



# Widespread sampling biases in herbaria revealed from large-scale digitization

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1 2	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 2.5<sup>th</sup> 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<sub>mean</sub> 224 (Abouheif, 1999), Blomberg's K (Blomberg *et al.*, 2003), and Pagel's lambda ( $\lambda$ ) (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  $\lambda$ 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  $\lambda$  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  $\lambda$  were calculated using the R package *phytools* (Revell, 2012). Abouheif's C<sub>mean</sub> 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 20<sup>th</sup> 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  $\sim 100$  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

#### **New Phytologist**

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 $\lambda$ = 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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#### 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 $\lambda$ ). 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 <sub>10</sub> -	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 <sub>10</sub> -	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 <sub>10</sub> -	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	

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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<sub>mean</sub>, 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; <sup>NS</sup>P > 0.05

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

942

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944

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

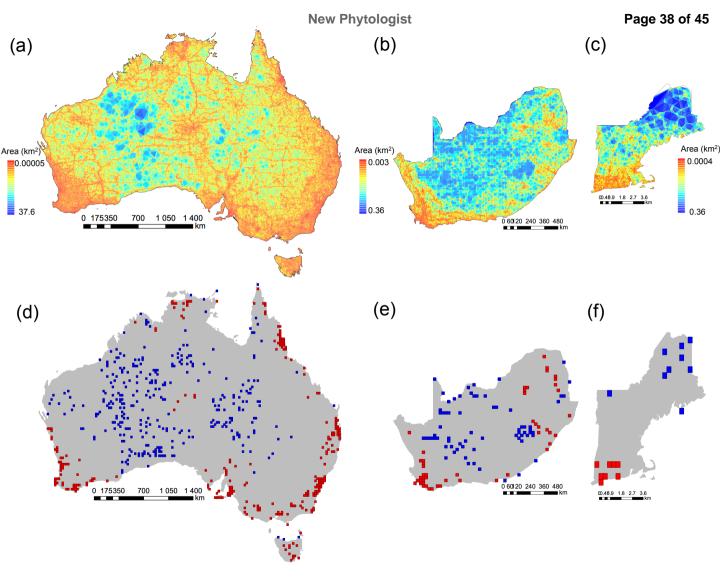
946 phylogenetic metrics of species uniqueness. BL, terminal branch length; ED, evolutionary distinctiveness; EDGE, evolutionary

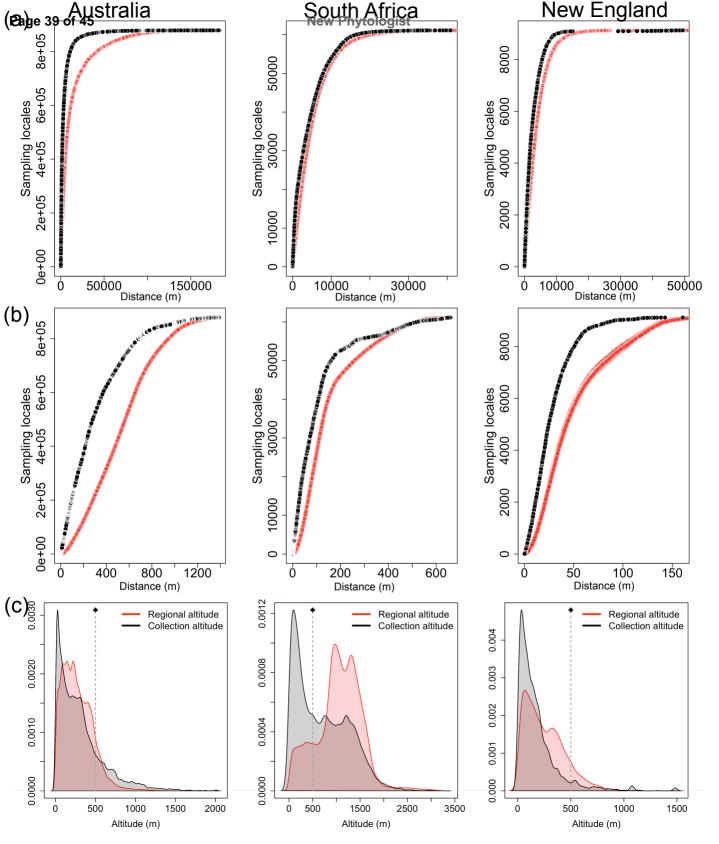
947 distinctiveness and global endangerment.

