $NDVI = (NIR - R)/(NIR + R)$ (Tucker & Sellers,
101 1986). NDVI values range from -1 to 1. NDVI values close to zero are associated with
102 diseased or sparse vegetation (presence of soil). Conversely, values very close to zero or
103 negative NDVI values may indicate the presence of snow, clouds or water (Aguayo &
104 CIREN, 2013). In this way, NDVI variations over time can show the vegetation's
105 phenological dynamics and changes in ground cover.

106

107 NDVI is an index obtained by remote sensing. This technique is based on
108 acquiring data from the Earth's surface through sensors carried on aircraft or satellites and
109 their subsequent processing for analysis and interpretation (Chuvieco Salinero, 2002).
110 Remote sensing has several advantages over other earth observation techniques, such as
111 aerial photography or direct observation. Among its benefits are that it offers global and
112 periodic coverage, allows information on regions of the non-visible electromagnetic
113 spectrum, and enables images to be obtained at different scales. Besides, the different
114 orbits traced by the satellites allow different time frequencies to be obtained over the same
115 territory. Finally, recording information in digital format speeds up interpretation
116 processes and facilitates data processing by geographic information systems (Martínez
117 Vega et al., 2010).

118

119 Nowadays, remote sensing is a key technology for monitoring vegetation indices
120 that give us an idea of the vegetative state of plant masses. In this respect, passive satellite
121 platform sensors have provided free satellite images with adequate spatial and temporal
122 resolution for over fifteen years. For example, the MODIS (Moderate Resolution Imaging
123 Spectroradiometer) sensor on board the AQUA and TERRA satellites of the National
124 Aeronautics and Space Administration (NASA) records reflection data of the Earth's
125 surface, which is stored in 36 bands of the electromagnetic spectrum. This sensor offers
126 many products that can be used for land use and land cover mapping, including NDVI at
127 different spatial resolutions. In addition, this sensor has been frequently used in vegetation
128 time series studies to characterize vegetation (Busetto et al., 2010; Hmimina et al., 2013;
129 Running et al., 2004).

130

131 These studies are key to deriving long-term habitat management guidelines under
132 global change scenarios. In particular, this paper focuses on assessing high mountain
133 Biosphere Reserves using remote sensing and time series. These protected areas are
134 important because they are specially designated areas for assessing and managing natural
135 and cultural resources and looking at their relationships in sustainable development
136 scenarios (Batisse, 1982). However, although they can be vital from a research point of
137 view for the management and management of the territories they contain, they have not
138 been studied much to date, and even less so utilizing remote monitoring techniques, as
139 proposed in this study.

140
141 In the framework of other NDVI studies carried out in Spanish mountain areas in
142 the last decades (most of them protected areas) (Alcaraz-Segura, Liras, et al., 2009; P.
143 Arrogante-Funes et al., 2018; Khorchani et al., 2018), in which positive trends of NDVI
144 proved to be related to global warming, topographic factors and anthropogenic factors,
145 the objectives of the present work are:

146
147 (1) To study the annual trends of NDVI in the Spanish high mountain Biosphere
148 Reserves, i.e. the Ordesa-Viñamala and Sierra Nevada Biosphere Reserves; and (2) to
149 study the possible association between NDVI trend patterns (from 2001 to 2016) and the
150 Biosphere Reserves of interest, their zones (core, buffer and transition), the mean annual
151 temperature trend, the annual water stress index trend, the elevation, and the distance to
152 population centres.

153 **2. Materials and Methods**

154 **2.1 Study Site**

155 The study sites in the present work are the High Mountain Biosphere Reserves of
156 Spain. These protected zones correspond to the Ordesa-Viñamala and Sierra Nevada
157 Biosphere Reserves.

158 159 **2.2.1 Ordesa-Viñamala**

160
161 The Ordesa-Viñamala Biosphere Reserve (Figure 1) was declared in 1977 and is
162 one of the first two announced in Spain. Later it underwent an extension in 2013. The
163 Biosphere Reserve is in the Central Aragonese Pyrenees, Huesca province, in "Axial
164 Pyrenees" and "Sierras Interiores". It has 117,364.1 ha, of which 13.71% correspond to
165 the core zone, 37.28% to the buffer zone and 49% to the transition zone. Inside, it has a
166 population of 5,639 inhabitants. The managing entity is the Ordesa-Viñamala Biosphere
167 Reserve Consortium (Calvo, 2019).

168
169 In addition, within the Reserve, there are the following protection figures:

- 170 - Ordesa y Monte Perdido National Park

- 171 - World Heritage Site, natural and cultural "Pyrenees Monte Perdido", declared by
- 172 UNESCO with mixed character (Spanish, French)
- 173 - Natural Monuments of the Pyrenean Glaciers
- 174 - Special Protection Areas for Birds (ZEPA)
- 175 - Places of Community Importance (SCI)
- 176 - Sobrarbe Geopark
- 177 - Viñamala and Circus Hunting Reserves

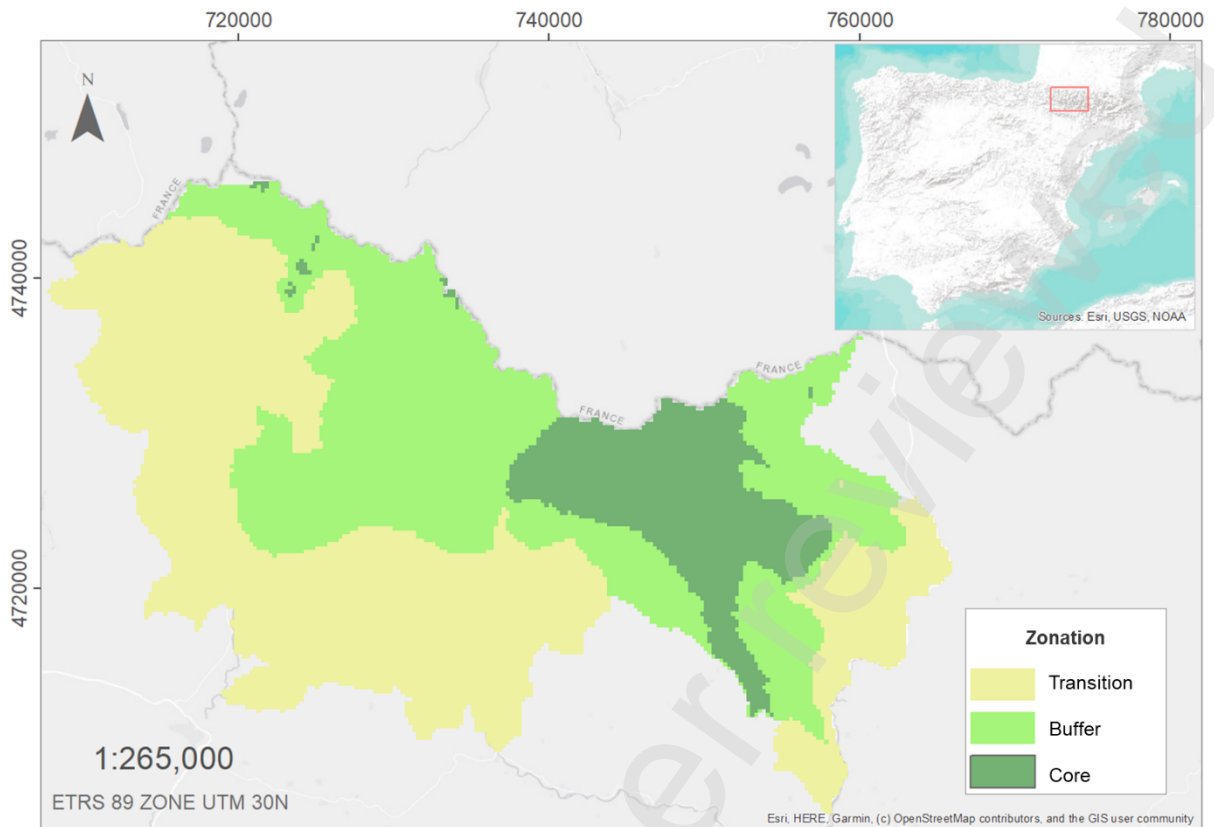
178 This Biosphere Reserve is one of the best representations of the Pyrenees
179 Mountains ecosystems. Typical high mountain landscapes with glaciers and
180 morphologies linked to glaciers give the landscape a significant relief (horns, ridges,
181 cirques, basins with lakes). Besides, the altitudinal variable allows coexisting bioclimatic
182 types, from sub-Mediterranean to mountainous ones. Also, some biogeographic units are
183 established, finding mixed deciduous forests, black pine forests, fir trees, and high
184 mountain pastures (OAPN, 2014). It is important to highlight that there are species of
185 flora and fauna of great importance for conserving biological diversity. In Ordesa-
186 Viñamala, local and regional endemism and thread species such as *Borderea pyrenaica*,
187 *Leontopodium alpinum*, *Cochlearia aragonensis* or *Cypripedium calceolus*. There are
188 also species of birds, such as the Bearded Vulture (*Gypaetus barbatus*) or the Grouse
189 (*Tetrao urogallus*) or amphibians, such as the Ocellated Lizard (*Timon lepidus*).

190

191 Regarding the socio-economic characteristics of the Ordesa-Viñamala Biosphere
192 Reserve, it is noticed that there is a clear example of sustainable development because it
193 is possible to reconcile the use of goods and resources offered by the environment with
194 ecosystem conservation. At present, the main productive activity is livestock to tourism.
195 In addition, the main tourist activities are linked to mountaineering, hiking and mountain
196 sports. Within the extension are 11 municipalities with a total population of 5,639
197 inhabitants.

198

199



200
201
202

Figure 1. Ordesa-Viñamala Biosphere Reserve map

203
204
205

2.2.2. Sierra Nevada

206
207
208
209
210
211

Sierra Nevada Biosphere Reserve (Figure 2) is situated between Almería and Granada, in the Autonomous Community of Andalusia, located 30 km from the coast. It was declared in 1986. Sierra Nevada Biosphere Reserve has 172,238 ha, of which 50% belongs to the core zone, 40% to the buffer zone and 10% to the transition zone. It has a population of 10,760 inhabitants and 60 municipalities.

