



Widespread sampling biases in herbaria revealed from large-scale digitization

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31	

32 SUMMARY

- Non-random collecting practices may bias conclusions drawn from analyses of herbarium records. Recent efforts to fully digitize and mobilize regional floras online offer a timely opportunity to assess commonalities and differences in herbarium sampling biases.
 We determined spatial, temporal, trait, phylogenetic, and collector biases in ~5
- million herbarium records, representing three of the most complete digitized floras of
- the world: Australia (AU), South Africa (SA), and New England, USA (NE).
- 40 3. We identified numerous shared and unique biases among these regions. Shared biases
- 41 included specimens i) collected close to roads and herbaria; ii) collected more
- 42 frequently during biological spring and summer; iii) of threatened species collected
- 43 less frequently; and iv) of close relatives collected in similar numbers. Regional
- 44 differences included i) over-representation of graminoids in SA and AU and of
- 45 annuals in AU; and ii) peak collection during the 1910s in NE, 1980s in SA, and
- 46 1990s in AU. Finally, in all regions, a disproportionately large percentage of
- 47 specimens were collected by very few individuals. We hypothesize that these mega-
- 48 collectors, and along with their associated preferences and idiosyncrasies, shaped

49 patterns of collection bias via 'founder effects	;'.
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- 50 4. Studies using herbarium collections should account for sampling biases, and future
 51 collecting efforts should avoid compounding these biases to the extent possible.
- 52 Keywords: Herbarium, collector bias, geographic bias, regional flora, sampling bias,
- 53 temporal bias, trait bias
- 54

INTRODUCTION 55 Herbaria contain a wealth of information about the ecological and evolutionary history of 56 living and extinct species (Funk, 2003). Despite the continuous decline in plant collecting 57 and declining support for herbaria (Dalton, 2003; Prather et al., 2004a, b), there has been 58 a recent surge of studies leveraging herbarium collections for diverse research projects 59 not focused on systematics (Pyke & Ehrlich, 2010; Lees et al., 2011; Feeley, 2012; 60 Lavoie, 2013; Hart et al., 2014). These studies include plant demography, current and 61 future species distributions, and temporal changes in phenology and morphology (e.g., 62 Miller-Rushing et al., 2006; Newbold, 2010; Pyke & Ehrlich, 2010; Lavoie, 2013; Staats 63 et al., 2013; Davis et al., 2015; Willis et al., 2017a,b). 64 Ideally, herbarium collections used for these studies would include statistically 65 unbiased samples of plant diversity across space and time. However, as the majority of 66 specimens were collected for qualitative taxonomic and/or systematic inquiries, they 67 were usually collected non-randomly and sampling designs were rarely quantified (Wolf 68 et al., 2011; Schmidt-Lebuhn et al., 2013). Because non-random samples may be 69

statistically biased, analyzing them without accounting for biases might lead to spurious
results (Syfert *et al.*, 2013).

72 Sampling biases fall into several broad categories. Taxonomic or phylogenetic bias is the unbalanced sampling of certain taxa or clades over others, typically resulting 73 74 from the scientific interests of a collector or the attractiveness of plants (Hortal *et al.*, 2007). Geographic bias occurs when specimens are collected more frequently in one 75 76 place than another, often because of differential accessibility (Hijmans *et al.*, 2000). Temporal bias occurs when collection activity is favored in certain years or parts of the 77 78 year (Cotterill et al., 1994; Funk & Morin, 2000; Norris et al., 2001). Meyer et al. (2016) evaluated worldwide terrestrial plant occurrence data using 120 million records from the 79 Global Biodiversity Information Facility (GBIF; Edwards et al., 2000). Their analyses 80 revealed large taxonomic gaps in global plant occurrence data ($\leq 25\%$ of species of land 81 82 plants were sampled); extensive spatial gaps across regions that harbor high 83 concentrations of plant diversity, especially in Asia, Central Africa, and Amazonia; and strong temporal discontinuities in occurrence records across decades, all of which can 84 hamper inferences about the effects on plants of recent and future environmental change. 85

86 Although Meyer *et al.*'s (2016) study represents the most comprehensive effort to assess biases in plant collections at a global scale to date, the vast majority of herbarium 87 88 collections have not been digitized, and of those that have, many are unavailable, in whole or in part, on GBIF. Thus, Meyer et al.'s (2016) assessment of biases may itself be 89 biased, or may inaccurately reflect biases in more complete, regional botanical 90 collections that have been more fully mobilized. Furthermore, over two-thirds of the plant 91 92 records in GBIF are not tied to physical specimens, and thus cannot be easily validated by others (Cotterill, 1995). For these reasons we suspect that an analysis of finer-grained 93 collection data, focused on specific regions that have been predominantly digitized and 94 validated, may reveal clearer patterns of sampling biases between regions than the global 95 trends identified by Meyer et al. (2016) (cf. Hijmans et al., 2000 for Bolivian potatoes). 96 Expanding upon Meyer et al.'s work, we explored spatial, temporal, and 97 taxonomic/phylogenetic sampling biases in collections from three of the most extensively 98 collected, digitized, and mobilized regional floras in the world: South Africa (SA), 99 Australia (AU), and the New England (NE) region of the United States. The SA flora is a 100 101 compilation of digitized herbarium specimens from all major herbaria across the country available in a single online portal (South African National Biodiversity Institute [SANBI], 102 103 2016; le Roux et al., 2017). The Australian Virtual Herbarium (AVH, 2016) is the main database for AU. It contains digitized herbarium specimens from all the major herbaria in 104 105 AU. The Consortium of the Northeast Herbaria database contains digitized specimens from 15 participating herbaria in the NE region of the United States (Schorn et al., 2016). 106 107 We also examined trait bias – sampling bias due to intrinsic life-history characteristics, including life cycle (annual vs. perennial), plant height, growth form (woody vs. 108 109 herbaceous), and species conservation status. Finally, we examined the contributions of individual collectors to each flora. We identified biases in all five of these categories 110 within each of these regional floras. Our results revealed both commonalities and 111 differences in regional collection biases and identified new sampling foci as collections 112 113 grow in the future. 114

115 MATERIAL AND METHODS

116 Sources and description of data

117 We obtained 12,488,200 herbarium specimen records of vascular plants from AU (Australia Virtual Herbarium [AVH], 2016); 2,049,905 herbarium specimen records from 118 119 SA including Lesotho and Swaziland (South African National Biodiversity Institute [SANBI], 2016); and 879,388 herbarium specimen records from the NE (USA) flora 120 121 (Consortium of Northeastern Herbaria [CNH], 2016). The records were cleaned in two steps (Fig. S1). First, we standardized the taxonomy of all species using the Taxonomic 122 123 Name Resolution Service v.4.0 (Boyle et al., 2013). This online tool corrects and standardizes plant names against reference taxonomies, such as Missouri Botanical 124 Garden's Tropicos (http://tropicos.org/) database or the PlantList (http://theplantlist.org/). 125 Second, we removed specimens that were duplicates from the same collection locality 126 and date; specimens with clearly erroneous locations (*i.e.*, in oceans); specimens with 127 zero coordinates and occurrences that fell outside the boundaries of our study; specimens 128 missing exact collection date or georeferenced location data; and field observation 129 records not tied to a physical specimen. Following this data cleaning, we retained 32% of 130 the initial specimens for further analysis: 24% of the AU records (31,966 taxa; 2,958,195 131 132 records); 49% (20,824 taxa, 1,008,206 records) from SA; and 75% (3719 taxa, 661,370 records) from NE. 133

134

135 Analyses

136 *Spatial biases*

First, we evaluated the density of sampling localities across the focal regions using 137 Delaunay triangulation polygons, which measure the land area covered by each sampling 138 locale (Fortune, 1992). Larger triangles indicate sparser collecting effort, whereas smaller 139 140 triangles indicate more concentrated effort. Second, we examined infrastructure bias by calculating the minimum distance of each collection locality to the nearest major road 141 (GADM, 2015) and herbarium (following Thiers, 2016). Our dataset of roads derives 142 from the publicly available Digital Chart of the World (http://maproom.psu.edu/dcw/), 143 144 which was compiled by the US Defense Mapping Agency from 1:1,000,000 scale paper maps (ESRI, 1992). All roads appearing at this scale were included in our analyses. 145 Although this dataset includes only larger roads and has not been updated since 1992, it 146 likely represents the most comprehensive digital record of roads around the world. We 147

148 then compared these distances to those generated by a null model (1000 iterations) in 149 which the same number of sample points was randomly (Poisson) distributed across each 150 geographic region. Third, we mapped geographic biases in sampling density, defined as areas of excessive (hotspots) or insufficient (coldspots) collection (Hijmans et al., 2000). 151 Hotspots and coldspots were determined at a spatial grain of $0.25^{\circ} \times 0.25^{\circ}$ based on the 152 number of specimens per grid cell, and identified using the 2.5% threshold (Ceballos & 153 Ehrlich, 2006; Orme et al., 2005; Daru et al., 2015), based, respectively, on the 97.5th and 154 2.5th percentile values in the number of specimens collected per grid cell. Spatial distance 155 calculations were computed with the functions *dist2Line* and *spDists* in the R packages *sp* 156 (Bivand et al., 2013) and geosphere (Hijmans, 2015), respectively. In our final predictive 157 model of sampling density, we also included human population density (CIESIN, 2016), 158 sampling localities, infrastructure (distance to herbaria and roads), number of specimens 159 collected, and elevation 160

161

162 Temporal bias

For each regional flora, we explored bias at several temporal scales. Collection dates 163 ranged from 20 May 1664 to 9 January 2016 (AU), 15 November 1656 to 6 June 2016 164 (SA), and 28 July 1687 to 4 May 2016 (NE). We hypothesized that collectors tended to 165 avoid fieldwork during unfavorable conditions (e.g., winter, wartime) or certain days of 166 167 the week (e.g., weekdays for non-professional botanists). To test for temporal bias, we first re-coded collection dates as days of the week (Sunday = 1, Monday = 2, etc.), and 168 day of the year (DOY; where January 1 = 1 DOY and December 31 = 365 DOY, etc.). 169 We then used a Rayleigh test of directional statistics in the R package *circular* 170 171 (Agostinelli & Lund, 2013) to test whether each of these collection dates were randomly distributed against all dates spanning the entire duration of plant collection. If $P < \alpha =$ 172 173 0.05, we rejected the null hypothesis of temporal uniformity at scales of weeks, days of the year, or decades. 174

175

176 Trait bias

177 We used customized R scripts to harvest information on growth duration (annual *vs*.

perennial), growth form (woody vs. herbaceous), and height for each species from online

179 regional databases (all accessed in June 2016), including: New South Wales Flora Online 180 (http://plantnet.rbgsyd.nsw.gov.au); JSTOR Global Plants (https://plants.jstor.org); Atlas 181 of Living Australia (http://bie.ala.org.au); Plants of Southwestern Australia (http://keys.lucidcentral.org); the African Plant Database (http://ville-ge.ch); Plants of 182 Southern Africa (http://plantzafrica.com); Plant Resources of Tropical Africa 183 (http://www.prota4u.org); Flora of North America (http://efloras.org); and the USDA 184 Plants Database (http://plants.usda.gov). We then manually checked these data for 185 inconsistencies in terminologies for defining certain traits. For example, 'vines' vs. 186 'lianas' for climbers, 'forbs' vs. 'herbs' for herbaceous life forms, 'biennial' for perennial 187 growth duration. Extinction risk assessments for each species were obtained from the 188 IUCN Red List database (www.iucnredlist.org, accessed August 2016), which uses the 189 following categories: Data Deficient (DD), Least Concern (LC), Lower 190 Risk/Conservation Dependent (LR/CD), Near Threatened (NT), Vulnerable (VU), 191 Endangered (EN), Critically Endangered (CR), and Extinct (EX). We grouped these 192 narrow categories into two broader threat categories, threatened (EX+CR+EN+VU) or 193 194 not threatened (LR/CD+NT+LC), following Yessoufou et al. (2012). Trait bias was evaluated using a chi-squared test to contrast the number of 195 observed specimens collected per species with the abundance of a species if specimen 196 collection was equal across all species for each trait category. Because of dramatically 197 198 unequal sampling effort in some species -e.g., Senna artemisioides with 10,167 199 specimens vs. Eucalyptus cordieri with only one – and the low coverage of taxa with 200 available trait data, we randomly sampled 50 specimens from each available species with trait data using 1000 randomizations. Species with less than 50 specimens were excluded 201

202 from this analysis.

