

LOTVS: a global collection of permanent vegetation plots

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Abstract

Analysing temporal patterns in plant communities is extremely important to quantify the extent and the consequences of ecological changes, especially considering the current biodiversity crisis. Long-term data collected through the regular sampling of permanent plots represent the most accurate resource to study ecological succession, analyse the stability of a community over time and understand the mechanisms driving vegetation change. We hereby present the LOnG-Term Vegetation Sampling (LOTVS) initiative, a global collection of vegetation time-series derived from the regular monitoring of plant species in permanent plots. With 79 datasets from five continents and 7789 vegetation time-series monitored for at least six years and mostly on an annual basis, LOTVS possibly represents the largest collection of temporally fine-grained vegetation time-series derived from permanent plots and made accessible to the research community. As such, it has an outstanding potential to support innovative research in the fields of vegetation science, plant ecology and temporal ecology.

Keywords: ecoinformatics; ecological succession; ecosystem stability; global scale; permanent plots; plant communities; plant diversity; temporal analysis; time-series; vegetation.

1. Background

Anthropogenic changes are severely impacting ecosystems (Bradshaw et al. 2021). The rate of species loss has now exceeded background extinction rates (Pimm et al. 2014), leading many scientists to claim a sixth mass extinction (Pereira et al. 2012; Ceballos et al. 2015). At the same time, a considerable proportion of natural habitats has been lost (Convention on Biological Diversity 2020) and a number of ecosystem functions and services are seriously at risk (IPBES 2019).

The analysis of time-based patterns in biological communities, especially when focused on primary producers like plants, represents an opportunity to quantify the extent and the consequences of such changes in biodiversity (Dornelas et al. 2014; Gonzalez et al. 2016; Blowes et al. 2019). This research field has potential for: i) unravelling the mechanisms that drive and maintain biodiversity over time (Jones et al. 2017; Hillebrand et al. 2018); ii) shedding light on how external drivers (e.g. global changes) affect community dynamics in natural habitats (Bernhardt-Römermann et al. 2015; Newbold et al. 2015) and iii) assessing relationships between community stability over time and the delivery of ecosystem services (Isbell et al. 2018). Reliable answers to these questions can only be provided by drawing upon ecological data collected at several points in time using consistent sampling procedures. For plant communities, long-term data collected by regularly sampling permanent plots probably constitute the most precise approach to detect temporal changes at the local scale (Bakker et al. 1996; Damgaard 2019; de Bello et al. 2020). First, due to their geographical position being kept “fixed” in the

21 field, permanent plots prevent relocation bias, i.e. the error derived from trying to find the original plot
22 location. This bias is inherent in vegetation resurveys (Verheyen et al. 2018). Second, the repeated
23 collection of vegetation data from permanent plots provides broad benefits to our understanding of
24 vegetation change, including the means to track detailed successional trajectories, monitor species
25 interactions over time and assess the stability of the community as a whole. For this reason, permanent
26 plots have been listed among the six most important developments in vegetation science over the past
27 three decades (Chytrý et al. 2019).

28 In recent decades, vegetation science has benefited from the development and maintenance of large
29 vegetation databases (Dengler et al. 2011). Historical vegetation relevés performed by early vegetation
30 ecologists, together with recent vegetation-plot data stemming from regional, but also national or
31 continental research and survey projects, have been carefully assembled and digitally archived in the
32 context of centralized initiatives (Chytrý et al. 2016; Wiser 2016; Bruelheide et al. 2019; Sabatini et al.
33 2021). Such global collections of vegetation-plot data are essential to investigate macroecological
34 patterns and provide spatially meaningful answers to global issues, i.e. to effectively perform global-
35 scale biodiversity research. In this context, a comparable effort specifically aimed at assembling and
36 maintaining global databases built on time-series of vegetation data is urgently needed to lay a common
37 ground for future studies focusing on i) providing global estimates of changes in plant diversity trends
38 over time; ii) monitoring the conservation status of natural habitats over time or iii) assessing the
39 stability of ecosystem functions and services. To the best of our knowledge, the BioTIME initiative
40 (Dornelas et al. 2018) represents the most important global collection of biodiversity time-series so far,
41 including abundance records measured in species assemblages belonging to the marine, freshwater and
42 terrestrial environments. Yet, the powerful spatial representation of BioTIME has limitations that
43 include: i) an often limited length and/or periodicity of the time-series, which particularly affects
44 vegetation data (29 datasets with at least 6 data points); ii) a poor focus on vegetation and terrestrial
45 plant biodiversity (96 datasets, corresponding to about 27% of the whole database). Given that a high
46 number of ecosystem functions and services strongly depend on plants (Maestre et al. 2012; van der
47 Plas 2019), we deem it crucial for the fields of vegetation science and ecology to be able to rely on a
48 consistent and standardized collection of datasets including high-quality time-series measured at regular
49 intervals and specific to plant communities.