212
213
214

The managing entity is the Ministry of Environment and Spatial Planning (Calvo, 2019). Within this Biosphere Reserve, other protection figures are shown as follows:

215
216
217
218
219
220
221
222
223

- National Park
- Natural Park
- Natural Monument (Falla de Nigüelas)
- Natura 2000 Network of the European Union
- Special Conservation Area (ZEC)
- Special Protection Area for Birds (ZEPA)
- Place of Community Importance (SCI)
- List of RAMSAR wetlands (Humadales y Peberas de Padul)
- European Charter for Sustainable Tourism (CETS)

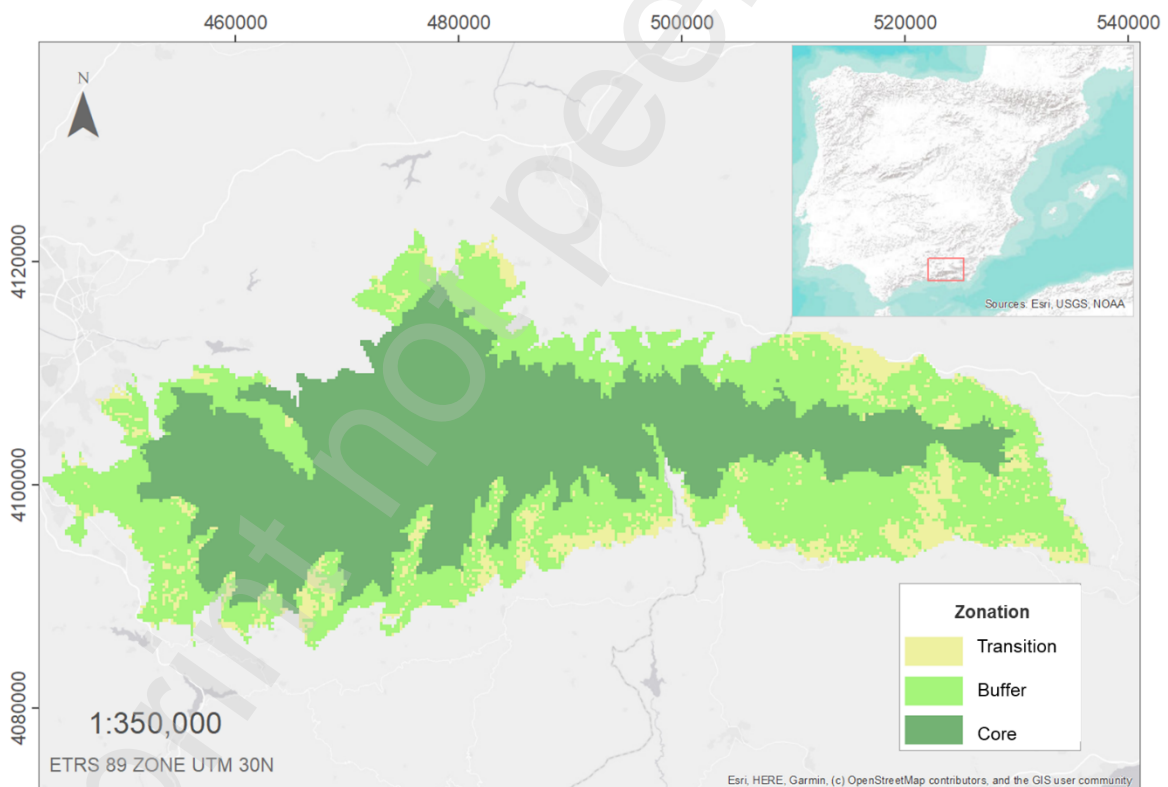
224

225 During the Quaternary, there were glaciers, which were the southernmost in Europe.
226 Consequently, morphologies associated with glacial erosion can be seen, such as cirques,
227 moraines, lagoons, and valleys. Sierra Nevada is the most important point of plant
228 diversity in the western Mediterranean Region, representing almost 30% of the flora of
229 mainland Spain. As for the fauna, there is a great diversity of fauna, highlighting the
230 avifauna.

231

232 The area of socio-economic influence of the Reserve gathers about 98,000
233 inhabitants, of which only 10% live within the limits of the Natural Area. Agricultural
234 and livestock crops represent 11% of land uses within the Reserve, although they have
235 suffered a progressive decline in recent decades. The arboreal species (almond, cherry,
236 olive tree, vineyards) represent 70% of the crops, while the irrigated herbaceous remains
237 30%. It is noteworthy in the industrial sector, meat industries for drying ham, wine,
238 construction materials, renewable energies, and the mineral water bottling plant in
239 Lanjarón. In addition, tourism is a fundamental source of income for the area's
240 development.

241



242

243 *Figure 2. Sierra Nevada Biosphere Reserve map*

244

245

246 **2.2 Data input**

247 Table 1 shows the data input used for the present work.

248

249

Table 1. Input material

Material	Product	Original spatial resolution	Source
NDVI	MOD13Q1	250 m	Google Earth Engine-NASA (last Access: 1 September 2020, Google Earth Engine Datasets Catalog)
Biosphere Reserve zonation layers			MITECO (Spain) (last access: 1 September 2020, http://www.miteco.gob.es/)
Land Cover	Land Cover CCI 2001	300 m	ESA (last access: 1 September 2020, https://www.esa-landcover-cci.org)
Annual mean temperature	MOD11	1 km	Google Earth Engine-NASA (last Access: 1 September 2020, Google Earth Engine Datasets Catalog)
Drought index	MOD16	500 m	Google Earth Engine-NASA (last Access: 1 September 2020, Google Earth Engine Datasets Catalog)
MDT (Digital Terrain Model)	MDT 200	200 m	IGN (Spain) (last access: 1 September 2020, http://centrodedescargas.cnig.es/)

250

251 - MODIS and MOD13q1 product

252

253 The product developed by the MODIS team to obtain the Vegetation Index is
 254 MOD13Q1. This product has been generated from the daily surface reflectance product.
 255 After applying an algorithm (Maximum Value Compositing), it is possible to obtain the
 256 best pixel value and a clean image for 16 days (Huete et al., 1999). Thus, the NDVI data
 257 is generated every 16 days with a 250-pixel-meter resolution. From 2001 to 2016
 258 (inclusive), 368 images were collected (Didan, 2015).

259

260 - CCI Land cover cartography

261 Climate Change Initiative (CCI) is a program dependent on the European Space
 262 Agency (ESA) whose objective is to group and understand all the information collected
 263 from Earth observation by ESA during the last 40 years to contribute to improving the
 264 databases required by the United Nations Framework Convention on Climate Change.
 265 The maps are free and can be downloaded from the ESA website within the CCI program
 266 (ESA, 2017).

267 The land cover layer has been used to observe changes in land use within our time
 268 series (2001-2016). Therefore, the land occupation layers were obtained for 2001 between
 269 2001 and 2016, and the pixels that have remained stable throughout that period are
 270 considered for the present work.

271

272 - Reserves Biosphere layers

273 The Nature Bank of the Ministry has provided these layers for Ecological Transition
274 and the Demographic Challenge (MITECO). It includes the zonation of each Reserve, i.
275 e., the core area, the buffer zone and the outer transition area, at a scale of 1: 50,000.
276

277 - EUNIS Habitat Classification

278 EUNIS Habitat Classification is a European reference classification based on the
279 habitat types listed in Annex I of the European Habitats Directive. Ecosystems are
280 mapped by interpreting different available land cover information according to the
281 European habitat classification. The EUNIS habitat classification covers the whole of
282 Europe's terrestrial and marine areas of Europe and establishes a total of 10 habitat types.
283

284 Generally, the scale used in this classification covers habitats of at least 100 m². However,
285 microhabitats (those below 1 m²) have also been described, and even combinations of
286 mosaics of several individual habitats can cover an area of at least 10 ha. EUNIS habitats
287 are grouped hierarchically into three levels, specifically the higher ones. To make this
288 classification, keys are used, which follow different criteria to categorize each habitat
289 (Davies et al., 2004). In the present work, only terrestrial habitats have been recorded for
290 both Biosphere Reserves, and the classification used is EUNIS level 1 (Table 2).
291

291

292 *Table 2. EUNIS Level 1 habitats used in this study and the number of pixels belonging to each class.*

EUNIS Level 1 categories		Total number of pixels
C	Inland surface waters	140
D	Mires, bogs and fens	6
E	Grasslands and land dominated by forbs, mosses or lichens	8475
F	Heathland, scrub and tundra	15799
G	Woodland, forest and other wooded land	22789
H	Inland unvegetated or sparsely vegetated habitats	4977
I	Arable land and market gardens	1382
J	Constructed, industrial and other artificial habitats	234

293

294 - Average annual temperature of the Earth's surface

295 The annual mean temperature data have been obtained using the MOD11 product of
296 the MODIS sensor for the 2001-2016 time series (both inclusive). The MOD11A1
297 Version 6 product records the Earth's surface temperature daily with a spatial resolution
298 of 1 km. The temperature value per pixel is derived from the MOD11L2 product, which
299 performs daily measurements every 5 minutes using the split-window algorithm (Wan et
300 al., 2015). To facilitate the processing of these data, Google Earth Engine software was
301 used, which allows an average of all the images available to obtain an annual average of
302 temperature for all years from 2001 to 2016, both inclusive.

303 - Annual average evapotranspiration

304 The annual mean evapotranspiration was obtained through the MOD16 product of the
305 MODIS sensor. The spatial resolution of the measurements for this product is 1 km.
306 According to Mu et al. (2011), the MODIS team uses the evapotranspiration algorithm.
307 For this product, the actual and potential evapotranspiration data have been used, during
308 the period 2001-2016, which will be used to calculate the drought index explained later.
309 In the same way as to obtain the mean annual temperature, the evapotranspiration data
310 were processed using Google Earth Engine, thus bringing a yearly mean for each year of
311 the time series under study.

312 - Digital model terrain

313 The elevation of the land has been obtained from the MDT200 cartography elaborated
314 by the National Geographic Institute (IGN) and downloaded through its website (see
315 Table 1).

316 This Digital Terrain Model (DTM) has a mesh pitch of 200 m and has been obtained
317 by interpolation from the terrain class of LIDAR flights of the first coverage of the
318 National Plan for Aerial Orthophotography (PNOA) (García Asensio & Lumbreras
319 Crespo, 1992).

320 **2.3 Data processing**

321

322 **2.3.1 NDVI Pixel Curve Smoothing and NDVI trend categories**

323

324 The annual mean NDVI values per pixel were smoothed using the Gaussian
325 algorithm of the TIMESAT software (Jönsson & Eklundh, 2002) so that no "spikes" of
326 outliers are found and the vegetation time series for the study area can be worked with.
327 In addition, TIMESAT provides a weighting mechanism such that some values in the
328 time series may be more influential than others. Therefore, high weights are assigned for
329 higher quality MODIS shots and low weights for lower quality. In this way, pixels with
330 a lower given weight have a lower weight in the curve fit (Jönsson & Eklundh, 2002).
331 Finally, the mean of each pixel included in the study area was obtained for 2001-2016.
332 Those pixels where the NDVI was less than zero were ignored because they are areas of
333 bare rock or soil without vegetation. Based on the annual mean NDVI values per pixel
334 obtained by smoothing, we analyzed the NDVI trends for the time series (2001-2016).