203

204 Phylogenetic bias

We assessed phylogenetic signal in collection frequency as a measure of phylogenetic
bias using two different tests following Wolkovich *et al.* (2013). A strong phylogenetic
signal – closely related species sharing similar collection frequency – would suggest
phylogenetic bias in collections. We first assembled a phylogeny using Phylomatic
(Webb & Donoghue, 2005), enforcing a topology that assumed the APG III (APG III,

210 2009) backbone (tree R20120829). This phylogeny included all species in our analysis, 211 but provided only an approximate degree of relatedness based on taxonomic hierarchy at 212 family level; many relationships, especially within genera, were unresolved. This is problematic because recent theoretical and empirical studies have shown that a lack of 213 214 resolution in a community phylogeny may mask significant patterns by reducing statistical power (Schaefer et al., 2011; Daru et al., 2017) or suggest significant 215 phylogenetic patterns that are not supported by more completely resolved phylogenies 216 (Davies et al., 2012). 217

To alleviate these concerns, we also tested for phylogenetic bias by including only those species sampled in the dated molecular phylogeny inferred from seven genes for 32,223 plant species (Zanne *et al.*, 2014). Although this phylogeny has been criticized (Edwards *et al.*, 2015), it nonetheless represents the single largest phylogeny to date for flowering plants. The taxon sampling for testing phylogenetic bias included 5814 species from AU, 3568 from SA, and 4269 from NE.

We estimated phylogenetic signal using three common metrics: Abouheif's C_{mean} 224 (Abouheif, 1999), Blomberg's K (Blomberg *et al.*, 2003), and Pagel's lambda (λ) (Pagel, 225 1999). Significance was assessed by comparing observed values to a null distribution 226 created by shuffling the trait values across the tips of the phylogeny 1000 times. Pagel's λ 227 uses a maximum-likelihood method with branch-length transformation to estimate the 228 229 best-fit of a trait against a Brownian model. Values of Pagel's λ range from 0 (no phylogenetic signal) to 1 (strong phylogenetic signal). Both Blomberg's K (a significant 230 231 phylogenetic signal is indicated by a K value > 1) and Pagel's λ were calculated using the R package *phytools* (Revell, 2012). Abouheif's C_{mean} was calculated using *adephylo* 232 233 (Jombart & Dray, 2008). We tested the sensitivity of our analysis by exploring phylogenetic signal in collecting effort across nine well-sampled clades as represented in 234 235 NE: Asteraceae, Brassicaceae, Cyperaceae, Ericaceae, Fabaceae, Lamiaceae, Poaceae, Ranunculaceae, and Rosaceae. 236

In addition to phylogenetic signal, we also used phylogenetic generalized least squares regressions (PGLS) in the R package *caper* (Orme *et al.*, 2012) to model collecting effort per species in each region as a function of species evolutionary ages, evolutionary distinctiveness (ED), and 'evolutionary distinctiveness and global

241 endangerment' (EDGE; Isaac et al., 2007). Species ages were measured as the length of 242 terminal branches (BL) linking species on a phylogenetic tree. ED measures the degree of 243 phylogenetic isolation of a species, whereas the EDGE metric was determined by calculating the ED score of each species (Isaac et al., 2007) and combining it with global 244 endangerment (GE) from IUCN conservation categories: $EDGE = ln(1 + ED) + GE \times$ 245 ln(2), where GE represents expected probability of species extinction over a 100-year 246 period (Redding & Mooers, 2006) categorized as follows: least concern = 0.001, near 247 Threatened and Conservation Dependent = 0.01, Vulnerable = 0.1, Endangered = 0.67, 248 and Critically Endangered = 0.999. 249 Last, we examined the phylogenetic structure of collecting efforts across decades 250

to test for patterns of phylogenetic overdispersion and clustering through time. Temporal
phylogenetic structure by decade (*i.e.*, 1901-1910, 1911-1920, *etc.*) was evaluated using
the net relatedness index (NRI) and nearest taxon index (NTI; Webb *et al.*, 2002, 2008).
NRI describes a tree-wide pattern of phylogenetic dispersion, whereas NTI evaluates
phylogenetic structure towards the tips of the phylogeny. Negative values of NRI or NTI
indicate phylogenetic overdispersion whereas positive values indicate phylogenetic

258

259 *Collector bias*

We determined collector bias by tabulating the number of specimens amassed by each collector in all three floras. We then examined Pearson's product-moment correlation between the numbers of specimens collected per collector with the number of species collected per collector.

264

265 Computation and availability of data and code

All statistical analyses were conducted using the Research Computing Clusters of

- 267 Harvard University (https://rc.fas.harvard.edu/). Data files and custom R scripts are
- available from the Harvard Forest Data Archive, dataset HF296

269 (http://harvardforest.fas.harvard.edu/data-archive).

270

271 **RESULTS**

272 Spatial bias

High sampling density was observed in southeast and southwest AU, the Cape region of 273 274 SA, and two of the six NE states (Connecticut and Massachusetts) relative to other parts of those regions (Fig. 1a-c). When we weighted each sampling locale by the number of 275 specimens, we found a mismatch between hotspots (top 2.5% quantiles) and coldspots 276 (lowest 2.5% quantiles) of sampling intensity (Fig. 1d-f), suggesting hotspots and 277 coldspots were not randomly distributed. Hotspots of collecting tend to cluster around 278 coasts in AU and SA, whereas coldspots were abundant in interior areas. In NE, hotspots 279 were concentrated in the south and coldspots occurred in the north. 280

Herbarium specimens tended to be collected closer than expected to roads and 281 herbaria (p<0.01; Fig. 2a, b). More than 50% of herbarium specimens were collected 282 within 2 km of roadsides in all three floras (p<0.01; Fig. 2a). Moreover, distance to 283 herbaria explained 45% of the variance in collecting effort in AU, 29% in SA and 12.3% 284 in NE, with a higher density of specimens closer to herbaria (Table 1). Despite substantial 285 gradients in altitudes in each region (-15 - 2022 m a.s.l. in AU; 1 - 3254 m a.s.l in SA;286 and -3 – 1485 m a.s.l. in NE), most specimens were collected below 500 m a.s.l in AU 287 and NE (81%, 44%, and 93% of specimens in AU, SA, and NE, respectively; Fig. 2c). 288 289 We also found a negative correlation between collecting effort and altitude in Australia and South Africa, suggesting a tendency for specimens to be collected at lower elevations. 290 291 However, the opposite was true for NE, where more specimens tended to be collected at higher elevations than expected by chance (Table 1). 292

293

294 *Temporal bias*

295 There were historical biases in collection efforts in the three floras: low sampling until 1880 in AU and SA, and a burst of collections in NE in the early 20th century (Fig. 3). 296 297 Conversely, there was a dramatic increase in botanical collection in SA and AU after World War II, peaking in the 1980s and 1990s, respectively (Fig. 3). This peaking 298 occurred ~ 100 years after peak collection activity in NE. Seasonally, specimen 299 collections were biased toward spring and summer for all three floras, with peak 300 collections ranging from September to December in AU and SA (Rayleigh Z = 0.189 and 301 Z = 0.251 respectively, both p < 0.001), and May to September in NE (Rayleigh Z = 302

303

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0.718, p < 0.001; Fig. 4a). There was a significant trend towards collection on weekends

304	(Saturdays and Sundays) in NE (Rayleigh test $Z = 1.0$, $p < 0.001$) and midweek in SA
305	and AU (Rayleigh test $Z = 0.105$ and $Z = 1.0$, respectively; both $p < 0.001$; Fig. 4a).
306	
307	Trait bias
308	Perennials were more frequently collected than annuals in terms of specimens per species
309	in SA and NE; the opposite was true for AU where there was a greater tendency for
310	annuals to be collected (Fig. 5a). Similarly, graminoid specimens per species were over-
311	represented relative to other habits in AU and SA, whereas herbs and trees were over-
312	represented in NE (Fig. 5b). Relatively short plants were more frequently represented
313	than taller plants in all three floras: 79.3%, 89.3% and 84.9% of the plants collected in
314	AU, SA and NE, respectively were less than 5 m in height (Fig. 5c).
315	Threatened species were collected significantly less often than non-threatened
316	plants across all three floras (all $p < 0.001$; Fig. 5d).
317	
318	Phylogenetic bias
319	It is possible that closely related species were collected similarly (either under-collected
320	or over-collected) than expected by chance. We tested this assumption phylogenetically
321	and found a significant, but weak phylogenetic signal in the abundance of specimens per
322	species across all three floras (Table 2). Specifically, closely related species tended to
323	have a more similar number of specimens than expected (Table 2; Fig. 6). This
324	phylogenetic bias was strongest in SA (Abuoheif's C_{mean} = 0.15 and λ = 0.32; both p <
325	0.01, but K = 0.0013 [NS]). For instance, in SA, collections from the genus <i>Protea</i>
326	averaged 115 specimens per species whereas only two specimens were collected for
327	species in the genus Rytigynia on average. Most Agoseris in NE were represented by < 10
328	specimens per species, whereas many fern genera were represented by high specimen
329	numbers (e.g., Onoclea with 845 specimens/species). Australian collections showed the
330	weakest phylogenetic bias (Abuoheif's $C_{mean} = 0.12$ and $\lambda = 0.18$, both $p < 0.01$, but $K =$

0.00085 [NS]; Fig. 6). Phylogenetic signal varied at the family level as well in NE, with

Asteraceae showing the strongest collection bias (Fig. 7), followed by Cyperaceae,

Poaceae, and Rosaceae (Table S1). These families were represented by much highercollection numbers than for other families.

EDGE was a significant predictor of collecting efforts in all three floras (p < 0.001), with variance ranging from 1.89% (NE) and 3.75% (AU), to 8.89% in SA. In general, EDGE species (distinctive, endangered species) were generally under-collected in terms of specimens per species (Table 3).