50 Based on these premises, we hereby present the LOnG-Term Vegetation Sampling (LOTVS) initiative,
51 a growing global collection of vegetation time-series derived from the regular (mostly, annual)
52 monitoring of plant species in permanent plots. By promoting the use, and supporting the visibility of
53 high-quality temporal data collected using permanent plots, LOTVS ultimately aims to provide the tools
54 to ask relevant ecological questions across a number of taxa, ecosystems and regions. The LOTVS
55 collection provides a platform for aggregating the currently disconnected datasets sampled around the

56 world based on permanent plots. As such, researchers are welcome to contribute to and, based on a
57 scientific proposal (see section 3.2.), use the available collection of data.

58 2. Description of LOTVS

59 As of December 2021, LOTVS encompasses 79 datasets (Fig. 1) for a total of 7789 vegetation time-
60 series, collected using permanent plots that were monitored for a minimum of six, and a maximum of
61 99 years (first quartile: 10; mean: 17.5; median: 16; third quartile: 23 years; see Fig 2). The vast majority
62 of LOTVS time-series has a fine-grained temporal resolution: measurements in permanent plots were
63 taken on 10% to 100% of the temporal interval (with 100% meaning that plots were sampled at least
64 annually; first quartile: 82.8%; mean: 87.4%; median: 100%; third quartile: 100%). A description of the
65 single datasets can be found in Valencia et al. (2020; Supplementary Material) and is now available on
66 Zenodo as online metadata (<https://doi.org/10.5281/zenodo.5807378>; see section 3.3). At present,
67 LOTVS includes vegetation time-series specifically focused on herbaceous species and shrubs mostly
68 belonging to grassland habitats, followed by mixed vegetation types (i.e. savannas, shrub steppes,
69 degraded stages of heathlands), shrublands (including heathlands), forest understoreys and wetlands
70 (mostly salt marshes; Fig. 4A; see Box 1). We note that, in this context, mixed vegetation types consist
71 of communities where grasses and shrubs co-exist in a mosaic landscape (naturally, or as the result of
72 anthropogenic disturbance processes) or are just co-dominant. Forest plots exclusively monitoring long-
73 term changes in tree species are, at the moment, excluded from LOTVS. The LOTVS collection
74 contains data from five continents (Fig. 1; an interactive map can be also accessed [here](#) and through the
75 proposal template available at <https://doi.org/10.5281/zenodo.5807378>), although Europe and North
76 America are so far the most represented areas. While Europe is the leading continent in terms of datasets
77 (38 out of 79), North America hosts the majority of plots (almost 70%, distributed across 30 datasets).