335

336 Taking as a reference the studies published by Arrogante et al. (2018) and Novillo
337 et al. (2019), Theil-Sen regressions were used to examine the trend. Thus, a positive slope
338 implies a growth in the NDVI value, while a negative slope indicates a decrease in the
339 NDVI value.

340 Subsequently, a Mann-Kendall test was applied to corroborate whether this
341 observed trend is significant. A significant trend between 2001 and 2016 was considered
342 for those pixels where the Mann-Kendall statistic had a p-value of less than 0.1 (90%

343 confidence interval). The Mann-Kendall test and Theil-Sen regression were performed
344 using the Clark Labs TerrSet Earth Trends Modeler (ETM) (Eastman, 2015). ETM is an
345 integrated set of tools for analyzing image time series data associated with remotely
346 sensed Earth observation imagery.

347

348 The result was a raster layer with significant trends generated with Mann-Kendall
349 and another with slopes generated with Theil-Sen. Then, using ArcGIS software, a single
350 layer was obtained with pixels with a significant positive NDVI trend (code = 1), pixels
351 with a significant negative NDVI trend (code = -1) and pixels with no significant trend
352 where the p-value was less than 0.1 (code = 0). Subsequently, the data from the significant
353 trend layer were crossed with the layer presenting the boundaries of the high mountain
354 Biosphere Reserve zones (Sierra Nevada and Ordesa-Viñamala) to obtain the pixels
355 within the study areas.

356

357 In addition, to eliminate the effect that could occur in the time trend of NDVI as
358 a result of variability in land occupation, the raster layer data were crossed with the CCI
359 layer of land occupations of each of the Biosphere Reserves to eliminate the pixels that
360 have undergone changes over the time series and thus obtain pure occupation pixels, i.e.,
361 the same occupation code in 2001, 2008 and 2016 cartography. For this purpose, the
362 different raster maps were combined to assign a single output value to each unique
363 combination of input values.

364

365 **2.3.2 Temperature trend**

366

367 The process of obtaining the time temperature trend was carried out in the same
368 way as that described in the previous section with the NDVI from the annual mean
369 temperature data obtained from the MOD11 product of the MODIS sensor. The only
370 difference is that we did not apply a smooth process.

371

372 **2.3.3 Drought index**

373

374 The actual and potential evapotranspiration data for the Drought Index has been
375 obtained from the MOD16 product from MODIS.

376

377 This drought index has been found by applying the Vapor Stress Index (ESI),
378 which allows the real evapotranspiration to be compared with the potential
379 evapotranspiration obtained by remote sensing using geostationary satellites. This index
380 does not require data on precipitation or subsurface soil characteristics (Anderson et al.,
381 2011). The process to trend the drought index was the same as for the NDVI and
382 temperature trends.

383

384 **2.3.4 Euclidean distance from population centres**

385

386 To examine how the distance to population centers impacts certain factors, the
387 analysis proceeded as follows:

388
389 Selection of Population Centers: The CCI raster layer from 2001 was utilized,
390 specifically identifying population centers with a designated class value (Class = 190).
391 This selection process aimed to pinpoint areas classified as population centers within the
392 raster data.

393
394 Assumption of Stability: It was assumed that population centers would remain
395 intact throughout the studied time series of 16 years. This assumption is grounded in the
396 expectation that urban areas typically maintain their category without significant changes
397 over this relatively short duration.

398
399 Calculation of Euclidean Distance: Using GIS software, the Euclidean distance
400 was computed from the chosen population centers. The Euclidean distance metric
401 measures the straight-line distance from each pixel or point to the nearest population
402 center. This analysis provides insights into the spatial relationship between various
403 locations and their proximity to the closest population center.

404
405 By conducting this analysis, valuable information can be obtained regarding the
406 influence of distance on different aspects or phenomena, such as land use patterns,
407 accessibility, or socio-economic characteristics.

408
409 All the variables were projected to ETRS 1989- UTM Zone 30 North, the official
410 projection in Spanish databases. The layers were resampled to 250 meters pixels, the
411 resolution from the MOD13q1 product, obtaining a total of 53,802 samples, of which
412 32,095 correspond to the Sierra Nevada Reserve, and 21,707 remain within the Ordesa-
413 Viñamala Reserve.

414 415 **2.4 Data analysis**

416
417 First, before performing the statistical analysis of the data, a descriptive study was
418 carried out where the distribution of frequencies between the explanatory variables and
419 the variable to be explained (NDVI trend) was examined. Then, bar graphs were made
420 for categorical variables, and box and whisker plots were used for numerical variables.

421 Statistical tests were performed using R software to test for statistically significant
422 relationships between the different variables. First, the Chi-square test was performed
423 between NDVI trends and categorical variables. Then, the Kruskal-Wallis test (see Table
424 2), analogous to an analysis of variance (ANOVA) but non-parametric, was performed to
425 assess whether there is a significant association between NDVI trends and numerical
426 variables. The Kruskal-Wallis test identifies substantial differences between the median
427 values of the numerical variable of a pair of categories of the variable to be explained
428 (positive, negative or no NDVI trend), so the post-hoc Wilcoxon signed-rank test was
429 carried out to identify which pairs differ.

430
431
432
433
434
435
436
437

Finally, for the categorical variables, a frequency analysis was carried out by creating contingency tables in which the difference in percentage between the observed and expected NDVI was calculated, with both significant positive and negative trends and no trend in the classes of each categorical variable, and graphs were made based on these differences.

Table 3. Input variables and tests performed for each analysis

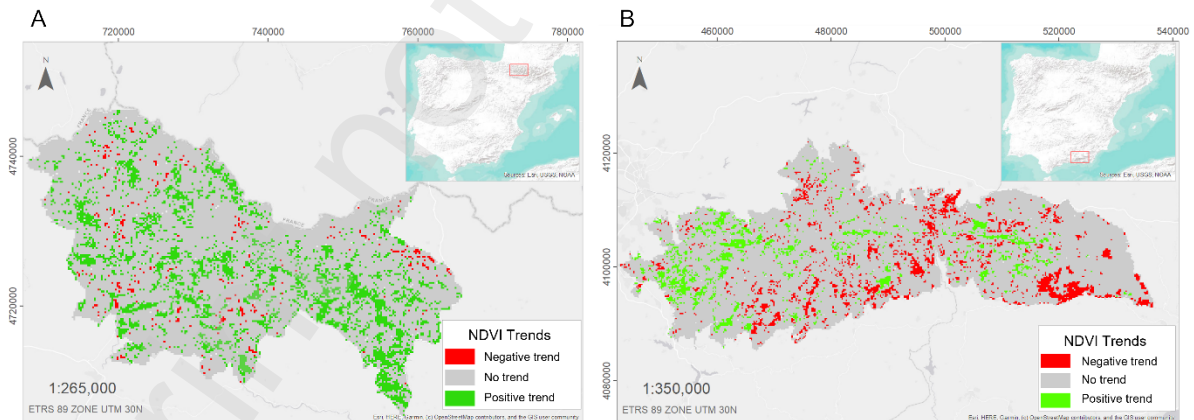
Analysis	Explained variable	Explanatory variables	Test
Analysis 1	NDVI trends	Reserve	Chi-square
Analysis 2	NDVI trends	Reserve zone	Chi-square
Analysis 3	NDVI trends	EUNIS Habitats	Chi-square
Analysis 4	NDVI trends	Annual mean temperature trend	Chi-square
Analysis 5	NDVI trends	Annual mean hydric stress trend	Chi-square
Analysis 6	NDVI trends	Elevation	Kruskal-Wallis
Analysis 7	NDVI trends	Distance to population centres	Kruskal-Wallis

438

439 3. Results

440 3.1. NDVI trends of the reserves

441 Figure 3 shows the geographical distribution of NDVI trends from 2001 to 2016 for each
442 Reserve. In general, significant positive NDVI trends are more frequent, accounting for
443 11.7%, while significant negative trends account for 6%, and there is 82.3% with no trend.
444



445
446
447

Figure 3. NDVI time trends in each Reserve. A. Ordesa-Viñamala Biosphere Reserve; B. Sierra Nevada Biosphere Reserve.

448

449 Figure 4A reveals the percentages of pixels with significant positive, negative and no
450 significant NDVI trends, for the 2001-2016 time series and each Reserve. Ordesa-
451 Viñamala is the Reserve with the highest percentage of pixels with significant trends,
452 20.15%, while the Sierra Nevada has 16.02% of pixels with significant trends. On the
453 other hand, in Ordesa-Viñamala, significant positive NDVI trends predominate at

454 18.47%, while significant negative trends represent only 1.68%. On the contrary, Sierra
 455 Nevada shows higher values in the negative NDVI trend with 9.02%.

456

457 Furthermore, when comparing the values of the difference between the observed and the
 458 expected number of pixels with positive, negative and no NDVI trend (Figure 4B), we
 459 observe that in the Sierra Nevada, the values of significant negative NDVI trend are more
 460 frequent than expected, while in Ordesa-Viñamala the frequency of negative NDVI trend
 461 is lower than expected; on the contrary, there are more positive values of NDVI trend
 462 than expected. In the latter case, it should be noted that the differences between observed
 463 and expected values are higher than 50% for positive and negative NDVI trends.