Lastly, floristic collecting showed a general trend of phylogenetic clustering within decades for all three floras. The collection of different clades of plants was not evenly distributed across time. NTI was significantly positive in each flora, indicating that clustering occurred near the tips of the phylogeny (Fig. 3). We only observed significant phylogenetic clustering at the deeper nodes of the phylogeny, as indicated by NRI, in SA (Fig. 3); deeper phylogenetic clustering was weak in NE and AU (Fig. 3).

345

346 *Collector bias*

347 The number of specimens per collector was highly skewed (Fig. 8). In AU, more than 50%

of the examined specimens were amassed by only 2% of the collectors, including A.C.

Beauglehole (46,728 specimens), B. Hyland (32,019 specimens), and P.I. Forster (30,280

specimens; Fig. 8a). In SA, more than 50% of the specimens were amassed by 9.5% of

collectors, including J.P.H Acocks (19,344 specimens), E.E. Esterhuysen (15,566

specimens), and E.E. Galpin (14,146 specimens; Fig. 8b). In NE, 50% of the specimens

were contributed by 3.2% of the collectors, including L.J. Mehrhoff (19,149 specimens),

M.L. Fernald (14,368 specimens), and A.S. Pease (12,238 specimens; Fig. 8c). The

number of specimens amassed by these collectors was very strongly positively correlated

with the number of species they collected, suggesting that these collectors were mainly

doing general collecting rather than focusing on a particular group of plants (r = 0.85 in

358 AU, 0.95 in SA and 0.84 in NE; all p < 0.01; Fig. S2).

359

360 **DISCUSSION**

361 Historically, the primary function of herbaria has been to serve as a resource for botanists

- 362 carrying out taxonomic and systematic research, allowing users to construct
- 363 classifications of plants, verify identifications, determine the ranges and morphological

364 characteristics of species, and develop local and regional floras (Greve *et al.*, 2016). Over time, new uses for specimens have arisen, and now more than ever, they are being used in 365 366 ways that collectors rarely imagined (Pyke & Ehrlich, 2010; Lavoie, 2013; Willis et al., 2017a,b; Nualart et al., 2017; Rudin et al., 2017). Accordingly, attempts to assess and 367 categorize biases inherent in these collections have been made (Rich & Woodruff, 1992; 368 Geri et al., 2013; Schmidt-Lebuhn et al., 2013; Meyer et al., 2016; Stropp et al., 2016). 369 370 Among these, the most comprehensive investigation is by Meyer et al. (2016), who proposed an important conceptual framework for analyzing gaps and biases along 371 taxonomic, geographical, and temporal dimensions. Although Meyer et al. (2016) 372 focused more on observational records than herbarium collections, they uncovered 373 numerous biases in 'digitally accessible information' regarding plants and provided an 374 important baseline for evaluating and improving global floristic coverage in collection 375 data. However, collection biases in large geographic areas are difficult to categorize 376 because the collections data are not vet digitized, and this may skew the global patterns 377 of bias noted by Meyer et al. (2016). By focusing on three of the most well-collected and 378 379 digitized floras in the world, we reduced effects of missing or unavailable data, and most importantly, could evaluate commonalities and differences in patterns of bias among 380 regional collections. 381

382

383 Spatial bias

Our data confirmed the tendencies for botanists to collect along roadsides (*e.g.*, Funk &

Richardson, 2002), near herbaria (*e.g.*, Hijmans *et al.*, 2000; Moerman & Estabrook,

2006; Pautasso & McKinney, 2007), in more accessible areas (Rich & Woodruff, 1992),

and at lower elevations. Before automobiles became common in the 1920s, botanists

often walked or rode domesticated animals to collection sites (Botkin, 1968; Belasco,

- 1979). As our modern infrastructure developed (*e.g.*, roads, railroads, and cities that
- contain herbaria) and access to automobiles increased, spatial biases associated with
- infrastructure likely increased as botanists tended to travel and collect using automobiles
- 392 (Everill *et al.*, 2014). Because roads are known alter local environmental conditions and
- facilitate biological invasions (e.g., Forman & Alexander, 1998; Hui et al., 2003; Griffith

394 et al., 2010; Li et al., 2014) and botanists and herbaria predominate in cities, specimens collected in proximity to either are unlikely to represent a random sample across species 395 396 distributions. Specifically, species collected along roadsides are likely to be overrepresented by species that thrive with disturbance, and under-represented by forest 397 interior and wetland species that are harmed by disturbance (Gutzwiller & Flather, 2011; 398 Rivers-Moore & Cowden, 2012). As the road network continues to expand and as people 399 become evermore concentrated in cities, this bias toward collecting near roads might 400 become stronger in coming decades. 401

Collection bias towards lower elevations (< 500 m) was striking in SA and AU, 402 despite extensive collection efforts in adjacent hyper-diverse hotspots such as the 403 mountains in the Cape Fold Belt (SA), and Mount Lesueur-Eneabba (Western AU). This 404 405 is likely due to the presence of the arid and relatively species-poor Great Karoo Plateau (SA), Great Sandy Desert (AU), which each encompass over a third of the respective 406 study sites, but account for only a small proportion of the biodiversity of each region. As 407 a result, the low-elevation collection bias in the floras may reflect actual species 408 409 abundance. In NE, the trend toward collecting at higher elevation might be due to the strong tendency for botanists to visit the White Mountains and Mount Katadhin to collect 410 alpine species. 411

Although we realize that patterns of species richness may not be randomly 412 413 distributed across the landscape, accounting for underlying patterns of richness or abundance is difficult because our knowledge of such patterns often derive from (and are 414 415 thus not independent from) these same (biased) collections. By comparing locations of samples (collections) against a Poisson set of points and specimens per species, and not 416 417 total collection numbers, we tested only for the non-random distribution of collection locations on a landscape. And indeed, we found that the collection locations were not 418 spatially distributed randomly (Poisson) on the landscape. It is also possible that 419 georeferencing might have introduced additional bias in some specimens. While 420 421 ascertaining the degree of accuracy of georeferenced records might be challenging because such information is often unavailable, our cumulative curves are likely less 422 affected. 423

424

425 Temporal and seasonal bias

Collections in AU and SA have increased through time until a few decades ago, but those 426 in NE peaked much earlier in the early 1900s. These differences between regional 427 collection activities may parallel broader societal factors influencing plant collection. In 428 429 NE, for example, the establishment of the New England Botanical Club during the 1890s (NEBC, 1899) preceded a surge and peak in collecting activity associated with prolific 430 botanical expeditions of the region coinciding with the 'Golden Age' of plant collecting 431 in Europe and North America (Whittle, 1970; Musgrave et al., 1999). In SA, collection 432 efforts began much later, peaking during the Apartheid Era (1948–1994), and declined 433 thereafter under the New Democratic Rule, concomitant with the general economic 434 decline of the country and concern for public safety (Ferreira & Harmse, 2000; Lemanski, 435 2004). In AU, the mass immigration of Europeans in 1948 after World War II included 436 numerous highly skilled professionals (Price, 1998; Leuner, 2007) and coincided with an 437 438 enormous increase in botanical collecting. Botanical collecting may have declined more recently owing to legislation in AU and SA to regulate collections activities, especially 439 those designed to protect rare and endangered species. 440

Collecting efforts within a season revealed common patterns of bias: specimens in 441 442 all three regions were collected overwhelmingly in biological spring and summer. Sampling during these time periods likely reflects efforts to collect plants in good 443 444 flowering and fruiting condition. However, this seasonal bias likely overlooks key developmental transitions (e.g., Poethig, 2013), including bud formation, bud break, leaf 445 out, fruit development, and leaf senescence (van der Schoot et al., 2014). Supporting this 446 argument, these temporal patterns were most pronounced in NE, which experiences the 447 harshest winter climates of the three regions. Plants collected during the winter season are 448 almost always in dormant condition, and often lack the leaves and reproductive structures 449 450 needed for taxonomic research. Collecting was also more likely during holidays and school vacations in NE and AU. 451

452

453 Trait bias

In all three regions, short to medium-height species were collected more frequently than tall species (>5 m). This pattern is presumably related to the relative ease of collecting specimens from shorter, often herbaceous, species, and because reproductive materials are more accessible and potentially more abundant. Specimens of trees with woody twigs also are typically bulkier and more difficult to prepare, which may reduce their collection frequency.

Threatened species were also greatly under-represented in all floras. This is 460 perhaps not surprising given their limited abundance (Palmer et al., 2002) and imposed 461 collecting restrictions (Klemens & Thorbjarnarson, 1995; Pritchard, 1996; Gibbons et al., 462 2000; Robinson, 2001). However, it is also true that collectors sometimes oversample 463 rare or threatened species because of their higher scientific value and avoid the more 464 common ones (Garcillán et al., 2008; Garcillán & Ezcurra, 2011; Minteer et al., 2014). 465 Regardless of past practices and contemporary formal restrictions, botanists now often 466 avoid over-collection of such species by following informal guidelines and collecting 467 plants only in areas with numerous individuals of the species (Iwanycki, 2009). Although 468 469 great care in collecting rare plants is important, under-collection of rare species may lead to incorrect extinction risk assessments (*i.e.*, that the species is rarer than it actually is) 470 and greatly limit opportunities to glean historic population and biogeographic data to 471 guide species conservation and restoration. 472

473 Annuals were over-represented relative to perennials in herbarium collections in 474 AU; the opposite was observed in SA and NE. There was also a high representation of 475 graminoids in herbarium collections in AU and SA. This result may stem from the higher likelihood of common species being collected multiple times by different individuals or 476 477 expeditions. Along these lines, much of AU is dominated by annual grasses, and the savannas of SA are populated by a variety of native and non-native perennial grasses 478 479 interspersed with forbs and woody plants (Bond & Parr, 2010). New England, on the other hand, is generally forested and has an abundance of shade tolerant shrubs and 480 481 perennial herbs. Graminoids are also considered harder to identify and may be avoided by 482 non-specialists. Lianas and vines simultaneously represent the smallest proportion of growth forms and comprise the least number of specimens per species in all three floras. 483 484 Such trait-based biases in botanical collections not only influence our perception of

species abundance and range, but can also lead to erroneous estimations of functional diversity and ecosystem services, especially for studies relying on specimen databases (Schmidt-Lebuhn *et al.*, 2013). Whether herbarium records represent true patterns of abundance and diversity remains difficult to untangle from human-mediated collecting biases. However, the large differences among the three floras in the traits of botanical collections almost certainly is reflective of genuine difference species abundance and diversity.

492

493 *Phylogenetic bias*

Taxonomic biases in collection data have been reported previously (Hijmans *et al.*, 2000;
Tobler *et al.*, 2007; Meyer *et al.*, 2016). However, our study is the first, to our knowledge,
to demonstrate explicit evidence for phylogenetic bias in herbarium collections.
Collection efforts in all three floras were concentrated in particular clades.