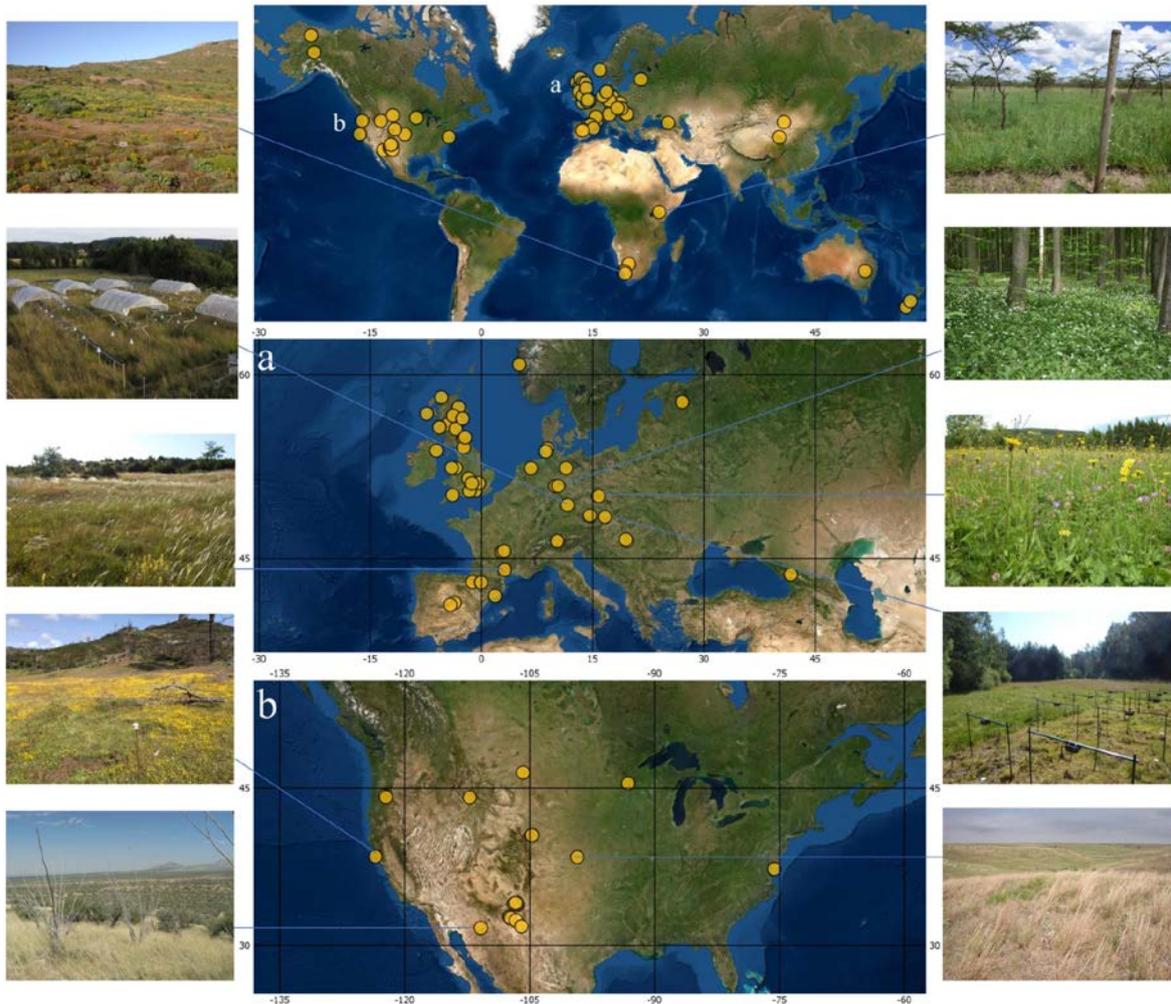
Habitat types

Forest	Vegetated land with a tree canopy cover > 10% and covering an area of more than 0.5 ha.
Forest understorey	The vegetative layer of a forest, consisting of tree seedlings, shrubs and herbaceous vegetation growing below the forest canopy.
Grassland	Open vegetation dominated by graminoids and forbs, characterized by a low cover of woody species (i.e. trees and shrubs).
Heathland	Evergreen formation mostly composed of heathers, i.e. dwarf shrubs of the <i>Ericaceae</i> family
Salt marshes	Coastal wetlands dominated by salt- and flood-tolerant vegetation (mostly grasses, rushes and sedges).

Savannas	Vegetation typical of tropical and subtropical seasonally dry climates, normally characterized by an open canopy of scattered trees and an understory dominated by grasses, often intermingled with large grassland patches.
Shrublands	Open vegetation dominated by shrubs or small (< 5 m tall) trees, often characterized by a single canopy layer.
Wetlands	Permanent or temporary areas of marsh, fen, peatland or water characterised by water that is static or flowing, fresh, brackish or salt.
Sampling method	
Biomass	Clipping, drying and weighing the aboveground phytomass present in a sampling unit (either as a whole or separately for each species).
Cover	Visual estimation of plant species cover in percentage (%) or through the use of cover-abundance classes.
Frequency	Estimation of species abundance based on recording the percentage of sub-units occupied by each species in a certain sampling unit. This can be done, e.g. by counting the number of contacts between each species and a pin.

78 **Box 1.** Working definitions of the main habitat types and sampling methods mentioned. Habitat
79 definitions were adapted from the Convention of Biological Diversity website
80 (<https://www.cbd.int/forest/definitions.shtml> and <https://www.cbd.int/drylands/definitions>), from
81 Goldstein & DellaSala (2020) and from the Ramsar Convention website
82 (<https://www.ramsar.org/about/the-convention-on-wetlands-and-its-mission>).

83 Datasets included in LOTVS span a wide climatic gradient, their mean annual temperature ranging
84 between -11.5 and 20.1°C, and their mean annual precipitation between 140 and 2592 mm (source:
85 WordClim 2; Fick & Hijmans 2017). As such, they are mostly included in the temperate seasonal forest,
86 temperate grassland/desert and in the woodland/shrubland biomes (sensu Whittaker 1975; see Fig 3).



87
 88 **Fig. 1.** Map showing the geographical location of the 79 datasets included in LOTVS. A more detailed
 89 view is given for areas featuring a high density of datasets: a) Europe; b) North America. ESRI World
 90 Satellite Imagery was used as the base map. For a subset of sites, representative vegetation types are
 91 shown. Photos were taken at: (left, starting from the top): Soebatsfontein (South Africa), Bayreuth
 92 (Germany), Roquefort-sur-Soulzon (France), McLaughlin Natural Reserve (California, USA), Santa
 93 Rita Experimental Range (Arizona, USA); (right, starting from the top): Laikipia (Kenya), Göttinger
 94 Wald (Germany), Krkonoše Mountains (Czech Republic), Ohrazení (Czech Republic), Hays (Kansas,
 95 USA).