464

465

466

467

468

469

470

471

472

473

474

475

476

477

478

479

480

481

482

483

484

485

486

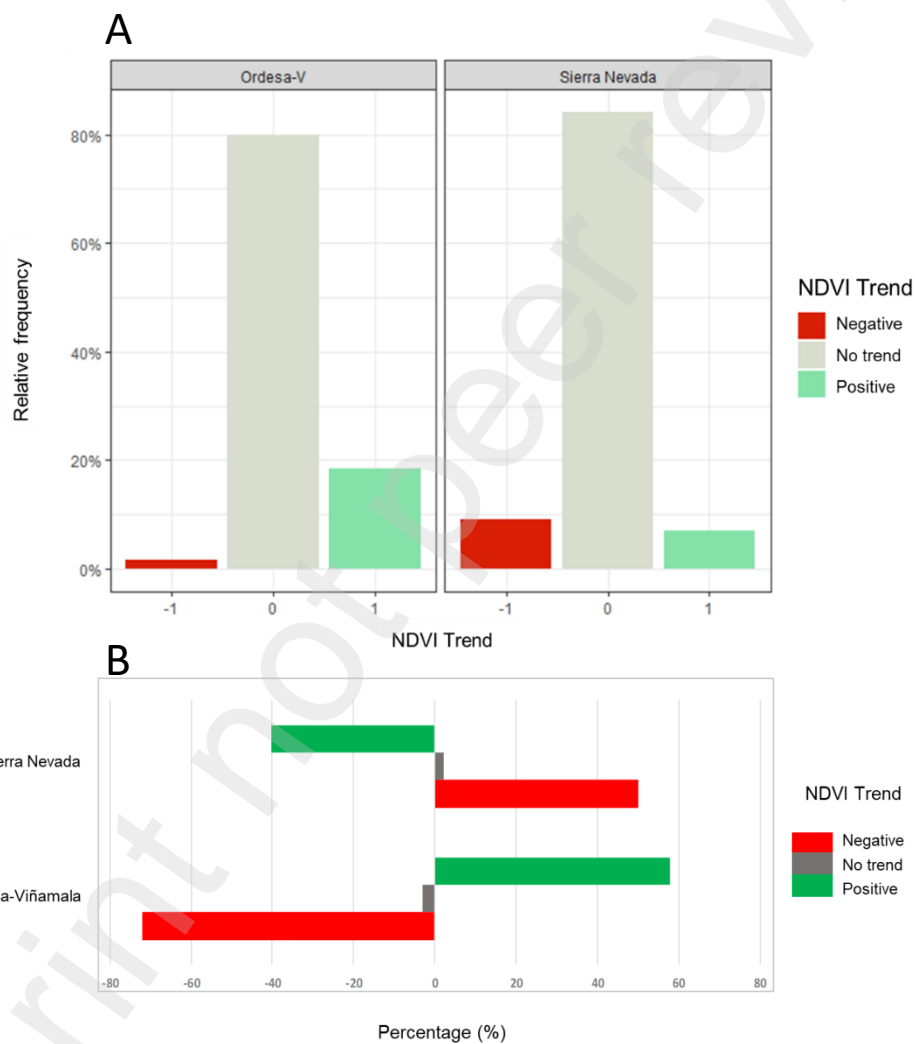
487

488

489

490

491



490 *Figure 4. A. Proportions of each category of NDVI trend in both reserves and B. Deviation of actual trend values*
 491 *from expected trend values in both reserves.*

492 3.2. Time trends of NDVI by important variables

493 3.2.1. Trend of NDVI according to reserve zonation

494 Figure 5.1 shows how Ordesa-Viñamala has a higher percentage of pixels with

495 positive than negative NDVI trends. In addition, the number of pixels with significant
 496 negative NDVI trends is lower in the core, slightly higher in the buffer zone and
 497 somewhat higher in the transition zone (0.48%, 1.48% and 2.14%, respectively). On the
 498 other hand, Sierra Nevada has a higher percentage of pixels with a negative NDVI trend
 499 in the transition zone than in the core zone, increasing from 7.72% in the core zone to
 500 19.42% in the transition zone. On the contrary, a decrease in the percentage of pixels with
 501 a positive trend is observed as we move towards the transition zone, with values of 8.32%
 502 in the core zone and 2.82% in the transition zone.

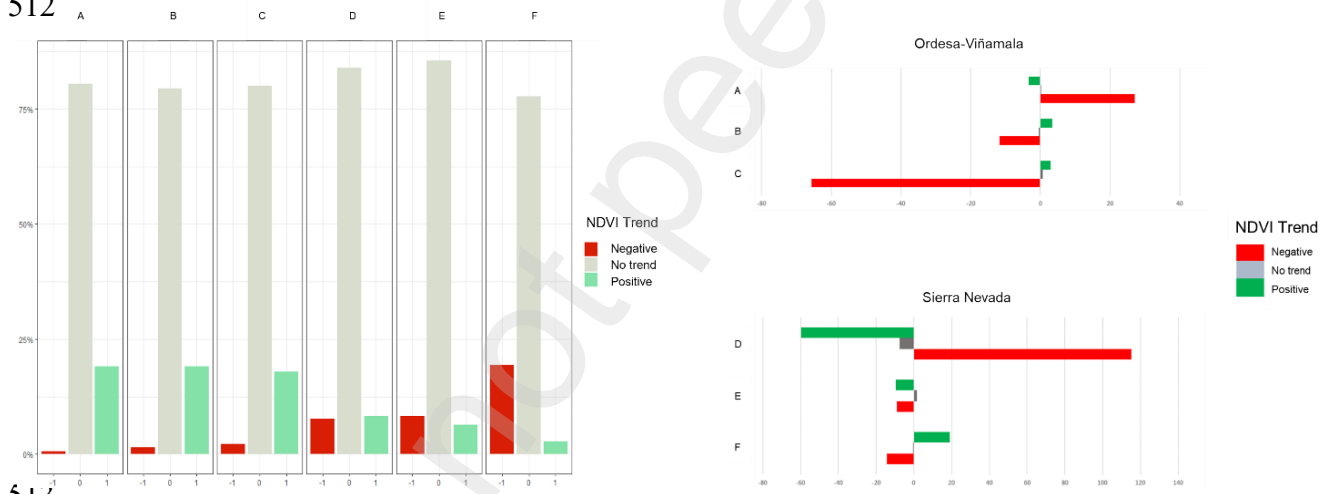
503

504 In addition, to verify this pattern, Figure 5.2 shows that in Ordesa-Viñamala, there
 505 is a lower percentage of negative NDVI trend than expected (less than 60%) for the core
 506 zone. The core zone of Sierra Nevada also reveals a lower rate of negative NDVI trend
 507 than expected (less than 10%) and a higher positive NDVI trend than expected (19%).
 508 Both reserves have more negative NDVI trends and less than-expected positive trends for
 509 the transition zone. Still, the high values observed in the Sierra Nevada stand out, with a
 510 115% higher-than-expected negative trend and a 60% lower-than-expected positive trend.

511

1

512



513

514 *Figure 5. 1. Proportion of each category of NDVI trend in each zone of both reserves. A. Core area of Ordesa-*
 515 *Viñamala; B. Buffer zone of Ordesa-Viñamala; C. Transition area of Ordesa-Viñamala; D. Core area of Sierra*
 516 *Nevada; E. Buffer zone of Sierra Nevada; F. Transition zone of Sierra Nevada. 2. Deviation of actual trend values*
 517 *from expected trend values in each zone of both reserves. A. Transition area of Ordesa-Viñamala; B. Buffer zone of*
 518 *Ordesa-Viñamala; C. Transition area of Ordesa-Viñamala; D. Core area of Sierra Nevada; E. Buffer zone of Sierra*
 519 *Nevada; F. Core area of Sierra Nevada.*

520 3.2.2. Trend of NDVI according to the trend of temperature

521

522 No significant decreasing trends in surface temperature over 2001-2016 were
 523 found for both reserves. However, Sierra Nevada is the Reserve with the highest
 524 increasing temperature trend for the time series, with 55% of pixels with such a trend. On
 525 the other hand, Ordesa-Viñamala has only 13% of pixels with a rising temperature trend.
 526 In both reserves, there is a decrease in the positive NDVI trend from areas without

527 significant temperature variation trends to regions with a significantly increasing
 528 temperature trend, as seen in Figure 6.1.

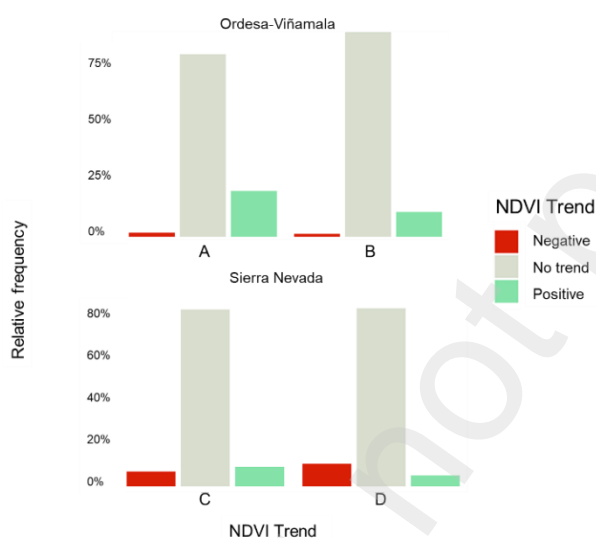
529

530 Figure 6.2 shows the difference between the observed and expected NDVI time
 531 trend pixels as a function of the temperature time trend within each Reserve. The
 532 difference in NDVI trend values between pixels with increasing temperature trends and
 533 pixels where no significant temperature trends were found is remarkable. The same
 534 pattern is observed in both reserves for growing temperature trends, where the positive
 535 NDVI trend values are lower than expected. Furthermore, there is a higher percentage of
 536 negative NDVI trends in the Sierra Nevada than expected. Notably, positive trends
 537 increased more than expected, with no significant temperature trends recorded. On the
 538 contrary, in Ordesa-Viñamala, positive and negative NDVI trends are lower than
 539 expected when there is a positive temperature trend. In contrast, when there are no
 540 significant temperature trends, there are higher positive NDVI trends than expected,
 541 although in a lower proportion than in the Sierra Nevada.