Previous examinations of taxonomic bias (e.g., Hijmans et al., 2000; Tobler et al., 498 2007; Meyer et al., 2016) did not use the full complement of modern phylogenetic 499 methods that included patterns of evolutionary relatedness, and so were limited in their 500 501 ability to detect details of taxonomic bias. In contrast, our phylogenetic approach not only captured taxonomic bias in favor of certain entire families (e.g., Asteraceae, Cyperaceae, 502 Poaceae, and Rosaceae in NE), but revealed that evolutionarily distinct and globally 503 504 endangered species are underrepresented in herbarium records relative to more common species. Such evolutionarily distinct species, which are threatened with extinction, 505 represent important targets for future documentation or prioritization for conservation 506 507 (Isaac et al., 2007). However, collecting threatened taxa requires specialized training, compliance with regulation, and awareness of actual collection needs (Minteer et al., 508 2014). Increasingly, DNA barcoding approaches, using small samples from living tissues, 509 510 combined with GPS-referenced digital photography might be an avenue to document 511 such species.

512

513 Collector bias

In all three regions, a large percentage of specimens was gathered by only a few 514 collectors (Fig. 8). Thus, the habits and preferences of a few individuals likely shaped the 515 516 establishment and formation of these herbarium collections. These 'founder effects' propagate across all the dimensions of collection bias examined above, and help us to 517 understand past collection behavior. For example, certain collectors may focus on 518 geographically circumscribed floristic zones, often near their place of residence, 519 520 workplace, or vacation home, and sample all species found therein, whereas others may focus on collecting species of a particular clade across various regions. Professional 521 botanists may tend to collect specimens on weekdays during any time of the year, 522 whereas amateurs and faculty with teaching responsibilities may focus their efforts on 523 weekends and vacation months. Those interested in function and physiology may only 524 collect plants of certain habits or life-histories (e.g., carnivorous, aquatic plants, or 525 succulent plants). These effects would likely be compounded when associated with mega-526 collectors. For instance, the Harvard University Herbaria's collection of Asian, especially 527 woody plants, was largely built by a few collectors and dates to the early establishment of 528 529 the institution, and continues to attract scholars of the flora of Asia and their collections. Investigating the historical significance and potential biases created and propagated by 530 these early pioneers is a ripe area for future research. 531

532

533 **Future collecting**

To ensure that herbaria continue to be vital centers for research beyond their importance 534 to taxonomy and systematics, herbarium directors and collectors should account for and, 535 whenever possible, reduce biases in plant collections. Biases can be accounted for to a 536 537 degree using statistical approaches (Droissart et al., 2012; Feeley, 2012; Grass et al., 2014; Engemann et al., 2015). For instance, inclusion of covariates for distances of 538 collections from herbaria, roads, or other infrastructure (McCarthy et al., 2012), using 539 rarefaction methods to predict abundances (Schmidt-Lebuhn et al., 2013), or including 540 541 the collector as a variable, would improve species distribution models and associated predictions of future changes across a flora. To remedy such biases, future collecting 542 expeditions should focus on "coldspots" of collection intensity (Hijmans et al., 2000), 543

544 that is, places that are under-represented in collections. Although some of the coldspots we identified likely represent more inaccessible environments, they often correspond to 545 546 unique ecosystems, including the Succulent Karoo of SA and the Great North Woods in northern NE that contain many species of interest. Some of these coldspots also may 547 indicate areas where herbarium specimens have yet to be mobilized, providing additional 548 focus for efforts to make collection data widely available. Equally important is the need 549 550 to continue modern collecting in well-established "hotspots" so that there are multiple temporal benchmarks against which change can be measured. This is particularly true for 551 non-native invasive species that have rapidly expanding distributions and vulnerable 552 native species that have ranges that are collapsing. 553

554 Phylogenetic and trait biases can be alleviated by targeting collection efforts 555 where we know species have been under-collected. Temporal bias is more difficult to 556 address, as we cannot add to historic collections. However, we can make efforts to 557 maintain consistent regional botanical records by conducting field surveys at regular 558 intervals. Also, by linking multiple herbaria into larger digital databases, the temporal 559 biases of individual herbaria can be smoothed out to some extent.

We acknowledge that some of the biases also may be attributed to longstanding 560 curation practices at the herbariums themselves. As herbarium collections were amassed 561 for qualitative floristic, taxonomic, and systematic research, duplicate specimens of 562 563 common species and non-reproductive material have sometimes been discarded, sent elsewhere, or not accepted in the first place. This trend is becoming even more 564 565 pronounced as many herbaria around the world are increasingly constrained by funding, labor, and space. As new uses for biological collections continue to proliferate, curation 566 567 practices should also change to accommodate different avenues of research, such as climate-change biology and rare plant conservation. This will often be most effective 568 through continued collecting of specimens to overcome past biases. And most 569 importantly, researchers analyzing herbarium specimens in a widening array of studies 570 571 needed to be aware of the biases in these collections, and apply appropriate statistical 572 techniques.

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- 574

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593	References
594	Abouheif E. 1999. A method for testing the assumption of phylogenetic independence in
595	comparative data. Evolutionary Ecology Research 1: 895–909.
596	Agostinelli C, Lund U. 2013. R package 'circular': Circular Statistics (version 0.4-7).
597	URL https://r-forge.r-project.org/projects/circular/
598	APG III (Angiosperm Phylogeny Group). 2009. An update of the angiosperm
599	phylogeny group classification for the orders and families of flowering plants:
600	APG III. Botanical Journal of the Linnean Society 161: 105–121.
601	AVH. 2016. Australia's Virtual Herbarium, Council of Heads of Australasian Herbaria,
602	http://avh.chah.org.au, accessed on 09 June 2016.
603	Belasco WJ. 1979. Americans on the road, from autocamp to motel 1910-1945.
604	Baltimore: Johns Hopkins University Press.
605	Bivand RS, Pebesma E, Gomez-Rubio V. 2013. Applied spatial data analysis with R,
606	Second edition. Springer, NY. http://www.asdar-book.org/
607	Blomberg SP, Garland T, Ives AR. 2003. Testing for phylogenetic signal in
608	comparative data: behavioural traits are more labile. Evolution 57: 717–745.
609	Bond WJ, Parr CL. 2010. Beyond the forest edge: ecology, diversity and conservation
610	of the grassy biomes. Biological Conservation 143: 2395-2404.
611	Botkin BA. 1968. Automobile humor: from the horseless carriage to the compact car.
612	The Journal of Popular Culture I: 395–402.
613	Boyle B, Hopkins N, Lu Z, Garay JAR, Mozzherin D, Rees T, Matasci N, Narro ML,
614	Piel WH, Mckay SJ, et al. 2013. The taxonomic name resolution service: an
615	online tool for automated standardization of plant names. BMC Bioinformatics 14:
616	16.
617	Ceballos G, Ehrlich PR. 2006. Global mammal distributions, biodiversity hotspots, and
618	conservation. Proceedings of the National Academy of Sciences USA 103: 19374–
619	19379.
620	CIESIN. 2016. Center for International Earth Science Information Network, Columbia
621	University. Gridded Population of the World, Version 4 (GPWv4): Population
622	Density. Palisades, NY: NASA Socioeconomic Data and Applications Center
623	(SEDAC). http://dx.doi.org/10.7927/H4NP22DQ. Accessed 29 August 2016.

624	CNH. 2016. Consortium of Northeastern Herbaria. http://portal.neherbaria.org/portal/
625	Cotterill FPD, Hustler CW, Broadley DG. 1994. Systematics and biodiversity. Trends
626	in Ecology & Evolution 9: 228.
627	Cotterill FPD. 1995. Systematics, biological knowledge and environmental conservation.
628	Biodiversity and Conservation 4: 183–205.
629	Dalton R. 2003. Natural history collections in crisis as funding is slashed. <i>Nature</i> 423:
630	6940.
631	Daru BH, Elliott TL, Park DS, Davies TJ. 2017. Understanding the processes
632	underpinning patterns of phylogenetic regionalization. Trends in Ecology &
633	Evolution doi: 10.1016/j.tree.2017.08.013
634	Daru BH, Van der Bank M, Davies TJ. 2015. Spatial incongruence among hotspots
635	and complementary areas of tree diversity in southern Africa. Diversity and
636	Distributions 21: 769–780.
637	Davies TJ, Kraft NJB, Salamin N, Wolkovich EM. 2012. Incompletely resolved
638	phylogenetic trees inflate estimates of phylogenetic conservatism. Ecology 93:
639	242–247.
640	Davis CC, Willis CG, Connolly B, Kelly C, Ellison AM. 2015. Herbarium records are
641	reliable sources of phenological change driven by climate and provide novel
642	insights into species' phenological cueing mechanisms. American Journal of
643	<i>Botany</i> 102 : 1599–1609.
644	Droissart V, Hardy OJ, Sonké B, Dahdouh-Guebas F, Stévart T. 2012. Subsampling
645	herbarium collections to assess geographic diversity gradients: A case study with
646	endemic Orchidaceae and Rubiaceae in Cameroon. Biotropica 44: 44-52.
647	Edwards EJ, de Vos JM, Donoghue MJ. 2015. Doubtful pathways to cold tolerance in
648	plants. <i>Nature</i> 521 : E5–E6.
649	Edwards JL, Lane MA, Nielsen ES. 2000. Interoperability of biodiversity databases:
650	biodiversity information on every desktop. Science 289: 2312-2314.
651	Engemann K, Enquist BJ, Sandel B, Boyle B, Jørgensen PM, Morueta-Holme N,
652	Peet RK, Violle C, Svenning J-C. 2015. Limited sampling hampers "big data"
653	estimation of species richness in a tropical biodiversity hotspot. Ecology and
654	<i>Evolution</i> 5 : 807–820.

655	ESRI 1992. Environmental Systems Research Institute, Digital chart of the world, 1:1M.
656	Environmental Systems Research Institute, Inc., Redlands, California.
657	Everill PH, Primack RB, Ellwood EE, Melaas EK. 2014. Determining past leaf-out
658	times of New England's deciduous forests from herbarium specimens. American
659	Journal of Botany 101: 1–8.
660	Feeley KJ. 2012. Distributional migrations, expansions, and contractions of tropical plant
661	species as revealed in dated herbarium records. Global Change Biology 18: 1335-
662	1341.
663	Ferreira SLA, Harmse AC. 2000. Crime and tourism in South Africa: international
664	tourists perception and risk. South African Geographical Journal 82: 80-85.
665	Forman RTT, Alexander LE. 1998. Roads and their major ecological effects. Annual
666	Review of Ecology and Systematics 29 : 207–31
667	Fortune S. 1992. Voronoi diagrams and Delaunay triangulations. Computing in
668	Euclidean Geometry 1: 193–233.
669	Funk V. 2003. The importance of herbaria. Plant Science Bulletin 49: 94–95.
670	Funk VA, Morin N. 2000. A survey of the herbaria of the southeast United States. Sida,
671	Botanical Miscellany 18: 35–52.
672	Funk VA, Richardson K. 2002. Biological specimen data in biodiversity studies: use it
673	or lose it. Systematic Biology 51: 303–316.
674	GADM. 2015. Global Administrative Areas, version 2.8 (www.gadm.org).
675	Garcillán PP, Ezcurra E, Vega E. 2008. Guadalupe Island: Lost paradise recovered?
676	Overgrazing impact on extinction in a remote oceanic island as estimated through
677	accumulation functions. Biodiversity and Conservation 17: 1613–1625.
678	Garcillán PP, Ezcurra E. 2011. Sampling procedures and species estimation: Testing
679	the effectiveness of herbarium data against vegetation sampling in an oceanic
680	island. Journal of Vegetation Science 22: 273–280.
681	Geri F, Lastrucci L, Viciani D, Foggi B, Ferretti G, Maccherini S, Bonini I, Amici V,
682	Chiarucci A. 2013. Mapping patterns of ferns species richness through the use of
683	herbarium data. Biodiversity and Conservation 22: 1679–1690.