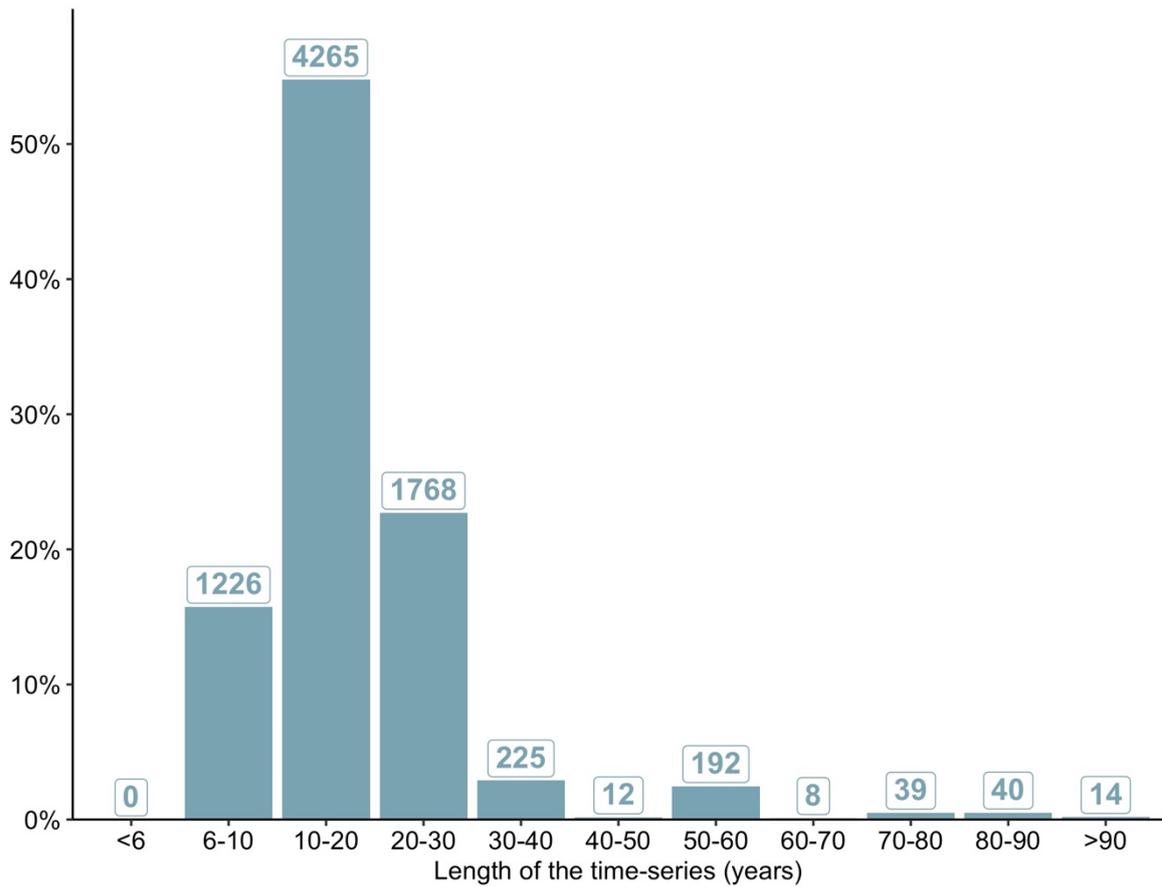
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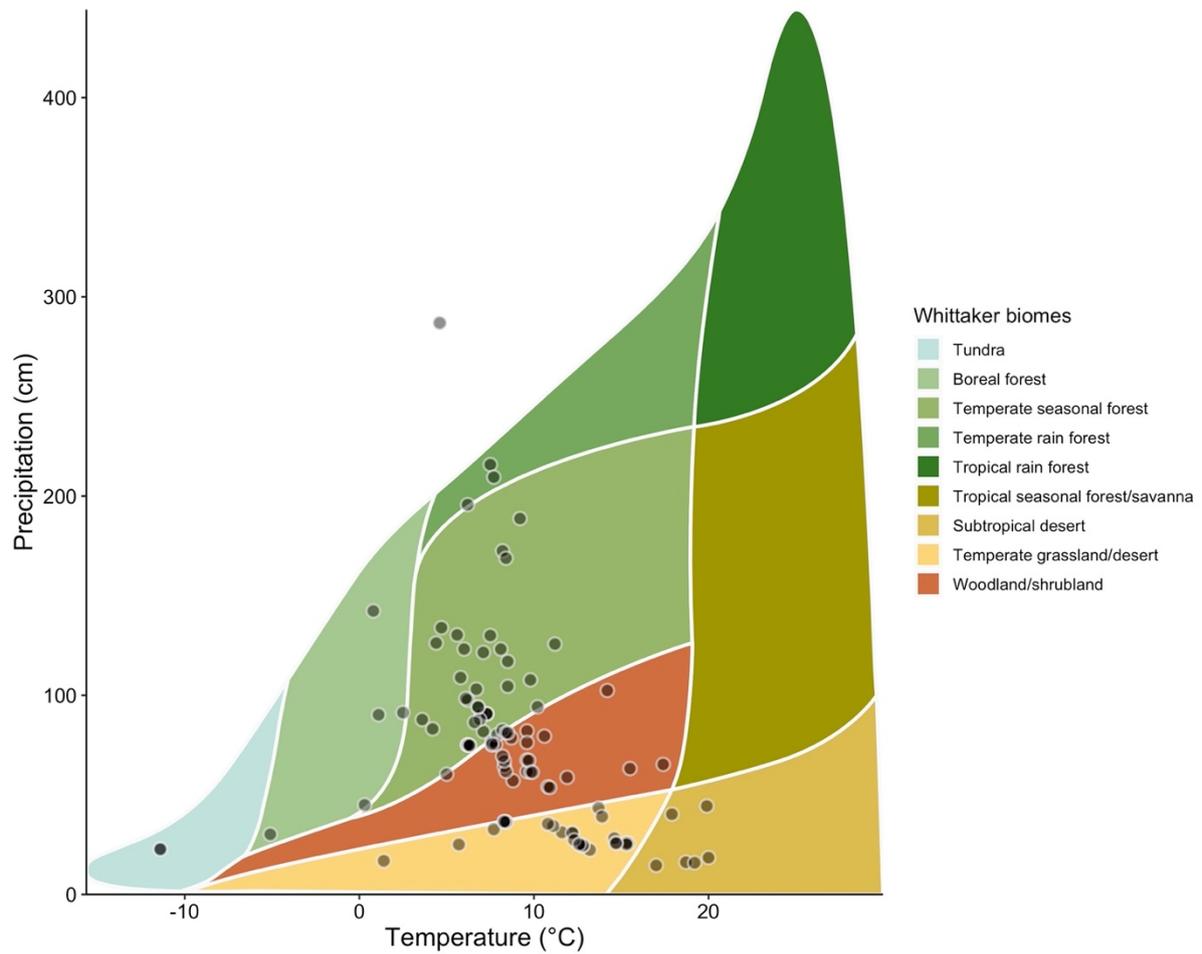
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103 **Fig. 2.** Distribution of the duration (in years) of the time-series included in LOTVS. The number of time-
104 series within each class is reported above the bars.