542

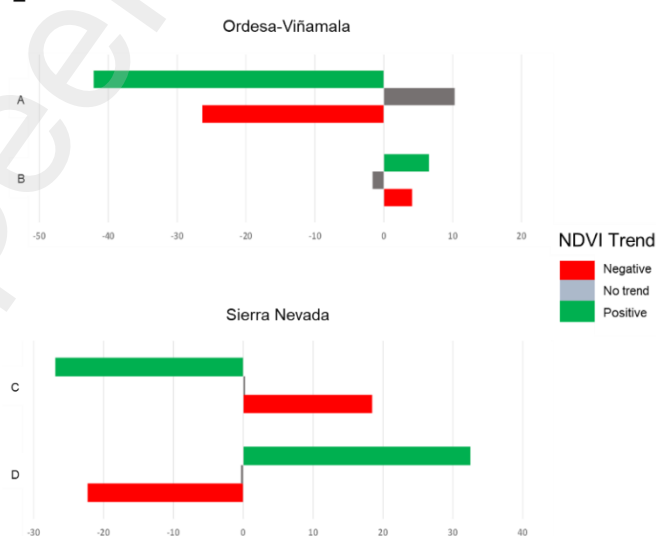
543

544 1



545
546
547
548
549
550
551
552
553
554
555
556
557

2



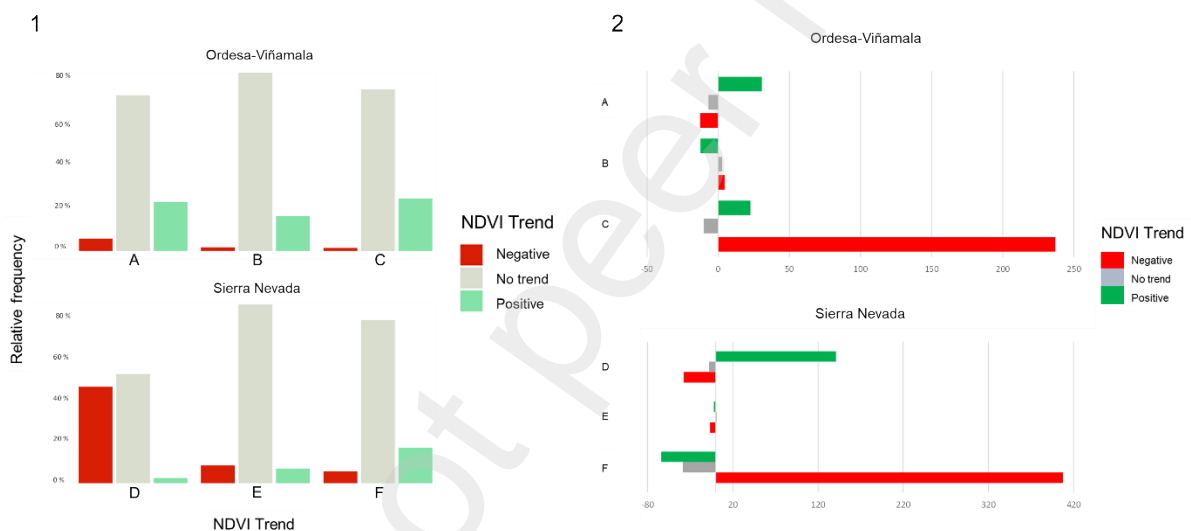
558 *Figure 6. 1. Proportion of each category of NDVI trend according to the time temperature trend in both reserves. A.*
 559 *No significant temperature trend in Ordesa-Viñamala Reserve; B. Significant positive temperature trend in Ordesa-*
 560 *Viñamala Reserve; C. No significant temperature trend in Sierra Nevada Reserve; D. Significant positive temperature*
 561 *trend in Sierra-Nevada Reserve. 2. Deviation of actual temperature trend values from expected temperature trend*
 562 *values in both reserves. A. Significant positive temperature trend in Ordesa-Viñamala Reserve; B. No significant*
 563 *temperature trend in Ordesa-Viñamala Reserve; C. Significant positive temperature trend in Sierra-Nevada Reserve;*
 564 *D. No significant temperature trend in Sierra Nevada Reserve.*

565 3.2.3 Trend of NDVI according to the time trend of hydric stress

566 Both reserves have similarities regarding the percentages of NDVI trend pixels as
 567 a function of significant water stress trends for 2001-2016 (Figure 7.1). For pixels with a
 568 positive water stress trend, i.e. lower water stress, there is a higher percentage of pixels
 569 with a positive NDVI trend, around 20% positive NDVI trend in both reserves. While the
 570 highest rates of negative NDVI trends in both reserves are found in areas with a negative

571 trend of water stress, i.e., greater water stress, Ordesa-Viñamala has 6% of negative NDVI
 572 trends. At the same time, in the Sierra Nevada, we observe a percentage of 46% of
 573 negative NDVI trends. Therefore, there are differences between reserves: Ordesa-
 574 Viñamala has higher percentages of significant positive NDVI trends, while Sierra
 575 Nevada has higher rates of negative NDVI trends.
 576

577 On the other hand, for negative water stress trends, a higher percentage of negative
 578 NDVI trends than expected was observed in both reserves (Figure 7.2). However, the
 579 rates were more elevated in the Sierra Nevada. For areas with positive water stress trends,
 580 higher positive NDVI trends and lower negative NDVI trends than expected were
 581 recorded in both reserves. Still, here, however, the positive NDVI trends were higher in
 582 the Sierra Nevada.
 583
 584



585
 586
 587
 588 *Figure 7. 1. Time NDVI trends according to time hydric stress trends in both reserves. A. Significant negative hydric*
 589 *stress trend in Ordesa-Viñamala Reserve; B. No significant hydric stress trend in Ordesa-Viñamala Reserve; C.*
 590 *Significant positive hydric stress trend in Ordesa-Viñamala Reserve; D. Significant negative hydric stress trend in*
 591 *Sierra Nevada Reserve; E. No significant hydric stress trend in Sierra Nevada Reserve; F. Significant positive hydric*
 592 *stress trend in Sierra Nevada Reserve. 2. Deviation of actual hydric stress trend values from expected hydric stress*
 593 *trend values in both reserves. A. Significant negative hydric stress trend in Ordesa-Viñamala Reserve; B. No significant*
 594 *hydric stress trend in Ordesa-Viñamala Reserve; C. Significant positive hydric stress trend in Ordesa-Viñamala*
 595 *Reserve; D. Significant negative hydric stress trend in Sierra Nevada Reserve; E. No significant hydric stress trend in*
 596 *Sierra Nevada Reserve; F. Significant positive hydric stress trend in Sierra Nevada Reserve.*

597 3.2.4. Time trend of NDVI according to EUNIS habitats

598
 599 Figure 13 shows significant differences in the NDVI trends according to habitat
 600 type in each Reserve. In Ordesa-Viñamala, positive NDVI trends are higher than negative
 601 ones for all habitat categories. The highest percentages of positive NDVI trends were
 602 observed in the habitats corresponding to grasslands and lands dominated by herbaceous,
 603 mosses or lichens (E) and heathlands, shrublands and tundra (F), with 23.34% and

604 20.65%, respectively. These were followed by habitats C and G, which are inland surface
 605 waters and woodland, forest and other wooded land, respectively. On the contrary,
 606 agricultural habitats (I) and urban, industrial or artificial areas (J) showed the highest
 607 percentage of negative trends. The most frequent habitats, i.e. the ones that concentrate
 608 the highest number of pixels, are wooded areas (G), with 43% of the total number of
 609 pixels, and habitat E, with 25%.

610

611 In the Sierra Nevada, there are more negative significant trends than positive
 612 trends in five of the eight identified habitats. Notably, 100% of the pixels belonging to
 613 wetlands (D) show negative NDVI trends for the period 2001-2016. This habitat is
 614 followed in the percentage of negative NDVI trends by habitats I and J, representing
 615 agricultural habitats and built and artificial areas, as in Ordesa-Viñamala. On the other
 616 hand, the habitats with the highest percentage of positive NDVI trends are habitats E and
 617 F, as in Ordesa-Viñamala. Finally, the most frequent habitats are G, as in Ordesa-
 618 Viñamala, and F, which correspond to woodland and heathland, scrub and tundra,
 619 representing 41.92% and 40.92% of the total number of pixels.

620

621

622

623

624

625

626

627

628

629

630

631

632

633

634

635

636

637

638

639

640

641

642

643

644

645

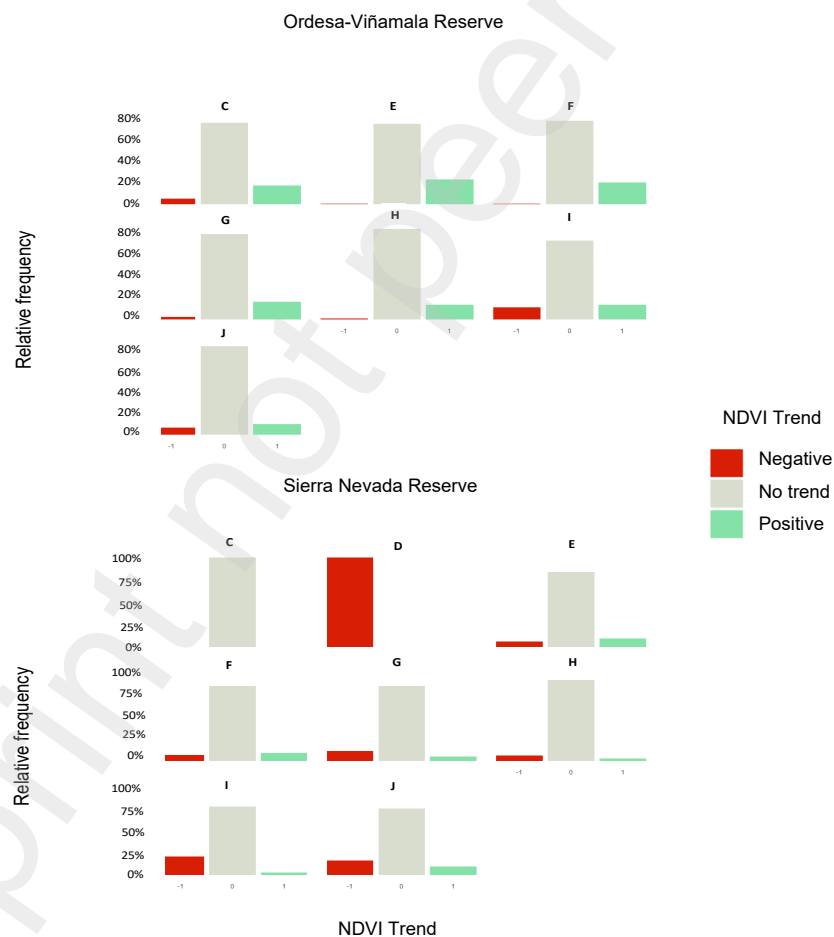


Figure 8. NDVI time trends distribution of frequencies according to EUNIS Habitats categories in both reserves.

646 3.2.5. Time trend of NDVI according to altitude and population centres

647

648 The results in Figure 8 have shown that positive NDVI trends occur at higher
649 altitudes than negative NDVI trends in both reserves. The variability of the negative
650 NDVI trends in Ordesa-Viñamala is lower than those recorded for positive NDVI trends,
651 which have a wider range of dispersion. These negative NDVI trends occur at lower
652 altitudes than the positive trends. In the Sierra Nevada, the negative NDVI trends show a
653 greater altitude variability, but they are still observed at lower altitudes than the positive
654 NDVI trends.

655 Concerning distance to population centres, the results have confirmed that the
656 distance is generally less in the Sierra Nevada than in Ordesa-Viñamala. In Ordesa-
657 Viñamala, negative NDVI trends are found in areas closer to population centres, and
658 positive trends are greater at greater distances from population centres. In contrast, in the
659 Sierra Nevada, positive NDVI trends are found at shorter distances from population
660 centres than negative trends (Figure 8).