684	Gibbons JW, Scott DE, Ryan T, Buhlmann K, Tuberville T, Greene J, Mills T,
685	Leiden Y, Poppy S, Winne C et al. 2000. The global decline of reptiles, déj·vu
686	amphibians. BioScience 50: 653–666.
687	Grass A, Tremetsberger K, Hössinger R, Bernhardt K. 2014. Change of species and
688	habitat diversity in the Pannonian region of eastern Lower Austria over 170 years:
689	Using herbarium records as a witness. Natural Resources 5: 583–596.
690	Greve M, Lykke AM, Fagg CW, Gereau RE, Lewis GP, Marchant R, Marshall AR,
691	Ndayishimiye J, Bogaert J, Svenning JC. 2016. Realising the potential of
692	herbarium records for conservation biology. South African Journal of Botany 105:
693	317–323.
694	Griffith EH, Sauer JR, Royle JA. 2010. Traffic effects on bird counts on North
695	American breeding bird survey routes. Auk 127: 387–393.
696	Gutzwiller KJ, Flather CH. 2011. Wetland features and landscape context predict the
697	risk of wetland habitat loss. Ecological Applications 21: 968–982
698	Hart R, Salick J, Ranjitkar S, Xu J. 2014. Herbarium specimens show contrasting
699	phenological responses to Himalayan climate. Proceedings of the National
700	Academy of Sciences USA 111: 10615–10619.
701	Hijmans RJ, Garrett KA, Huaman Z, Zhang DP, Schreuder M, Bonierbale M. 2000.
702	Assessing the geographic representation of genebank collections: the case of the
703	Bolivian wild potatoes. Conservation Biology 14: 1755-1765.
704	Hijmans RJ. 2015. geosphere: Spherical Trigonometry. R package version 1.4-3.
705	http://CRAN.R-project.org/package=geosphere
706	Hortal J, Lobo JM, Jiménez-Valverde A. 2007. Limitations of biodiversity databases:
707	case study on seed-plant diversity in tenerife, canary islands. Conservation
708	<i>Biology</i> 21 : 853–863.
709	Hui C, Shuang-cheng L, Yi-li Z. 2003. Impact of road construction on vegetation
710	alongside Qinghai-Xizang highway and railway. Chinese Geographical Science
711	13 : 340–346.
712	Isaac NJ, Turvey ST, Collen B, Waterman C, Baillie JE. 2007. Mammals on the
713	EDGE: conservation priorities based on threat and phylogeny. PLoS ONE 2: e296.

714	Iwanycki N. 2009. Guidelines for collecting herbarium specimens of vascular plants.
715	Royal Botanical Gardens Canada, Hamilton, Canada.
716	Jombart T, Dray S. 2008. adephylo: exploratory analyses for the phylogenetic
717	comparative method. <i>Bioinformatics</i> 26: 1907–1909.
718	Klemens MW, Thorbjarnarson JB. 1995. Reptiles as a food resource. Biodiversity and
719	<i>Conservation</i> 4 : 281–298.
720	Lavoie C. 2013. Biological collections in an ever changing world: Herbaria as tools for
721	biogeographical and environmental studies. Perspectives in Plant Ecology,
722	Evolution and Systematics 15: 68–76.
723	le Roux MM, Wilkin P, Balkwill K, Boatwright JS, Bytebier B, Filer D, Klak C,
724	Klopper RR, Koekemoer M, Livermore L et al. 2017. Producing a plant
725	diversity portal for South Africa. Taxon 66: 421-431.
726	Lees DC, Lack HW, Rougerie R, Hernandez-Lopez A, Raus T, Avtzis ND, Augustin
727	S, Lopez-Vaamonde C. 2011. Tracking origins of invasive herbivores through
728	herbaria and archival DNA: the case of the horse-chestnut leaf miner. Frontiers in
729	Ecology and the Environment 9: 322–328.
730	Lemanski C. 2004. A new apartheid? The spatial implications of fear of crime in Cape
731	Town, South Africa. Environment & Urbanization 16: 101-111.
732	Leuner B. 2007. Migration, multiculturalism and language maintenance in Australia.
733	Peter Lang, Oxford.
734	Li Y, Yu J, Ning K, Du S, Han G, Qu F, Wang G, Fu Y, Zhan C. 2014. Ecological
735	effects of roads on the plant diversity of coastal wetland in the Yellow River Delta.
736	The Scientific World Journal 2014: 952051.
737	McCarthy KP, Fletcher JR RJ, Rota CT, Hutto RL. 2012. Predicting species
738	distributions from samples collected along roadsides. Conservation Biology 26:
739	68–77.
740	Meyer C, Weigelt P, Kreft H. 2016. Multidimensional biases, gaps and uncertainties in
741	global plant occurrence information. Ecology Letters 19: 992-1006.
742	Miller-Rushing A, Primack R, Mukunda S. 2006. Photographs and herbarium
743	specimens as tools to document phenological changes in response to global
744	warming. American Journal of Botany 93: 1667–1674.

745	Minteer BA, Collins JP, Love KE, Puschendorf R. 2014. Avoiding (Re)Extinction".
746	<i>Science</i> 344 : 260-261.
747	Moerman DE, Estabrook GF. 2006. The botanist effect: counties with maximal species
748	richness tend to be home to universities and botanists. Journal of Biogeography
749	33 : 1969–1974.
750	Musgrave T, Gardner C, Musgrave W. 1999. The plant hunters. Two hundred years of
751	adventure and discovery. Seven Dials, London, UK.
752	NEBC. 1899. Editorial announcement. Rhodora 1: 1-2
753	Newbold T. 2010. Applications and limitations of museum data for conservation and
754	ecology, with particular attention to species distribution models. Progress in
755	Physical Geography 34 : 3–22.
756	Norris WR, Lewis DQ, Widrlechner MP, Thompson JD, Pope RO. 2001. Lessons
757	from an inventory of the Ames, Iowa, flora (1859-2000). Journal of the Iowa
758	Academy of Science 108: 34–63.
759	Nualart N, Ibáñez N, Soriano I, López-Pujol J. 2017. Assessing the relevance of
760	herbarium collections as tools for conservation biology. Botanical Review 83:
761	303–325.
762	Orme CD, Davies RG, Burgess M, Eigenbrod F, Pickup N, Olson VA, Webster AJ,
763	Ding TS, Rasmussen PC, Ridgely RS, et al. 2005. Global hotspots of species
764	richness are not congruent with endemism or threat. Nature 436: 1016–1019.
765	Orme D, Freckleton R, Thomas G, Petzoldt T, Fritz S, Isaac N, Pearse W. 2012.
766	caper: Comparative Analyses of Phylogenetics and Evolution in R. R package
767	version 0.5.http://CRAN.R-project.org/package = caper.
768	Pagel M. 1999. Inferring the historical patterns of biological evolution. <i>Nature</i> 401: 877–
769	884.
770	Palmer MW, Earls PG, Hoagland BW, White PS, Wohlgemuth T 2002. Quantitative
771	tools for perfecting species list. Environmetrics 13: 121–137.
772	Pautasso M, McKinney ML. 2007. The Botanist Effect Revisited: Plant Species
773	Richness, County Area, and Human Population Size in the United States.
774	Conservation Biology 21: 1333–1340.

775	Poethig, RS. 2013. Vegetative phase change and shoot maturation in plants. Current
776	Topics in Developmental Biology 105: 125–152.
777	Prather LA, Alvarez-Fuentes O, Mayfield MH, Ferguson CJ. 2004a. The decline of
778	plant collecting in the United States: a threat to the infrastructure of biodiversity
779	studies. Systematic Botany 29: 15–28.
780	Prather LA, Alvarez-Fuentes O, Mayfield MH, Ferguson CJ. 2004b. Implications of
781	the decline in plant collecting for systematic and floristic research. Systematic
782	<i>Botany</i> 29 : 216–220.
783	Price CA. 1998. Post-war immigration: 1945-1998. Journal of the Australian Population
784	Association 15: 17.
785	Pritchard PCH. 1996. The Galápagos tortoises: nomenclatural and survival status.
786	Chelonian Research Foundation in association with Conservation International
787	and Chelonia Institute, Lunenburg, MA.
788	Pyke GH, Ehrlich PR. 2010. Biological collections and ecological/environmental
789	research: a review, some observations and a look to the future. Biological Reviews
790	5 : 247–266.
791	Redding DW, Mooers AØ. 2006. Incorporating evolutionary measures into conservation
792	prioritization. Conservation Biology 20: 1670–1678.
793	Revell LJ. 2012. phytools: An R package for phylogenetic comparative biology (and
794	other things). Methods in Ecology and Evolution 3: 217–223.
795	Rich TCG, Woodruff ER. 1992. Recording bias in botanical surveys. Watsonia 19: 73-
796	95.
797	Rivers-Moore NA, Cowden C. 2012. Regional prediction of wetland degradation in
798	South Africa. Wetlands Ecology and Management 20: 491-502.
799	Robinson JG. 2001. Using 'sustainable use' approaches to conserve exploited
800	populations. In: Reynolds JD, Mace GM, Redford KH, Robinson JG, eds.
801	Conservation of exploited species. Cambridge: Cambridge University Press, 485-
802	498.
803	Rudin SM, Murray DW, Whitfeld TJS. 2017. Retrospective analysis of heavy metal
804	contamination in Rhode Island based on old and new herbarium specimens.
805	Applications in Plant Sciences 5: 1–13.