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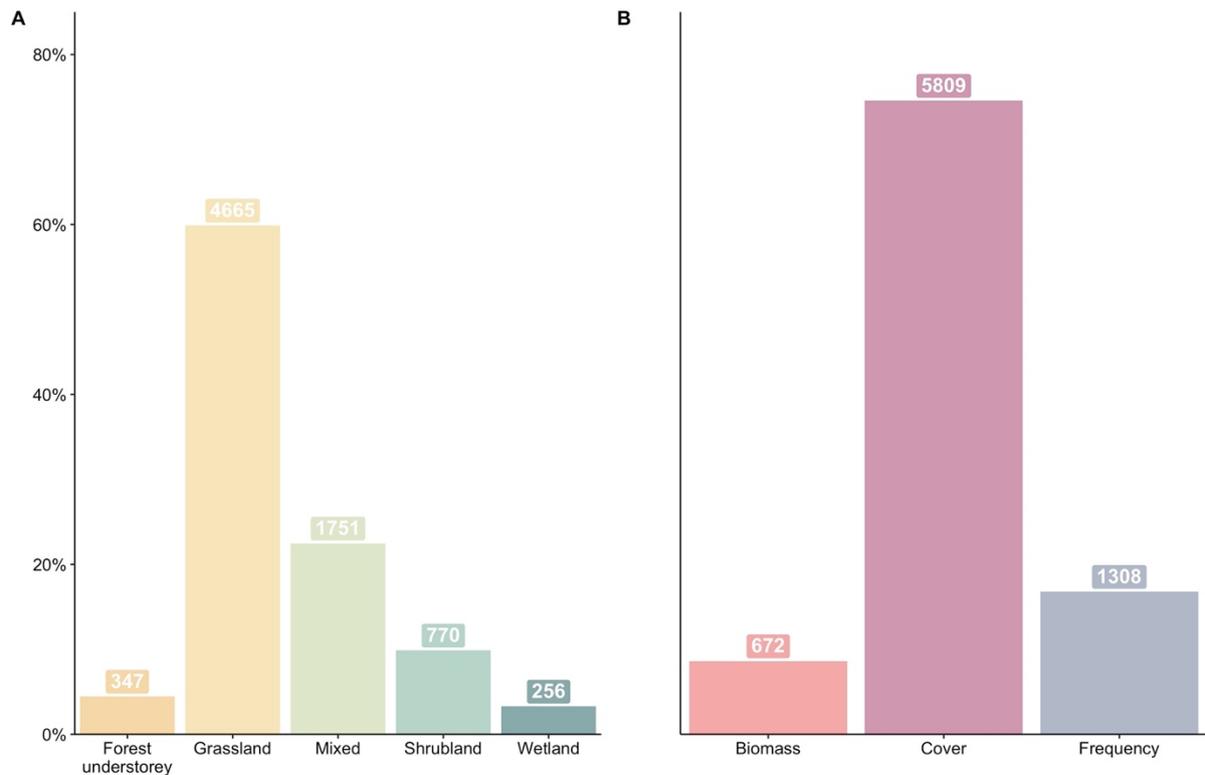