661

662

663

664

665

666

667

668

669

670

671

672

673

674

675

676

677

678

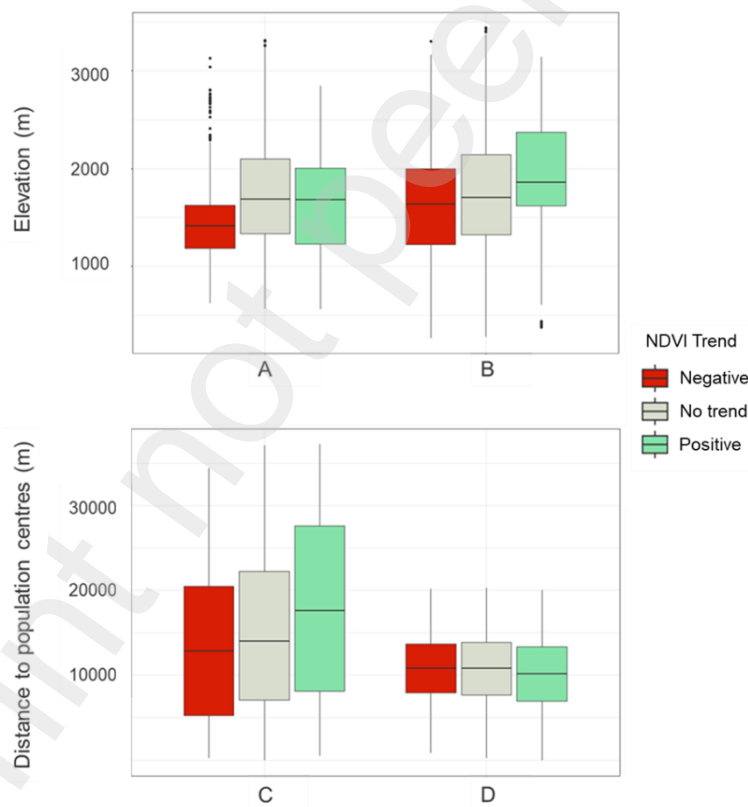
679

680

681

682

683



682 *Figure 9. NDVI time trends distribution of frequencies according to elevation and distance to population centres in*
683 *each zone of both reserves.*

684 3.3. Results of the Chi-square and Kruskal-Wallis analysis

685

686 Tables 3 and 4 show the results obtained for the statistical tests applied for
 687 categorical (Table 3) and numerical variables (Table 4).

688
 689

Table 4. Chi-square test results. SN: Sierra Nevada Reserve; OV: Ordesa-Viñamala Reserve

Test number	Explained variable	Explanatory variable	Biosphere Reserve	p value	Degrees of freedom	Chi-square value
1	NDVI trend	Reserves	Todas	<0.05	2	2599.68
2	NDVI trend	Reserve zone	SN	<0.05	4	40.77
3	NDVI trend	Reserve zone	OV	<0.05	4	527.40
4	NDVI trend	Annual mean temperature trend	SN	< 0.05	2	306.06
5	NDVI trend	Annual mean temperature trend	OV	<0.05	2	142.55
6	NDVI trend	Annual mean hydric stress trend	SN	<0.05	4	2682.00
7	NDVI trend	Annual mean hydric stress trend	OV	<0.05	4	143.50

690
 691
 692

Table 5. Kruskal-Wallis test results. SN: Sierra Nevada Reserve; OV: Ordesa-Viñamala Reserve

Test number	Explained variable	Explanatory variable	Biosphere Reserve	p value	Degrees of freedom	H value	Post-hoc Wilcoxon Ranks
8	NDVI trend	Elevation	OV	< 0.05	2	172.42	All pairs are differentiated
9	NDVI trend	Elevation	SN	< 0.05	2	432.21	All pairs are differentiated
10	NDVI trend	Distance to population centres	OV	< 0.05	2	302.25	All pairs are differentiated
11	NDVI trend	Distance to population centres	SN	< 0.05	2	38.37	0 is not differentiated from -1

693
 694
 695
 696
 697
 698
 699
 700
 701

A statistical association was found between NDVI trends and all the categorical variables analyzed in each Reserve. For the numerical variables, significant differences were found for all variables except for the slope variable in Ordesa-Viñamala, where no significant differences were found for the negative and positive NDVI trend categories. Also, there were no significant differences between areas with no NDVI trend and regions with a negative NDVI trend regarding the distance to population centres variable.

702 4. Discussion

703
 704
 705
 706
 707
 708
 709
 710
 711
 712
 713

The statistical results show a significant association between NDVI trends and the reserves studied. In general, it has been observed that significant positive NDVI trends are more frequent than negative trends in high mountain Biosphere Reserves for the period from 2001 to 2016. These results coincide with those obtained by Arrogante et al. (2018) for the same period and in the Pyrenees and Southern Baetic mountain regions, where the Reserves under study are located. These results have also been observed in continental Spain (Novillo et al., 2019). This general increase in photosynthetic activity also occurs globally (Nemani et al., 2003). According to different studies, this is related to the rise in mean annual temperature, favouring plant development, especially in spring and summer (Myneni et al., 1997).

714

715 NDVI trend results for the two Reserves have followed different patterns. In
716 Ordesa-Viñamala, more positive NDVI trends have been observed than negative ones.
717 Moreover, the positive trends recorded are higher than expected, while the negative trends
718 are lower than expected. However, this pattern is reversed in the Sierra Nevada, where
719 negative NDVI trends are predominant and higher than expected, while positive trends
720 occur less frequently. These differences can be explained by climatic factors, such as
721 temperature, precipitation and radiation, although anthropogenic factors may also play a
722 role (Liu et al., 2015).

723

724 According to Papagiannopoulou et al. (2017), the limiting factors on vegetation
725 in the north of the Peninsula are temperature and solar radiation received. In the south,
726 the limiting factor is precipitation. Khorchani et al. (2018) state that when energy is the
727 limiting factor, an increase in temperature results in positive NDVI trends; however when
728 precipitation is the limiting factor, this increase in temperature causes a decrease in the
729 NDVI trend.

730

731 In Ordesa-Viñamala, vegetation growth has a high seasonality, the limiting factor
732 of this growth being temperature, especially in winter, but the availability of water and
733 solar energy received, especially in summer, allows vegetation development (Alcaraz-
734 Segura, Cabello, et al., 2009). The increase in mean annual temperatures is causing this
735 seasonality to be lower, which could be favouring vegetation development. However, the
736 climatic conditions of Sierra Nevada mean that the main limiting factor is precipitation
737 (Alcaraz-Segura, Cabello, et al., 2009). In addition, rising temperatures are causing an
738 increase in the frequency of summer droughts (Sergio M Vicente-Serrano et al., 2014).
739 On the other hand, altitude may also be causing the temperature to be a limiting factor in
740 winter.

741

742 No significant decreasing surface temperature trends have been found from 2001
743 to 2016. This fact makes sense in the current global warming scenario (IPCC, 2014). A
744 significant association was found between surface temperature trends and NDVI trends,
745 and it was observed that Sierra Nevada Reserve has the highest percentage of increasing
746 temperature trends. Besides, when the temperature trend has grown for both reserves, the
747 positive NDVI trends have decreased compared to areas where no significant temperature
748 trends were found. However, when comparing the negative NDVI trends, there are
749 differences between the reserves: in the Sierra Nevada, the negative NDVI trends increase
750 from areas with no significant temperature trend to places where there has been an
751 increase in temperature, while in Ordesa-Viñamala, there is no such increase in the
752 negative NDVI trends, there is even a slight decrease. These differences highlight the
753 vulnerability of the Mediterranean region to rising temperatures, which could explain why
754 we find more negative and less positive NDVI trends than expected in the Sierra Nevada
755 and the opposite in Ordesa-Viñamala (García-Ruiz et al., 2011).

756

757 Concerning the above, the increase in temperatures and the decrease in
758 precipitation has led to a rise in the severity of droughts, especially in the Mediterranean
759 area (García-Ruiz et al., 2011; Sergio M. Vicente-Serrano et al., 2014), so it would be
760 expected that in areas with negative trends of water stress, we would observe higher
761 negative and lower positive trends of NDVI than expected. Again, these trends would be
762 accentuated in the Sierra Nevada.

763

764 The results demonstrate this, although the common trend in both reserves is to
765 find more negative NDVI trends than expected for areas with an increase in water stress
766 and more positive trends than expected where there has been a decrease in water stress.
767 Sierra Nevada seems to be more affected by this variable, which could indicate that water
768 stress is a more limiting factor for vegetation development in the Mediterranean bioregion
769 than in the alpine (Alcaraz-Segura, Cabello, et al., 2009). It is noteworthy that in the areas
770 of Sierra Nevada where there has been a decrease in water stress, higher positive NDVI
771 trends than expected have been found than for the same regions of Ordesa-Viñamala,
772 which seems to explain how, when neither water availability nor temperature is limiting
773 factors, vegetation growth is favoured.

774

775 Regarding EUNIS habitats, in Ordesa-Viñamala, grasslands and shrublands are
776 the habitats with a higher percentage of positive NDVI trends since they could be
777 benefiting from the increase in temperatures, which would allow attenuating the effect of
778 low temperatures in winter, as explained Alcaraz-Segura et al. (2009) and Khorchani et
779 al. (2018). On the contrary, the habitats with the most NDVI negative trends are
780 agricultural habitats and built and constructed areas, the same as in Sierra Nevada. These
781 trends could be explained by the abandonment of traditional practices on farmland, which,
782 together with unfavourable climatic cycles, hinder vegetation development (Valladares
783 et al., 2004).

784

785 In the Sierra Nevada, the whole wetlands area shows negative NDVI trends, which
786 indicates how temperature increases and lack of precipitation affect the vegetation
787 associated with these habitat types. Agricultural and built areas also host more negative
788 trends than positive ones, as in Ordesa-Viñamala. The latter could be because agricultural
789 areas are suffering a widespread impact due to climate change (Anderson et al., 2020;
790 Pathak et al., 2018). These habitats are followed in the percentage of negative trends by
791 woodlands and unvegetated lands. On the contrary, the habitats with the highest
792 percentage of positive NDVI trends are shrublands and grasslands. The latter mentioned
793 could indicate that land abandonment is triggering the development of woody vegetation
794 in these areas (Lasanta et al., 2017).

795

796 On the other hand, the most extensive habitats in the Sierra Nevada, woodlands
797 and forests, which are composed mostly of coniferous forests and evergreen broadleaved
798 forests, and heathland, shrubland and tundra are the most extensive ones in Ordesa-
799 Viñamala as well. It indicates how, despite being high mountain Reserves and having the

800 same habitat types, there are significant differences depending on the bioregion where the
801 Reserve is located.

802

803 In this sense, habitats in the Ordesa-Viñamala Biosphere Reserve, located in the
804 north of Spain, have a much more favourable evolution than the habitats of the Sierra
805 Nevada located in the south. Ordesa-Viñamala is not in arid conditions of the
806 Mediterranean zone, so the progressive increase in temperatures favours vegetation
807 growth and, therefore, generates positive NDVI trends (Martínez-Vilalta et al., 2008). In
808 this lower latitude, they endure conditions of higher average temperatures and water stress
809 (Lamprecht et al., 2021) and, therefore, potential aridification in the face of climate
810 change (Oliva et al., 2011; Ramos-Román et al., 2018).

811

812 In addition, it has to be taken into account that Biosphere reserves represent a
813 dynamic conservation concept, different from other protected areas, whose main purpose
814 is conservation. In addition to preserving ecosystems, biosphere reserves seek to maintain
815 harmony between human action and the natural environment and, therefore, the cultural
816 landscapes created by anthropic activity (Vericad Corominas & Balcells Rocamora,
817 1981).