806	SANBI. 2016. South African National Biodiversity Institute. Botanical Database of
807	Southern Africa (BODATSA), http://newposa.sanbi.org/, accessed on 22 July
808	2016.
809	Schaefer H, Hardy OJ, Silva L, Barraclough TG, Savolainen V. 2011. Testing
810	Darwin's naturalization hypothesis in the Azores. Ecology Letters 14: 389-396.
811	Schmidt-Lebuhn AN, Knerr NJ, Kessler M. 2013. Non-geographic collecting biases in
812	herbarium specimens of Australian daisies (Asteraceae). Biodiversity and
813	<i>Conservation</i> 22 : 905–919.
814	Schorn C, Weber E, Bernardos R, Hopkins C, Davis CC. 2016. The New England
815	Vascular Plants Project: 295,000 specimens and counting. Rhodora 118: 324-325.
816	Staats M, Erkens RHJ, van de Vossenberg B, Wieringa JJ, Kraaijeveld K, Stielow B,
817	Geml J, Richardson JE, Bakker FT. 2013. Genomic treasure troves: complete
818	genome sequencing of herbarium and insect museum specimens. PLoS ONE 8:
819	e69189.
820	Stropp J, Ladle RJ, Malhado ACM, Hortal J, Gaffuri J, Temperley, WH, Skøien JO.
821	Mayaux, P. 2016. Mapping ignorance: 300 years of collecting flowering plants in
822	Africa. Global Ecology and Biogeography 25: 1085–1096.
823	Syfert MM, Smith MJ, Coomes DA. 2013. The effects of sampling bias and model
824	complexity on the predictive performance of MaxEnt species distribution models.
825	<i>PLoS ONE</i> 8 : e55158.
826	Thiers B. 2016. Index Herbariorum: A global directory of public herbaria and
827	associated staff. New York Botanical Garden's Virtual Herbarium.
828	http://sweetgum.nybg.org/science/ih/, accessed on 29 September 2016.
829	Tobler M, Honorio E, Janovec J, Reynel C. 2007. Implications of collection patterns of
830	botanical specimens on their usefulness for conservation planning: an example of
831	two neotropical plant families (Moraceae and Myristicaceae) in Peru. Biodiversity
832	and Conservation 16: 659–677
833	van der Schoot C, Paul LK, Rinne PLH. 2014. The embryonic shoot: a lifeline through
834	winter. Journal of Experimental Botany 65: 1699–1712.

835	Webb CO, Ackerly DD, Kembel SW. 2008. PHYLOCOM: software for the analysis of
836	phylogenetic community structure and trait evolution. Bioinformatics 24: 2098-
837	2100.
838	Webb CO, Ackerly DD, McPeek MA, Donoghue MJ. 2002. Phylogenies and
839	community ecology. Annual Review of Ecology and Systematics 33: 475–505.
840	Webb CO, Donoghue MJ. 2005. Phylomatic: tree assembly for applied phylogenetics.
841	Molecular Ecology Notes 5: 181–183.
842	Whittle T. 1970. The Plant Hunters. Heinemann, London.
843	Willis CG, Ellwood ER, Primack RB, Davis CC, Pearson KD, Gallinato AS, Yost
844	JM, Nelson G, Mazer SJ, Rossington NL et al. 2017a. Old plants, new tricks:
845	phenological research using herbarium specimens. Trends in Ecology & Evolution
846	32 : 531–546.
847	Willis CG, Law E, Williams AC, Franzone BF, Bernardos R, Brun L, Hopkins C,
848	Schorn C, Weber E, Parks DS et al. 2017b. CrowdCurio: an online
849	crowdsourcing platform to facilitate climate change studies using herbarium
850	specimens. New Phytologist 215: 479-488.
851	Wolf A, Anderegg WRL, Ryan SJ, Christensen J. 2011. Robust detection of plant
852	species distribution shifts under biased sampling regimes. Ecosphere 2: 115.
853	Wolkovich EM, Davies TJ, Schaefer H, Cleland EE, Cook BI, Travers SE, Willis
854	CG, Davis CC. 2013. Temperature-dependent shifts in phenology contribute to
855	the success of exotic species with climate change. American Journal of Botany
856	100 : 1407–1421.
857	Yessoufou K, Daru BH, Davies TJ. 2012. Phylogenetic patterns of extinction risk in the
858	Eastern Arc ecosystems, an African biodiversity hotspot. PLoS ONE 7: e47082.
859	Zanne AE, Tank DC, Cornwell WK, Eastman JM, Smith SA, FitzJohn RG,
860	McGlinn DJ, O'Meara BC, Moles AT, Reich PB, et al. 2014. Three keys to the
861	radiation of angiosperms into freezing environments. Nature 506: 89-92.
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864 Supporting Information

Additional supporting information may be found in the online version of this article.

866

- **Fig. S1** Analytical workflow representing the different steps in the development of this
- study from data compilation, collation, to statistical analysis.

869

- Fig. S2 Relationships of the number of specimens collected per species with number of
- species collected in each flora for Australia (left), South Africa (middle), and New
- 872 England of the USA (right).

873

Table S1 Results of the tests of phylogenetic signal in the number of specimens collected

875 per species.

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877

LEGENDS TO FIGURES 878 Fig. 1 Spatial bias in herbarium collections. Geographic distribution of herbarium 879 880 collecting activity depicting the spatial variation in sampling effort using Delaunay polygon tiles for (a) Australia (857,245 locales), (b) South Africa (n = 61,130 locales), 881 and (c) New England (n = 130, 374 locales). Hotspots (red) and coldspots (blue) of 882 herbarium sampling within quarter degree grids for (d) Australia, (e) South Africa and (f) 883 New England. The hotspots and coldspots are the top and lowest 2.5% quantiles 884 respectively of the number of specimens per locale. 885 886 Fig. 2 Comparison of geographic sampling bias of herbarium records in relation to (a) the 887 minimum distance to roads; (b) minimum distance to herbaria; and (c) regional altitudes 888 at sampling locales. Black lines in (a) and (b) correspond to sampling locales and red 889 890 indicates an equal number of random points generated 1000 times. Dark grey shading in (c) corresponds to sampling locales in relation to the regional altitudes, *i.e.*, all other 891 altitudes (in red) for all three floras, Australia (left), South Africa (middle) and New 892 England (right). Dotted line in (c) indicates altitude at 500 m above sea level. 893 894 Fig. 3 Timeline of herbarium specimen collection density in relation to major historical 895 events in time (indicated in red text) for the three floras: Australia, South Africa and New 896 England. Analysis of phylogenetic structure through time by binning sequences of 897 collection dates into decades and testing for overdispersion vs. clustering, are indicated in 898 899 black font. The red trend line indicates the gross domestic product of each region. NRI, 900 net relatedness index; NTI, nearest taxon index. 901 902 Fig. 4 Temporal biases in herbarium collections. (a) Comparison of density plots of collection dates by seasons of the year of herbarium records (blue line) with the dates 903 904 spanning the entire duration of collection (red line); blue lines outside the red lines 905 indicate over-collecting at a particular time of year, and (b) Distribution of collection 906 dates by days of the week for the three floras. Australia (n = 4,579,321 collection dates), 907 South Africa (n = 771,991 collection dates), and New England (n = 562,587 collection dates). 908

909	
910	Fig. 5 Assessment of bias in plant traits: (a) growth duration; (b) growth form; (c) height;
911	and (d) extinction risk for the floras of Australia (left pane), South Africa (middle pane)
912	and New England (right pane). Error bars in (a), (b), and (d) represent +/- SE.
913	
914	Fig. 6 Distribution of phylogenetic bias, the tendency of closely related species to be
915	similarly collected in herbarium records for three floras: (a) Australia; (b) South Africa;
916	and (c) New England. Collecting effort is not phylogenetically random, but tends to be
917	clustered in few selected lineages. The color scales correspond to log_{10} numbers of
918	specimens per species and ranges from red (low number of specimens per species) to blue
919	(high number of specimens per species).
920	
921	Fig. 7 Phylogenetic bias in collection frequency for exemplar families in New England
922	flora. Phylogenetic bias is indicated by significant phylogenetic signal in at least one of
923	three metrics (Abouheif's C_{mean} , Blomberg's K and Pagel's λ). The color bar illustrates
924	values within families: log_{10} numbers of specimens per species and ranges from red (low
925	number of specimens per species) to blue (high number of specimens per species). **P $<$
926	0.001; *P < 0.01; NS P > 0.05
927	
928	Fig. 8 Collector bias in herbarium collections. The number of herbarium specimens
929	amassed per collector for three regional floras in (a) Australia; (b) South Africa; and (c)
930	New England. The top five collectors in each flora are highlighted in red. Numbers
931	within parentheses correspond to lifespans of the collectors, with collectors that have died
932	highlighted in red and currently living ones in black.
933	

934

Table 1. Model coefficients for multiple regressions of collecting effort in the number of specimens collected per locality.

AUSTRALIA	Predictors (log ₁₀ -	Percentage of	P values	Model adjusted	Model slope	Model
	transformed)	variance		R^2		intercept
		explained				
		(%)				
	Distance to roads	0.14	0.001	0.4571	-0.02	11.45
	Distance to herbaria	45.03	0.001	•	-0.89	
	Human population	0.50	0.001		0.11	
	density					
	Altitude	0.041	0.001		-0.046	
SOUTH AFRICA	Predictors (log ₁₀ -	Percentage of	P values	Model adjusted	Model slope	Model
	transformed)	variance		R^2		intercept
		explained				
		(%)				
	Distance to roads	0.00001	0.0003	0.3075	-0.011	11.33
	Distance to herbaria	29.13	0.001		-0.73	
	Human population	0.0009	0.001	1	-0.03	
	density					

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	Altitude	1.62	0.001		-0.15	
NEW ENGLAND	Predictors (log ₁₀ -	Percentage of	P values	Model adjusted	Model slope	Model
	transformed)	variance		R^2		intercept
		explained				
		(%)				
	Distance to roads	0.07	0.0009	0.17	0.13	7.03
	Distance to herbaria	12.3	0.001	•	-0.87	
	Human population	4.68	0.001	•	0.30	
	density					
	Altitude	0.04	0.001		0.046	

936

937 938

- **Table 2**. Results of the tests of phylogenetic signal in the number of specimens collected per species using three methods (Abouheif's
- 940 C_{mean}, Blomberg's K and Pagel's λ). Phylogenetic data is derived from Zanne *et al.* (2014). All tests are based on 1000 randomizations.
- 941 **P < 0.001; ^{NS}P > 0.05

	Australia (n = 5814 species)	South Africa (n = 3568	New England ($n = 4269$	
		species)	species)	
Abouheif's C _{mean}	0.12**	0.15**	0.12**	
Blomberg's K	0.00085^{NS}	0.0013 ^{NS}	0.0030 ^{NS}	
Pagel's λ	0.18**	0.32**	0.29**	

942

943

944

945 **Table 3** Multiple regressions of phylogenetic generalized least squares of collecting effort (frequency) of herbarium specimens with

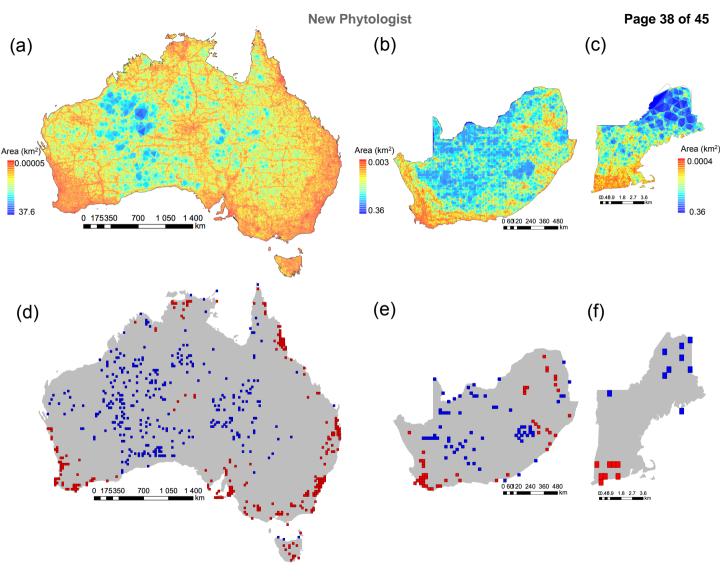
946 phylogenetic metrics of species uniqueness. BL, terminal branch length; ED, evolutionary distinctiveness; EDGE, evolutionary