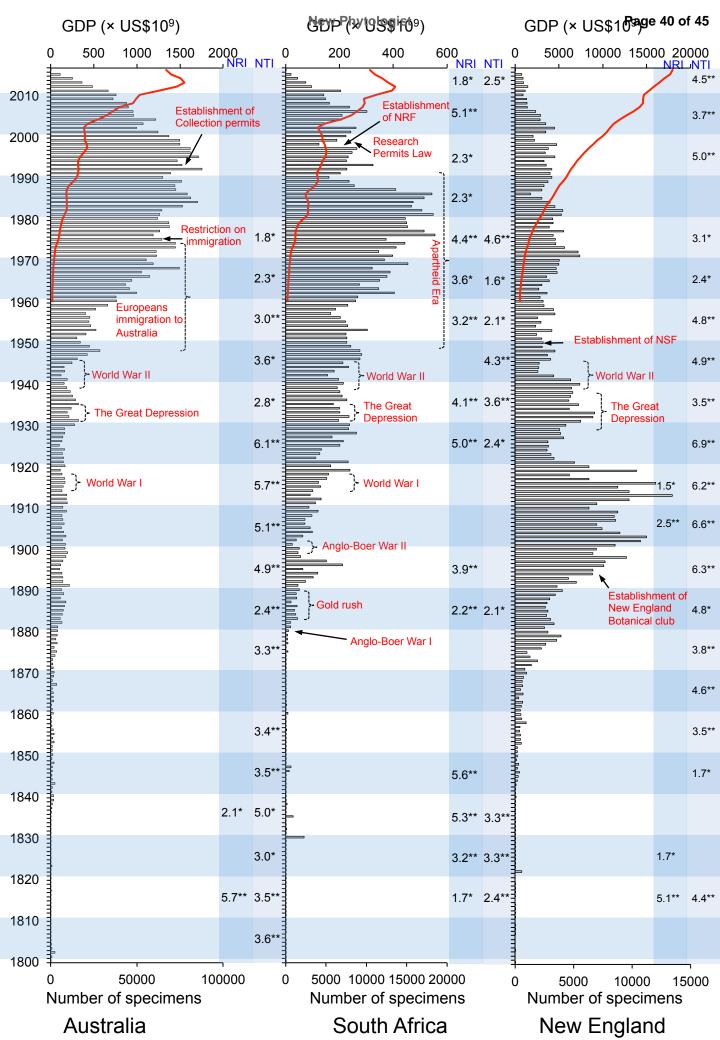
	Percentage of	P values	Model	Model slope	Model	
transformed)	variance		adjusted R <sup>2</sup>		intercept	
	explained (%)					
BL	1.36	0.7	0.049	0.035	4.37	
ED	0.2	0.008		0.44		
EDGE	3.75	< 0.001		-1.23		
Predictors (log <sub>10</sub> -	Percentage of	P values	Model	Model slope	Model	
transformed)	variance		adjusted R <sup>2</sup>		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 <sub>10</sub> -	Percentage of	P values	Model	Model slope	Model	
transformed)	variance		adjusted R <sup>2</sup>		intercept	
-	BL ED EDGE Predictors (log <sub>10</sub> - transformed) BL ED EDGE Predictors (log <sub>10</sub> -	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 <sub>10</sub> - transformed)       Percentage of variance explained (%)       P values       Model adjusted R <sup>2</sup> 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 $explained (\%)$ & $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $$ $	

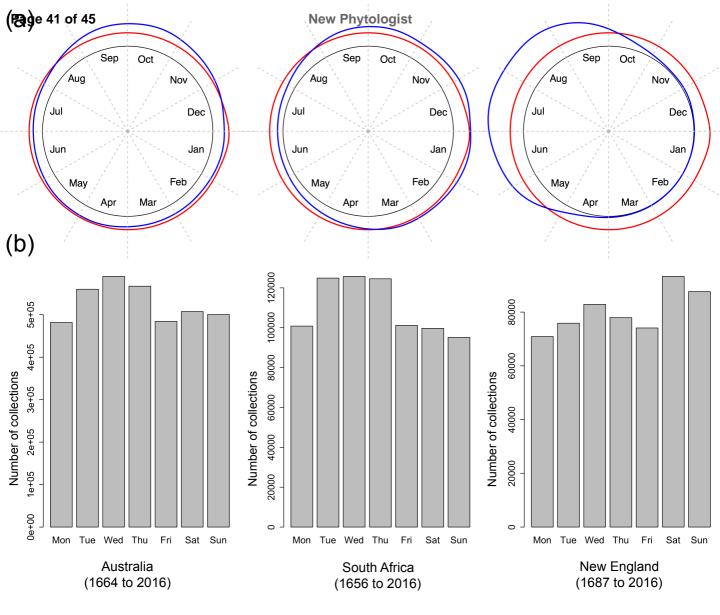
	explained (%)				
BL	0.09	0.94	1.70E-02	-0.0052	3.89
ED	0.054	0.0045		0.79	
EDGE	1.87	< 0.001		-2.28	