818

819 However, within the biosphere reserve concept, there are different approaches or
820 objectives. It should be noted that the Ordesa-Viñamala biosphere reserve has a great
821 environmental value and a more purely conservation focus (González González, 2020)
822 than the Sierra Nevada biosphere reserve, which hosts recreational activities, like tourism,
823 occupying a considerable area (Moreno-Llorca et al., 2020). It is possible that the
824 preservation of the state of the ecosystems in Ordesa-Viñamala, together with its
825 latitudinal situation and, thus, its climatic characteristics, has led to a significantly higher
826 proportion of positive NDVI trends and a lower proportion of negative trends than in the
827 Sierra Nevada reserve in general, which translates into a better situation of tree stands.

828

829 On the other hand, a significant statistical association was found between the
830 protection zones of each Reserve. The same trend has been observed in both reserves, i.e.
831 negative NDVI trends increase, and positive trends decrease as we move away from the
832 core area. However, both trends remain more stable in Ordesa-Viñamala. These results
833 make sense, since the objective of these protected areas, according to UNESCO's MaB
834 Programme, is precisely to offer greater protection in the core area, to allow the
835 development of environmentally compatible activities in the buffer zone and to allow the
836 development of activities that favour socio-economic development in the buffer zone
837 (Batisse, 1982). Furthermore, this is evident because both reserves also show the same
838 differences in observed and expected NDVI trends. The results obtained by Arrogante et
839 al. (2018) for the mountainous area of Sierra Nevada show more negative and less
840 positive trends than expected in protected areas and suggest that the configuration and
841 management in this area could also explain the NDVI trends found.

842

843 Regarding the orographic variables studied, the statistical results show significant
differences between altitude and NDVI trends in both reserves. These differences are due

844 to the distribution of vegetation, which is partly dependent on topography (Peco et al.,
845 1998). Positive NDVI trends occur at higher altitudes than negative ones in both reserves.
846 These results coincide with those obtained by Pauli et al. (2012), where it is stated that
847 the increase in temperatures induces the altitudinal ascent of vegetation in search of more
848 temperate zones and greater water availability. Regarding slope, no significant
849 differences were found between negative and positive NDVI trends in Ordesa-Viñamala,
850 i.e. slope does not seem to be associated with finding significant positive or negative
851 NDVI trends in the area. On the other hand, in Sierra Nevada, significant associations
852 were found between NDVI trends and slope, although the proximity of the medians seems
853 to indicate that this variable does not have a great influence on NDVI trends.

854

855 NDVI trends were significantly associated with distance to population centres in
856 Ordesa-Viñamala. In this Reserve, negative NDVI trends have been observed at distances
857 closer to population centres, and positive trends have been recorded in more distant areas.
858 These results are logical since the greater the distance from population centres, the less
859 anthropogenic activity capable of causing disturbances to the vegetation can be expected.
860 In the Sierra Nevada, no significant differences were found between the distance to
861 population centres and areas with negative and no NDVI trends. However, there are
862 significant differences between negative and positive NDVI trends. Notably, positive
863 trends were observed in the Sierra Nevada at a shorter distance from population centres
864 than negative trends. The latter mentioned could be explained by the fact that the Sierra
865 Nevada has 60 municipalities within the reserve boundary and, in addition, its elongated
866 and irregular shape means that the central zone is not very isolated from the outer zones
867 (Shafer, 2008), which would mean that the municipalities are closer to the core and
868 transition zones, which are precisely where the highest NDVI trends have been recorded.

869

870 In summary, global warming is increasing the vulnerability of these habitats,
871 especially in peninsular Spain. Studying and understanding environmental changes in
872 these protected areas is crucial for taking appropriate conservation measures and
873 preserving their services. Biosphere Reserves are internationally protected areas that seek
874 to protect biodiversity and promote sustainable development. Evaluating these protected
875 areas is essential to verify environmental changes and establish priorities in their
876 management. In addition, these studies, like the one shown here, help identify threats and
877 pressures they face and sites requiring greater attention and conservation efforts. Finally,
878 by studying the temporal trends of the Normalized Difference Vegetation Index (NDVI)
879 in a spatially continuous manner and their relationship with different factors such as
880 temperature, water stress, altitude, slope, habitat type, biogeographic region, zonation,
881 and distance to population centres, valuable information can be obtained about the factors
882 influencing the health and resilience of mountain ecosystems. The latter allows for a
883 better understanding of environmental processes and helps establish more effective
884 conservation strategies.

885 **5. Conclusions**

886 This study assessed NDVI trends for 2001-2016 in the high mountain Biosphere
887 Reserves: Sierra Nevada and Ordesa-Viñamala. The main conclusions are the following:

888 - Generally, there are more significant positive NDVI trends in the high mountain
889 Biosphere Reserves analyzed as a whole. These results are similar to other studies carried
890 out both in Spain and globally.

891 - Significant differences in NDVI trends were found between the two reserves. However,
892 the Ordesa-Viñamala Biosphere Reserve has shown higher significant positive NDVI
893 trends and fewer negative trends, while this pattern is reversed in the Sierra Nevada.

894 - The analysis of the variables has allowed us to explain the differences found between
895 the reserves, and it has been observed that temperature and water stress affect Sierra
896 Nevada Reserve to a greater extent.

897 - The NDVI trends analyzed according to the zonation of each Reserve show the same
898 pattern in both reserves, with higher positive trends around the core and lower positive
899 trends around the transition zone.

900 - Altitude is highly related to vegetation development in both reserves.

901 - The significant association between the distance to population centres and the NDVI
902 trend shows that anthropogenic factors limit vegetation development.

903 - Remote sensing and GIS are essential for this kind of study as they provide access to
904 spatial data, enable long-term monitoring, facilitate spatial analysis, and support decision-
905 making. By analyzing trends in NDVI and integrating environmental factors such as
906 distance to population centers, altitude, and habitat type, valuable insights are gained
907 regarding the health and resilience of mountain ecosystems. In addition, these tools help
908 understand the effects of global warming, identify conservation priorities, and plan
909 sustainable development within Biosphere Reserves. Therefore, remote sensing and GIS
910 are crucial in assessing environmental changes and informing effective management
911 strategies for high mountain areas.

912 **Funding**

913 The preparation of this paper was supported by Rey Juan Carlos University through the
914 project called ECOFORIA from the IMPULSO call of research projects (code:
915 2023/00004/005). It was also funded through the project called INFOLANDYN, financed
916 by the Ministerio de Ciencia, innovación y Universidades (PID2020-115509RB-100). It
917 also is supported by the Comunidad de Madrid funding under the research contracts
918 program (INVESTIGO-Grants URJC-AI-57 y URJC-AI-18). Furthermore, this research

919 was also funded by the Ministerio de Ciencia, Innovación y Universidades (grant no.
920 PRE2019-089208).

921

922 **Acknowledgements**

923

924 This research was conducted within the URJC project (ECOFORIA) framework from the
925 IMPULSO call of projects (code: 2023/00004/005). Furthermore, this paper's preparation
926 was also supported through the project called INFOLANDYN, financed by the Ministerio
927 de Ciencia, innovación y Universidades (PID2020-115509RB-100). Furthermore,
928 Ariadna Álvarez-Ripado was supported by a research work contract from the Comunidad
929 de Madrid- subsidy from the INVESTIGO Program (URJC-AI-57). In addition, Dina
930 Osuna was also supported by a research work contract from the Comunidad de Madrid-
931 a grant from the INVESTIGO Program (URJC-AI-18). Finally, Fátima Arrogante-Funes
932 was supported by a predoctoral scholarship (FPI) from the Spanish Ministry of Science,
933 Innovation and Universities (PRE2019-089208).

934

935 **References**

- 936 Aguayo, P. M., & CIREN, G. (2013). Índices de vegetación.
937 Alcaraz-Segura, D., Cabello, J., Paruelo, J. M., & Delibes, M. (2009). Use of
938 descriptors of ecosystem functioning for monitoring a national park network: a
939 remote sensing approach. *Environmental Management*, 43, 38-48.
940 Alcaraz-Segura, D., Liras, E., Tabik, S., Paruelo, J., & Cabello, J. (2009). Evaluating
941 the consistency of the 1982-1999 NDVI trends in the Iberian Peninsula across
942 four time-series derived from the AVHRR sensor: LTDR, GIMMS, FASIR, and
943 PAL-II. *Sensors*, 10(2), 1291-1314.
944 Allen, C. D., Macalady, A. K., Chenchouni, H., Bachelet, D., McDowell, N., Venetier,
945 M., . . . Hogg, E. T. (2010). A global overview of drought and heat-induced tree
946 mortality reveals emerging climate change risks for forests. *Forest ecology and
947 management*, 259(4), 660-684.
948 Anderson, M. C., Hain, C., Wardlow, B., Pimstein, A., Mecikalski, J. R., & Kustas, W.
949 P. (2011). Evaluation of drought indices based on thermal remote sensing of
950 evapotranspiration over the continental United States. *Journal of Climate*, 24(8),
951 2025-2044.
952 Anderson, R., Bayer, P. E., & Edwards, D. (2020). Climate change and the need for
953 agricultural adaptation. *Current opinion in plant biology*, 56, 197-202.
954 Arrogante-Funes, P., Novillo, C., & Romero-Calcerrada, R. (2018). Monitoring NDVI
955 Inter-Annual Behavior in Mountain Areas of Mainland Spain (2001-2016).
956 *Sustainability*, 10(12), 4363.
957 Arrogante-Funes, P., Novillo, C. J., & Romero-Calcerrada, R. (2018). Monitoring
958 NDVI Inter-Annual Behavior in Mountain Areas of Mainland Spain (2001-
959 2016). *Sustainability*, 10(12), Article 4363. <https://doi.org/10.3390/su10124363>
960 Batisse, M. (1982). The biosphere reserve: a tool for environmental conservation and
961 management. *Environmental Conservation*, 9(2), 101-111.
962 Bradley, B. A., & Mustard, J. F. (2008). Comparison of phenology trends by land cover
963 class: a case study in the Great Basin, USA. *Global Change Biology*, 14(2), 334-
964 346.
965 Busetto, L., Colombo, R., Migliavacca, M., Cremonese, E., Meroni, M., Galvagno, M., .
966 . . . Pari, E. (2010). Remote sensing of larch phenological cycle and analysis of