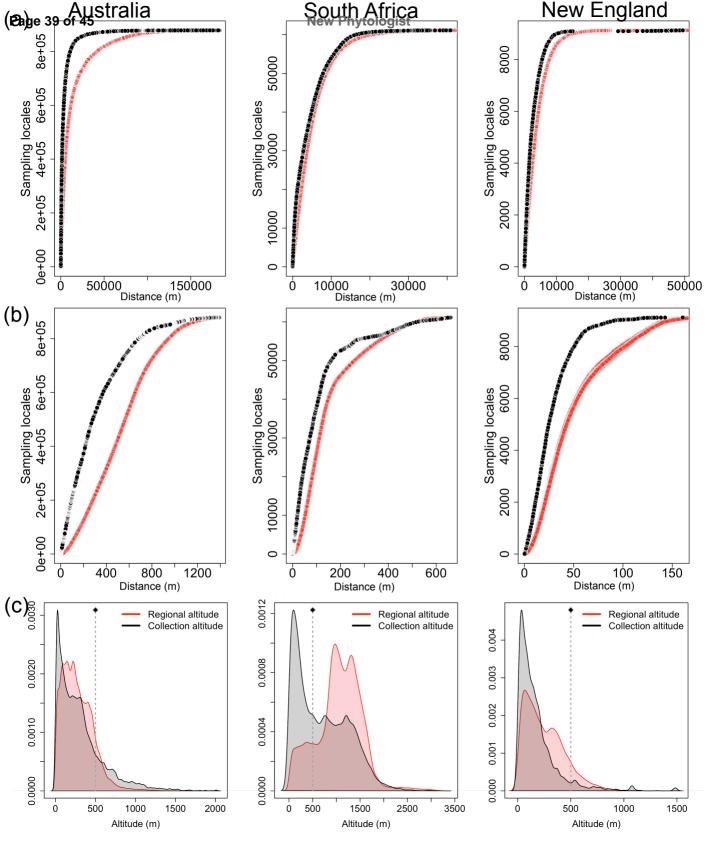
947 distinctiveness and global endangerment.

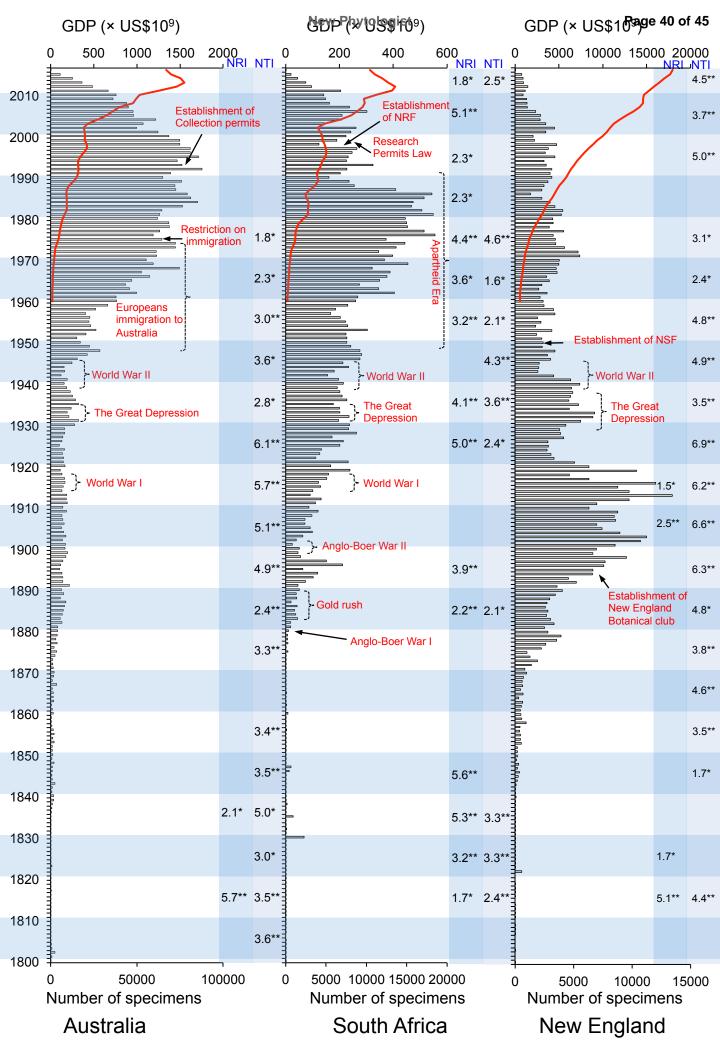
	Percentage of	P values	Model	Model slope	Model	
transformed)	variance		adjusted R ²		intercept	
	explained (%)					
BL	1.36	0.7	0.049	0.035	4.37	
ED	0.2	0.008		0.44		
EDGE	3.75	< 0.001		-1.23		
Predictors (log ₁₀ -	Percentage of	P values	Model	Model slope	Model	
transformed)	variance		adjusted R ²		intercept	
	explained (%)					
BL	0.47	0.3	0.09	-0.063	3.63	
ED	0.000015	0.001		0.63		
EDGE	8.89	< 0.001		-1.3		
Predictors (log ₁₀ -	Percentage of	P values	Model	Model slope	Model	
transformed)	variance		adjusted R ²		intercept	
-	BL ED EDGE Predictors (log ₁₀ - transformed) BL ED EDGE Predictors (log ₁₀ -	BL1.36ED0.2EDGE3.75Predictors (log10- transformed)Percentage of variance explained (%)BL0.47ED0.000015EDGE8.89Predictors (log10- Percentage of Percentage of Percentage of Percentage of Percentage of Percentage of Percentage of	explained (%) explained (%) BL 1.36 0.7 ED 0.2 0.008 EDGE 3.75 <0.001	explained (%) explained (%) BL 1.36 0.7 0.049 ED 0.2 0.008 0.001 EDGE 3.75 <0.001 Predictors (log ₁₀ - transformed) Percentage of variance explained (%) P values Model adjusted R ² BL 0.47 0.3 0.09 EDGE 8.89 <0.001 EDGE 8.89 <0.001	$\begin{tabular}{ c c c c c } \hline $explained (\%)$ & $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $	

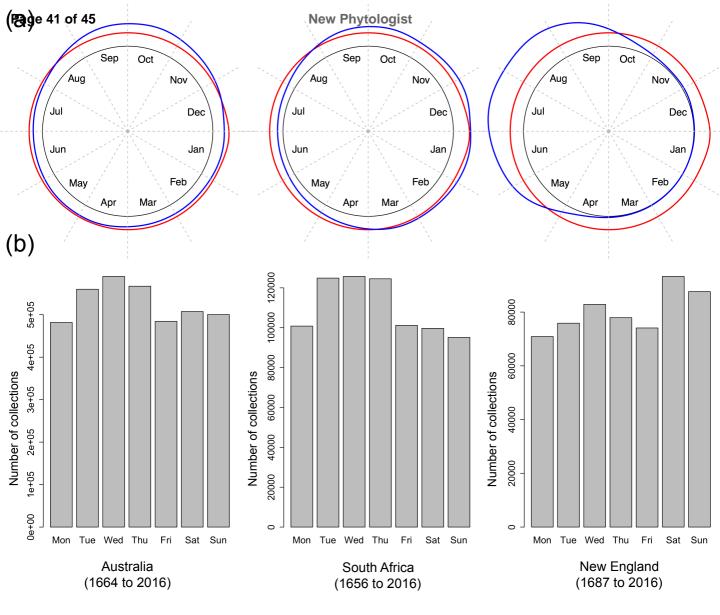
	explained (%)				
BL	0.09	0.94	1.70E-02	-0.0052	3.89
ED	0.054	0.0045		0.79	
EDGE	1.87	< 0.001		-2.28	

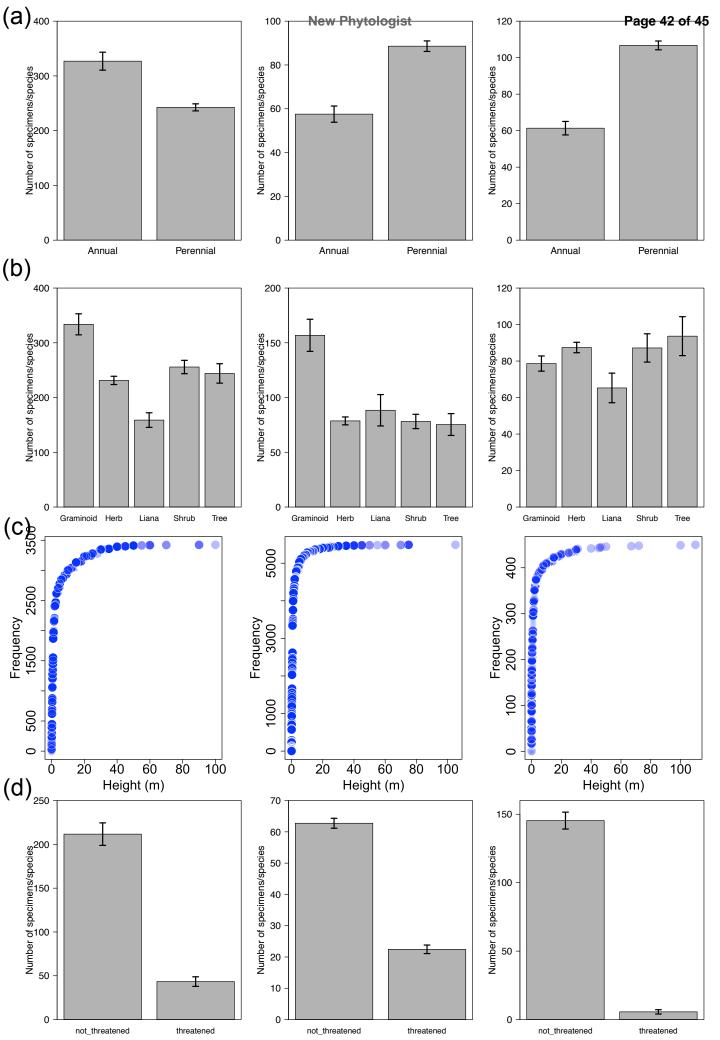
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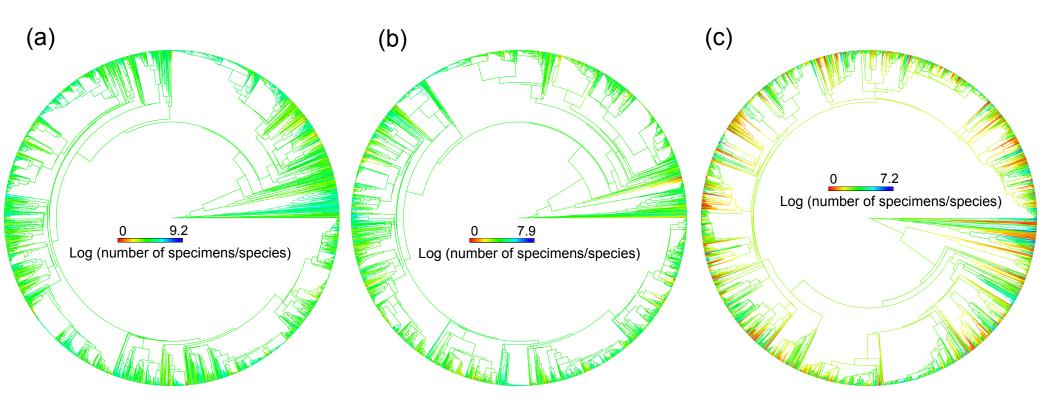