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107 **Fig. 3.** Climatic summary of the datasets included in LOTVS. Mean annual temperature and mean
 108 annual precipitation are plotted on the x and y axes, respectively. Each dot represents mean climatic
 109 conditions characterizing sites within each dataset. Dots are superimposed on Whittaker biomes (i.e.
 110 indicating potential vegetation; Whittaker, 1975), as redrawn from Ricklefs (2008). The plot was
 111 created using the R package “plotbiomes” (Stefan and Levin 2021).

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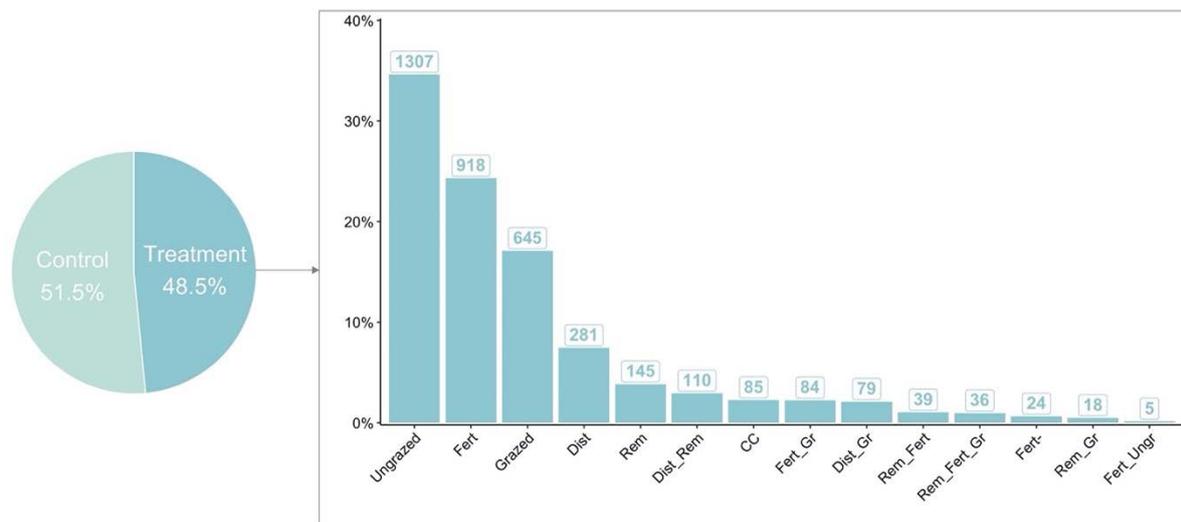
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 115 **Fig. 4.** *Habitat types (A) and sampling methods/approaches (B) covered in LOTVS. The number of time-*
 116 *series within each class is reported on top of the bars. See Box 1 for working definitions of habitat type*
 117 *and sampling methods.*

118
 119 In almost half of the permanent plots included in LOTVS (48.5%), vegetation has been subjected to
 120 experimental treatments manipulating abiotic or biotic conditions. The most frequent treatment types
 121 are herbivore exclusion, fertilizer application and grazing intensification (applied to ~35, 24 and 17%
 122 of the treated plots, respectively; Fig. 5). Yet, even in the absence of such treatments, LOTVS includes
 123 plots subjected to regular management regimes, such as mowing or grazing, that are necessary to
 124 maintain traditional land-use in given habitats.

125 Permanent plots in the LOTVS collection are surveyed using different techniques. The vast majority
 126 (~85%) are quadrat plots, but line transects and quadrat plots arranged along a transect are also present.
 127 Plot size ranges from 0.04 to 400 m²; ~ 80% of the plots range from 0.04 to 1.25 m², with 1 m² being
 128 the most frequent (49%) plot size in the collection. Information on plot size is missing for 90 plots,
 129 corresponding to 0.8% of the whole LOTVS collection. The method used to quantify species abundance
 130 also varies among the 79 datasets (Fig. 4B). The most frequent approach uses visual estimation of
 131 species cover (75%) followed by recording the frequency of individual species across a given number
 132 of subplots (17%) and third by the collection of aboveground biomass for each species in the plot (9%).
 133 In most cases, biomass clipping is intended to mimic mowing. All plots included in LOTVS were
 134 permanently marked in the field; geographic coordinates are available for either specific plots or for

135 unique localities of each dataset. Besides estimating plant species abundance, some of the datasets
 136 within LOTVS also include information about bare ground cover or the abundance of other taxa (e.g.
 137 bryophytes, lichens).