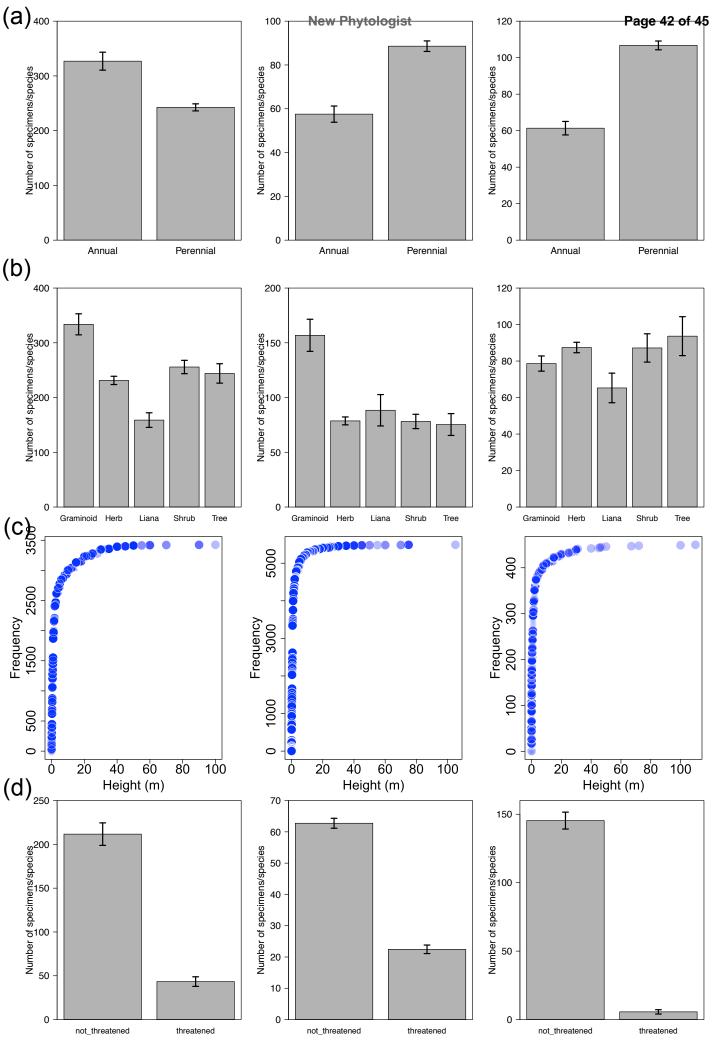
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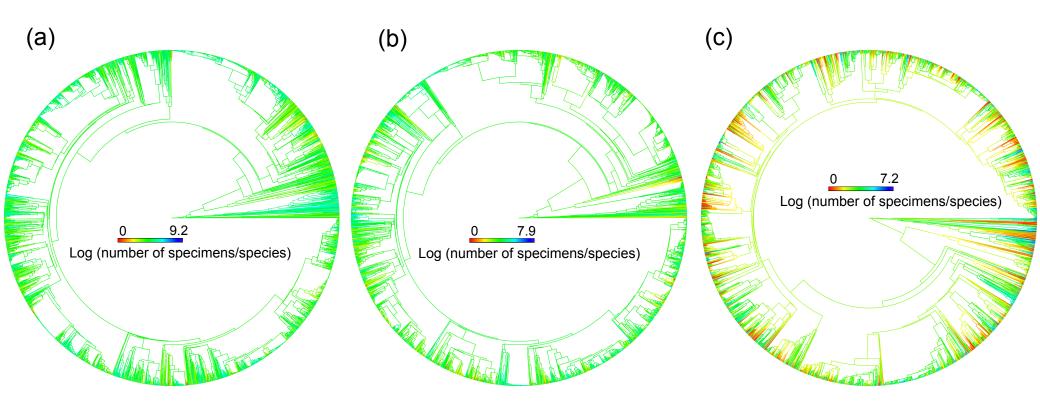