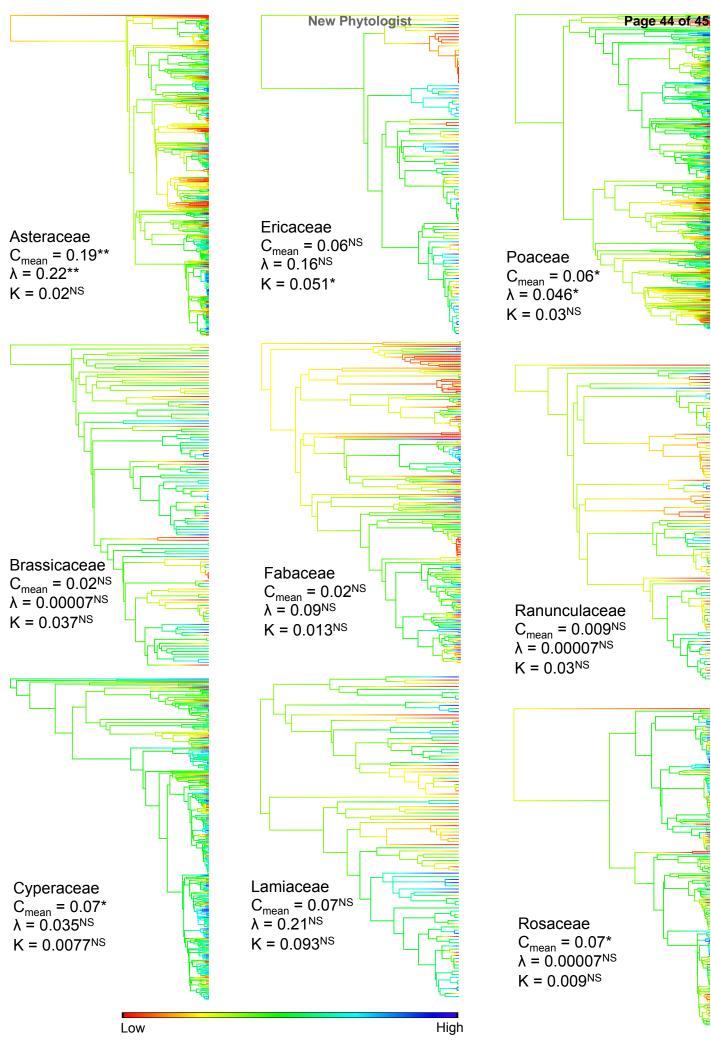


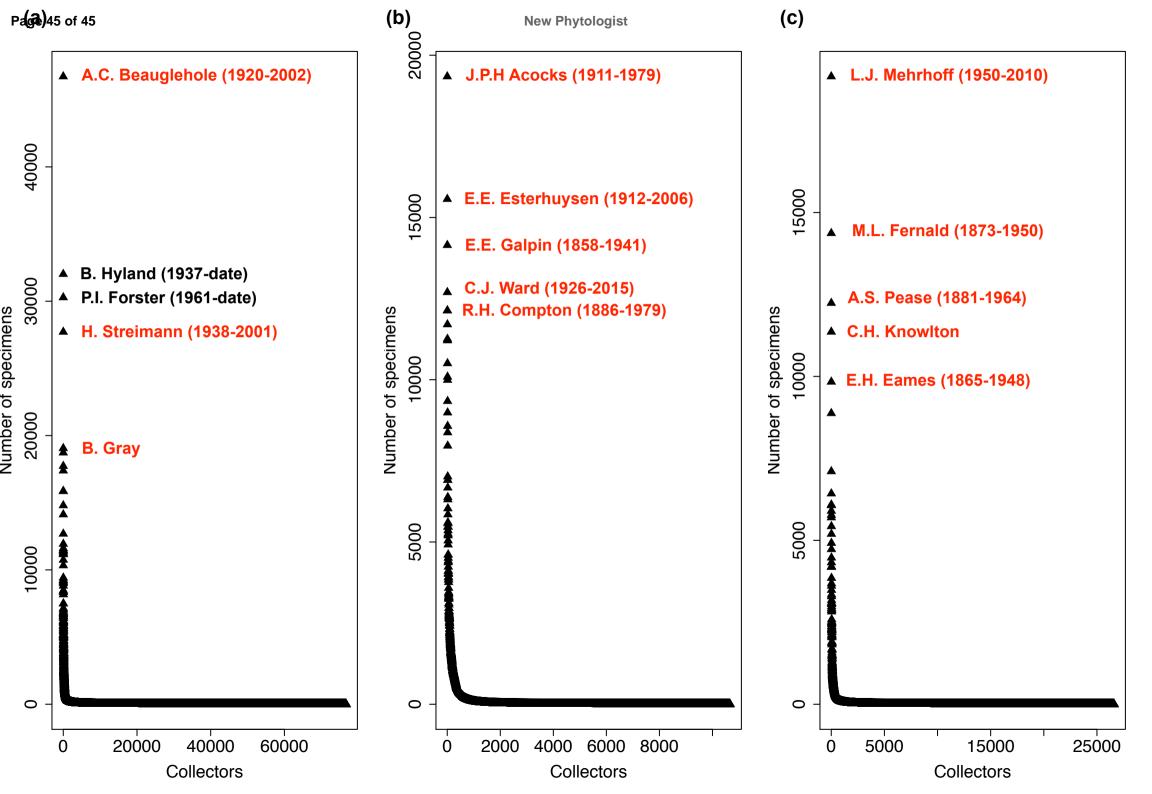












New Phytologist Supporting Information Figs S1 & S2 and Table S1 Article title: **Widespread sampling biases in herbaria revealed from large-scale digitization**

Authors: Barnabas H. Daru, Daniel S. Park, Richard B. Primack, Charles G. Willis, David S. Barrington, Timothy J. S. Whitfeld, Tristram G. Seidler, Patrick W. Sweeney, David R. Foster, Aaron M. Ellison and Charles C. Davis Article acceptance date: 18 September 2017

The following Supporting Information is available for this article:

Fig. S1 Analytical workflow representing the different steps in the development of this study from data compilation, collation, to statistical analysis.

Fig. S2 Relationships of the number of specimens collected per species with number of species collected in each flora for Australia, South Africa, and New England of the USA.

Table S1 Results of the tests of phylogenetic signal in the number of specimens collected

 per species using three methods for nine exemplar clades in New England.

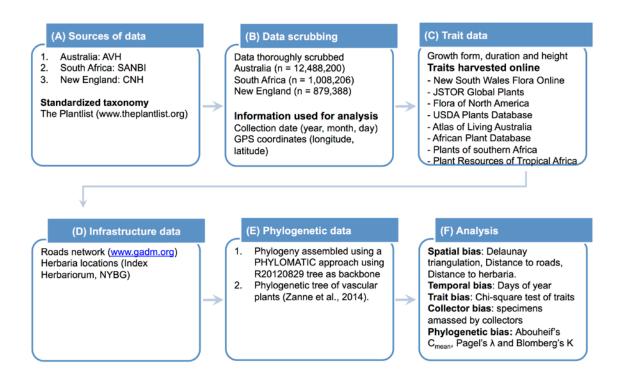


Fig. S1 Analytical workflow representing the different steps in the development of this study from data compilation, collation, to statistical analysis.

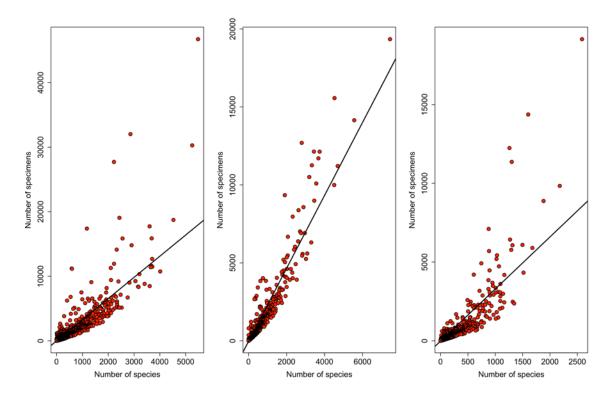


Fig. S2 Relationships of the number of specimens collected per species with number of species collected in each flora for Australia (left), South Africa (middle), and New England of the USA (right).

Table S1 Results of the tests of phylogenetic signal in the number of specimens collected per species using three methods (Abouheif's C_{mean} , Blomberg's K and Pagel's λ) for nine exemplar clades in New England: Asteraceae, Brassicaceae, Cyperaceae, Ericaceae, Fabaceae, Lamiaceae, Poaceae, Ranunculaceae, and Rosaceae. Phylogenetic data is derived from Phylomatic (Webb & Donoghue 2005). All tests are based on 1000 randomizations. **P < 0.001; *P < 0.01; NS, P > 0.05

	Asteraceae	Brassicaceae	Cyperaceae	Ericaceae	Fabaceae	Lamiaceae	Poaceae (n	Ranunculaceae	Rosaceae
	(n = 593	(n = 146)	(n = 518)	(n = 158)	(n = 255	(n = 146)	= 565	(n = 169)	(n = 346
	species)				species)		species)		species)
Abouheif's	0.11**	-0.055 ^{NS}	0.016 ^{NS}	0.15*	0.028 ^{NS}	-0.014 ^{NS}	0.0026 ^{NS}	-0.04 ^{NS}	0.098*
C _{mean}									
Blomberg's	0.11 ^{NS}	0.57 ^{NS}	0.55 ^{NS}	0.42 ^{NS}	0.092 ^{NS}	0.56 ^{NS}	0.20*	0.093 ^{NS}	0.17 ^{NS}
К									
Pagel's	0.14**	0.00006 ^{NS}	0.053*	0.27 ^{NS}	0.23*	0.02 ^{NS}	0.00008 ^{NS}	0.00007 ^{NS}	0.19 ^{NS}
lambda									

References to Supporting Information

- Webb CO, Donoghue MJ. 2005. Phylomatic: tree assembly for applied phylogenetics. *Molecular Ecology Notes* 5: 181–183.
- Zanne AE, Tank DC, Cornwell WK, Eastman JM, Smith SA, FitzJohn RG, McGlinn DJ, O'Meara BC, Moles AT, Reich PB, et al. 2014. Three keys to the radiation of angiosperms into freezing environments. Nature 506: 89–92.