138 Taxonomic standardization is fundamental when compiling vegetation databases. On the one
 139 hand, it is a necessary step to (i) addressing the issue of nomenclature redundancy caused by the
 140 multitude of synonyms characterizing botanical literature (Kalwij 2012) and (ii) performing
 141 comparative analyses across datasets; on the other hand, it allows a later integration of data with
 142 ancillary information linked to taxonomic entities, e.g. species functional traits. Original datasets
 143 included in LOTVS show considerable variation in the chosen taxonomical references reflecting
 144 different regional and national traditions as well as the time when the work was undertaken. As such, a
 145 taxonomic standardization was deemed necessary. To do this, we standardized the nomenclature of
 146 plant species following The Plant List, currently the most widely used global reference list (Kalwij
 147 2012). This was done in R (R Core Team 2019) using the package “Taxonstand” (Cayuela et al. 2019),
 148 which allows the automatic standardization of plant names by running an internal query to The Plant
 149 List (<http://www.theplantlist.org>) and returning standardized species names, eventually resolving
 150 synonyms and homogenizing intraspecific taxonomic entities to the level of species. Nevertheless, a
 151 non-standardized version of the LOTVS collection, including (for each dataset) the original
 152 nomenclature of the species, is also available for potential users (see 3.2).



153
 154 **Fig. 5.** Left: pie chart describing the distribution of time-series included in the LOTVS collection
 155 according to the absence (“control”) or presence (“treatment”) of treatment. Right: distribution of the
 156 type of treatments present in LOTVS. The number of time-series within each treatment is reported on
 157 the top of the bars. Names for the treatments were abbreviated as follows: “Ungrazed”: grazing
 158 enclosure; “Fert”: fertilization; “Grazed”: grazing; “Dist”: disturbance; “Rem”: removal of plant

159 *species*; “*Dist_Rem*”: *disturbance + removal*; “*CC*”: *climate change*; “*Fert_Gr*”: *fertilization +*
160 *grazing*; “*Dist_Gr*”: *disturbance + grazing*; “*Rem_Fert*”: *removal + fertilization*; “*Rem_Fert_Gr*”:
161 *removal + fertilization + grazing*; “*Fert-*”: *decrease of productivity*; “*Rem_Gr*”: *removal + grazing*;
162 “*Fert_Ungr*”: *fertilization + grazing exclosures*.

163

164 **3. Data usage**

165 As an unprecedented collection of vegetation time-series, LOTVS has a huge potential to support timely
166 and innovative research in the fields of vegetation science, plant ecology and temporal ecology. It
167 should be noted, though, that installing and maintaining permanent plots is a very time- and resource-
168 consuming task, and thus a powerful collection of vegetation plots such as LOTVS can only arise from
169 collaborative efforts that stem from an impressive amount of work carried out by many data
170 contributors, whose effort must be acknowledged. To this end, anyone can contribute to LOTVS with
171 original data, as long as the data fully comply with LOTVS requirements (see section 3.2). At the same
172 time, data included in LOTVS can be requested and, based on their accessibility level (see section 3.2),
173 used following a simple procedure that is intended to support well-grounded research projects.

174 **3.1. Contributing data**

175 Detailed information on how to contribute to LOTVS, as well as specific data requirements can be
176 found on the dedicated website <https://lotvs.csic.es/contribute/>. LOTVS welcomes datasets including
177 vegetation time-series collected from permanent plots with a fixed (i.e. permanently marked)
178 geographical position in the field, possibly replicated in space, maintained for a minimum of six years
179 and sampled at annual intervals. In principle, the time-series should be continuous, i.e. no gaps should
180 be present. However, exceptions are allowed, provided that observations for some years are only
181 missing for a reduced number of plots. In cases of missing years for all of the permanent plots, the new
182 dataset can only be incorporated to LOTVS if a) only a very limited number of years is missing and b)
183 their distribution within the time-series is irregular, i.e. LOTVS is not intended to accept permanent
184 plots that, according to their original scope, are only sampled every n years. Following these
185 requirements, we are also not looking to include data collected in the context of so-called resurveying
186 studies, where historic vegetation plots are revisited after a longer time period and re-recorded. In fact,
187 these are the subject of other databases (e.g. the ReSurveyEurope initiative, [http://euroveg.org/eva-](http://euroveg.org/eva-database-re-survey-europe)
188 [database-re-survey-europe](http://euroveg.org/eva-database-re-survey-europe)). Also, whereas data collected using different sampling approaches are
189 welcomed (e.g. visual estimation of species cover, biomass, frequency, number of individuals), the
190 sampling approach should be consistent over time. LOTVS does not plan, at present, to incorporate
191 permanent plots that only record species occurrence (i.e. presence/absence data). To be included in
192 LOTVS, permanent plots should be preferably representative of natural or semi-natural vegetation. The
193 former can be defined as vegetation that developed in the absence of human influence and/or has long

194 been left undisturbed by humans; as to the latter, its existence and maintenance depend on human
195 practices (e.g. grazing or mowing) carried out for either production, conservation, or a mix of the two
196 purposes. As such, time-series data recorded from artificial seed mixtures such as those sown in
197 biodiversity experiments are not currently accepted.