- 967 relationships with climate in the Alpine region. *Global change biology*, 16(9),
968 2504-2517.
- 969 Chuvieco Salinero, E. (2002). Teledetección ambiental La observación de la Tierra
970 desde el Espacio.
- 971 Didan, K. (2015). MOD13Q1 MODIS/Terra Vegetation Indices 16-Day L3 Global
972 250m SIN Grid V006. distributed by NASA EOSDIS LP DAAC,
973 <https://doi.org/10.5067/MODIS/MOD13Q1.006>.
- 974 Dudley, N. (2008). *Guidelines for applying protected area management categories*.
975 IUCN.
- 976 Eastman, J. R. (2015). TerrSet. Clark University, Worcester. Google Scholar.
- 977 Fensholt, R., Langanke, T., Rasmussen, K., Reenberg, A., Prince, S. D., Tucker, C., . . .
978 Eastman, R. (2012). Greenness in semi-arid areas across the globe 1981-2007:
979 An Earth Observing Satellite based analysis of trends and drivers. *Remote*
980 *sensing of environment*, 121, 144-158.
- 981 Field, C. B., Barros, V., Stocker, T. F., & Dahe, Q. (2012). *Managing the risks of*
982 *extreme events and disasters to advance climate change adaptation: special*
983 *report of the intergovernmental panel on climate change*. Cambridge University
984 Press.
- 985 García Asensio, L., & Lumberras Crespo, J. J. (1992). El modelo digital del terreno
986 MDT200 del Instituto Geográfico Nacional: descripción general y resultados.
987 *Mapping*, 1(3), 38-42.
- 988 García-Ruiz, J. M., López-Moreno, J. I., Vicente-Serrano, S. M., Lasanta-Martínez, T.,
989 & Beguería, S. (2011). Mediterranean water resources in a global change
990 scenario. *Earth-Science Reviews*, 105(3-4), 121-139.
- 991 González González, M. J. (2020). Analysis, Systemization of the Impacts of Planning
992 on the Territory: Applied to the Ordesa National Park. *Land*, 9(12), 527.
- 993 Hmimina, G., Dufrêne, E., Pontailier, J. Y., Delpierre, N., Aubinet, M., Caquet, B., . . .
994 Granier, A. (2013). Evaluation of the potential of MODIS satellite data to
995 predict vegetation phenology in different biomes: An investigation using
996 ground-based NDVI measurements. *Remote Sensing of Environment*, 132, 145-
997 158.
- 998 Huete, A., Justice, C., & Van Leeuwen, W. (1999). MODIS vegetation index (MOD
999 13). Version 3. Algorithm theoretical basis document. *Greenbelt MD NASA*
1000 *Goddard Space Flight Cent, Greenbelt, MD, USA*, 7.
- 1001 Jönsson, P., & Eklundh, L. (2002). Seasonality extraction by function fitting to time-
1002 series of satellite sensor data. *IEEE Transactions on Geoscience and Remote*
1003 *Sensing*, 40(8), 1824-1832.
- 1004 Khorchani, M., Vicente-Serrano, S. M., Azorin-Molina, C., Garcia, M., Martin-
1005 Hernandez, N., Peña-Gallardo, M., . . . Domínguez-Castro, F. (2018). Trends in
1006 LST over the peninsular Spain as derived from the AVHRR imagery data.
1007 *Global and Planetary Change*, 166, 75-93.
- 1008 Lamprecht, A., Pauli, H., Fernández Calzado, M. R., Lorite, J., Molero Mesa, J.,
1009 Steinbauer, K., & Winkler, M. (2021). Changes in plant diversity in a water-
1010 limited and isolated high-mountain range (Sierra Nevada, Spain). *Alpine Botany*,
1011 131(1), 27-39.
- 1012 Lasanta, T., Arnáez, J., Pascual, N., Ruiz-Flaño, P., Errea, M., & Lana-Renault, N.
1013 (2017). Space-time process and drivers of land abandonment in Europe. *Catena*,
1014 149, 810-823.

- 1015 Liu, Y. Y., Van Dijk, A. I., De Jeu, R. A., Canadell, J. G., McCabe, M. F., Evans, J. P.,
1016 & Wang, G. (2015). Recent reversal in loss of global terrestrial biomass. *Nature*
1017 *Climate Change*, 5(5), 470-474.
- 1018 Martínez Vega, J., Martín, M. P., Díaz Montejo, J. M., López Vizoso, J. M., & Muñoz
1019 Recio, F. J. (2010). Guía didáctica de teledetección y medio ambiente.
- 1020 Martínez-Vilalta, J., López, B. C., Adell, N., Badiella, L., & Ninyerola, M. (2008).
1021 Twentieth century increase of Scots pine radial growth in NE Spain shows
1022 strong climate interactions. *Global change biology*, 14(12), 2868-2881.
- 1023 Moreno-Llorca, R., Vaz, A., Herrero, J., Millares, A., Bonet-García, F., & Alcaraz-
1024 Segura, D. (2020). Multi-scale evolution of ecosystem services' supply in Sierra
1025 Nevada (Spain): An assessment over the last half-century. *Ecosystem Services*,
1026 46, 101204.
- 1027 Myneni, R. B., Keeling, C. D., Tucker, C. J., Asrar, G., & Nemani, R. R. (1997).
1028 Increased plant growth in the northern high latitudes from 1981 to 1991. *Nature*,
1029 386(6626), 698.
- 1030 Nemani, R. R., Keeling, C. D., Hashimoto, H., Jolly, W. M., Piper, S. C., Tucker, C. J.,
1031 . . . Running, S. W. (2003). Climate-driven increases in global terrestrial net
1032 primary production from 1982 to 1999. *science*, 300(5625), 1560-1563.
- 1033 Novillo, C. J., Arrogante-Funes, P., & Romero-Calcerrada, R. (2019). Recent NDVI
1034 Trends in Mainland Spain: Land-Cover and Phytoclimatic-Type Implications.
1035 *ISPRS International Journal of Geo-Information*, 8(1), 43.
- 1036 Oliva, M., Schulte, L., & Ortiz, A. G. (2011). The role of aridification in constraining
1037 the elevation range of Holocene solifluction processes and associated landforms
1038 in the periglacial belt of the Sierra Nevada (Southern Spain). *Earth Surface*
1039 *Processes and Landforms*, 36(10), 1279-1291.
- 1040 Papagiannopoulou, C., Miralles, D. G., Dorigo, W. A., Verhoest, N. E. C., Depoorter,
1041 M., & Waegeman, W. (2017). Vegetation anomalies caused by antecedent
1042 precipitation in most of the world. *Environmental Research Letters*, 12(7),
1043 074016.
- 1044 Pathak, T. B., Maskey, M. L., Dahlberg, J. A., Kearns, F., Bali, K. M., & Zaccaria, D.
1045 (2018). Climate change trends and impacts on California agriculture: a detailed
1046 review. *Agronomy*, 8(3), 25.
- 1047 Pauli, H., Gottfried, M., Dullinger, S., Abdaladze, O., Akhalkatsi, M., Alonso, J. L. B., .
1048 . . Calzado, R. F. (2012). Recent plant diversity changes on Europe's mountain
1049 summits. *Science*, 336(6079), 353-355.
- 1050 Peco, B., Espigares, T., & Levassor, C. (1998). Trends and fluctuations in species
1051 abundance and richness in Mediterranean annual pastures. *Applied Vegetation*
1052 *Science*, 1(1), 21-28.
- 1053 Pepin, N., Bradley, R. S., Diaz, H. F., BaraÑ«r, M., Caceres, E. B., Forsythe, N., . . .
1054 Liu, X. D. (2015). Elevation-dependent warming in mountain regions of the
1055 world. *Nature Climate Change*, 5(5), 424.
- 1056 Pettorelli, N., Vik, J. O., Mysterud, A., Gaillard, J.-M., Tucker, C. J., & Stenseth, N. C.
1057 (2005). Using the satellite-derived NDVI to assess ecological responses to
1058 environmental change. *Trends in ecology & evolution*, 20(9), 503-510.
- 1059 Ramos-Román, M. J., Jiménez-Moreno, G., Camuera, J., García-Alix, A., Anderson, R.
1060 S., Jiménez-Espejo, F. J., & Carrión, J. S. (2018). Holocene climate aridification
1061 trend and human impact interrupted by millennial-and centennial-scale climate
1062 fluctuations from a new sedimentary record from Padul (Sierra Nevada, southern
1063 Iberian Peninsula). *Climate of the Past*, 14(1), 117-137.

- 1064 Rodà, F., Ibáñez, J., & Gracia, C. (2003). L'estat dels boscos. *L'estat del Medi Ambient*
1065 *a Catalunya. Generalitat de Catalunya.*
- 1066 Running, S. W., Nemani, R. R., Heinsch, F. A., Zhao, M., Reeves, M., & Hashimoto, H.
1067 (2004). A continuous satellite-derived measure of global terrestrial primary
1068 production. *Bioscience*, 54(6), 547-560.
- 1069 Schröter, D., Cramer, W., Leemans, R., Prentice, C. I., Araújo, M. B., Arnell, N. W., . . .
1070 Gracia, C. A. (2005). Ecosystem service supply and vulnerability to global
1071 change in Europe. *science*.
- 1072 Shafer, C. L. (2008). Terrestrial nature reserve design at the urban/rural interface.
1073 *Urban ecology: an international perspective on the interaction between humans*
1074 *and nature*, 715-737.
- 1075 Tucker, C. J., & Sellers, P. J. (1986). Satellite remote sensing of primary production.
1076 *International journal of remote sensing*, 7(11), 1395-1416.
- 1077 Valladares, F., CAMARERO, J., PULIDO, F., & GIL-PELEGRÍN, E. (2004). El
1078 bosque mediterráneo, un sistema antropizado y cambiante. *Ecología del bosque*
1079 *mediterraneo en un mundo cambiante. Organismo Autónomo de Parques*
1080 *Nacionales. Ministerio de Medio Ambiente, Madrid.*
- 1081 Valladares, F., Peñuelas, J., & de Luis Calabuig, E. (2005). Impactos sobre los
1082 ecosistemas terrestres. *Evaluación preliminar de los impactos en España por*
1083 *efecto del cambio climático*, 67.
- 1084 Vericad Corominas, J. R., & Balcells Rocamora, E. (1981). La Reserva de la Biosfera
1085 Ordesa-Viñamala y el interés socio-económico de su estudio.
- 1086 Vicente-Serrano, S. M., Lopez-Moreno, J.-I., Beguería, S., Lorenzo-Lacruz, J.,
1087 Sanchez-Lorenzo, A., García-Ruiz, J. M., . . . Trigo, R. (2014). Evidence of
1088 increasing drought severity caused by temperature rise in southern Europe.
1089 *Environmental Research Letters*, 9(4), 044001.
- 1090 Vicente-Serrano, S. M., Lopez-Moreno, J. I., Beguería, S., Lorenzo-Lacruz, J., Sanchez-
1091 Lorenzo, A., García-Ruiz, J. M., . . . Trigo, R. (2014). Evidence of increasing
1092 drought severity caused by temperature rise in southern Europe. *Environmental*
1093 *Research Letters*, 9(4), 044001.
- 1094 Wan, Z., Hook, S., & Hulley, G. (2015). MOD11A1 MODIS/Terra Land Surface
1095 Temperature/Emissivity Daily L3 Global 1km SIN Grid V006. 2015, Distributed
1096 by NASA EOSDIS Land Processes DAAC. In.

1097

1098

1099

1100

1101

1102

1103

1104