198

199 **3.2. Requesting data**

200 Because of the effort needed to maintain permanent plots in time and collect temporal vegetation data
201 (Mills et al. 2015), access to individual time-series or datasets included in LOTVS is governed by a data
202 policy that allows data owners and contributors to remain in full control of their data if they so choose
203 (see <https://lotvs.csic.es/contribute/> for detailed information). The availability and access of single
204 datasets depend on the choice of individual data owners and contributors, who decide the accessibility
205 level of their data (from “restricted data”, i.e. data are only usable upon consent from data
206 owners/contributors, that should be expressed each time their dataset is requested; to “free data” i.e.
207 data that are freely available to use through the LOTVS platform). We note that about 65% of LOTVS
208 datasets are publicly available, either because they belong to Long-Term Ecological Research (LTER)
209 Programs, or because they were archived and published by their data owners. Also, several of the
210 LOTVS datasets are publicly available in their own right, via contact with their owners. Depending on
211 the accessibility level specified, data owners and contributors hold (or not, in case of freely usable data)
212 the right to request authorship on eventual publications based on the proposal submitted by the
213 applicants when they request the data. In all cases, to request data included in LOTVS, a short and
214 sound scientific proposal describing the aims of the project and the type of data required should be
215 prepared and submitted to the LOTVS’ Supervising Committee. This process is intended to i) minimize
216 conceptual overlap of proposals addressing highly similar research questions and ii) make sure that all
217 data owners are informed about the possible use of their data and are free to decline it if they wish so.
218 To help potential users get familiar with the data and facilitate data requests, we provide both a metadata
219 sheet describing the LOTVS datasets and a proposal template (both available on Zenodo:
220 <https://doi.org/10.5281/zenodo.5807378>), that will be regularly updated every time new datasets are
221 added to the LOTVS collection (or current datasets are updated). Together with a brief description of
222 each dataset, the metadata sheet contains information on several features (e.g. accessibility level,
223 number of plots, length of the time-series, habitat type, number of surveyed years, presence and type of
224 treatments, data type) which, in our opinion, should guide the users in their data requests. The proposal
225 template also includes a link to an interactive map displaying the geographical location of sampling
226 sites within the LOTVS collection, along with key dataset features.

227 **Perspectives**

228 In its present form, LOTVS includes vegetation time-series for almost 8000 permanent plots installed
229 and maintained in natural and semi-natural plant communities worldwide. Still, as we explained in
230 section 2, the geographical representation of both individual datasets and permanent plots is not
231 homogeneous; it is biased towards Europe and North America, and many habitats, such as forest
232 understoreys, are strongly underrepresented. In order to promote a more equal representation in terms
233 of geographical areas and habitats, one of the goals of this paper is to encourage new datasets including
234 time-series recorded using permanent plots located in currently underrepresented continents such as
235 Africa, Asia, Australia and South America. Similarly, time-series recorded in forest understoreys,
236 tundra and coastal areas would be particularly welcome. Furthermore, we are very interested in datasets
237 featuring spatial as well as temporal replication (minimum 6 years as mentioned above) to disentangle
238 the differences between temporal and spatial changes. Finally, to broaden the range of potential
239 applications of LOTVS (and in line with what has been done by other global initiatives, see Bruelheide
240 et al. 2019), we are planning to integrate it with information on environmental variables (climate, micro-
241 climate) and species functional traits. Such integration will eventually allow users complementary
242 access to ancillary data crucial to explore the evolution of different facets of diversity over time (Monnet
243 et al. 2014; Sperandii et al. 2021).

244 **Conclusions**

245 LOTVS possibly represents the largest collection of temporally fine-grained vegetation time-series
246 made accessible to the research community derived from permanent plots addressing the study of plant
247 communities through time. As such, LOTVS can be highly useful to perform timely research on a wide
248 range of topics in the field of vegetation science: investigating patterns and drivers of ecological
249 succession in natural plant communities, quantifying vegetation changes through time, as well as
250 assessing community stability and identifying its driving mechanisms. At the same time, because it
251 includes a considerable proportion of permanent plots subjected to some kind of treatment (e.g. grazing,
252 fertilization etc.), LOTVS can also support the development of large-scale studies aiming to understand
253 how temporal dynamics are affected by different treatments in the context of global changes. Last but
254 not least, we believe LOTVS could also serve as a valuable resource to conduct methodological research
255 addressing topics related to, for example, methods to quantify dissimilarity through time and their
256 partitioning (Baselga 2010; Legendre & Condit 2015) or quantitative approaches to investigate
257 community dynamics and more specifically, stability.

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