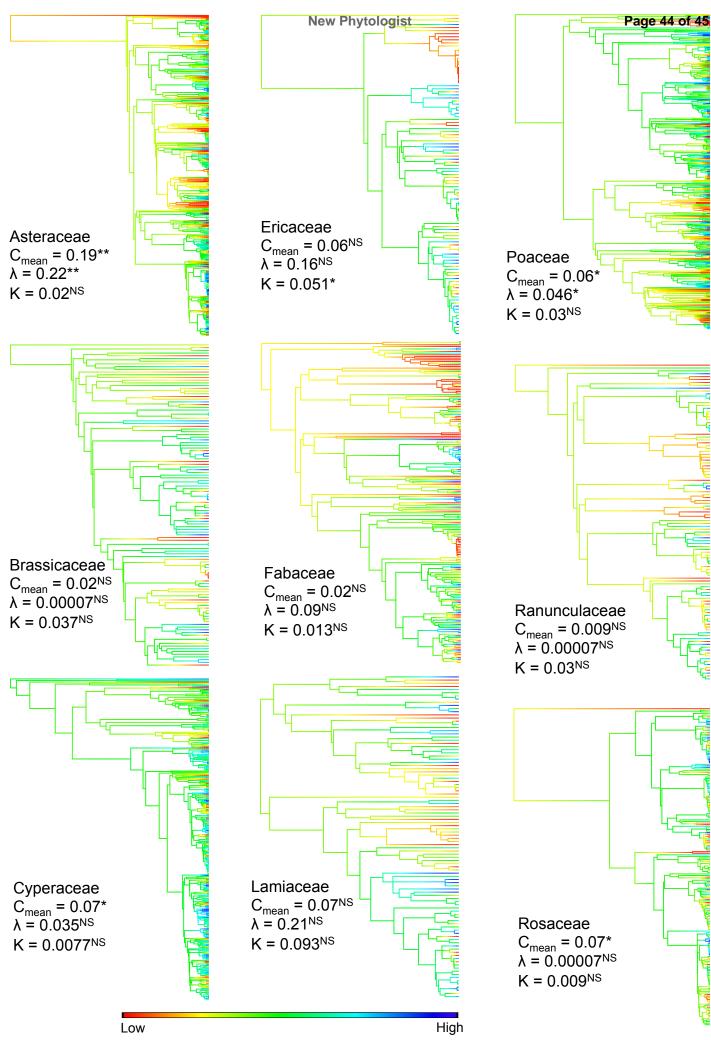


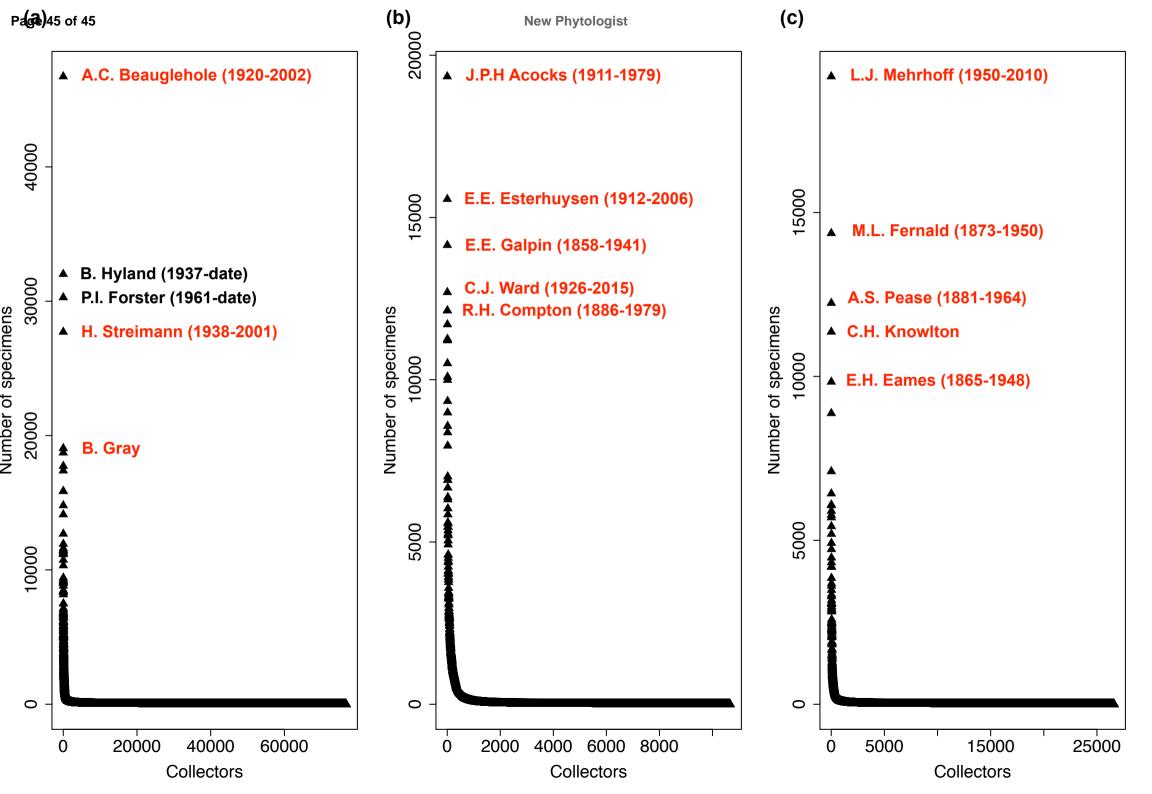












New Phytologist Supporting Information Figs S1 & S2 and Table S1 Article title: **Widespread sampling biases in herbaria revealed from large-scale digitization** 

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

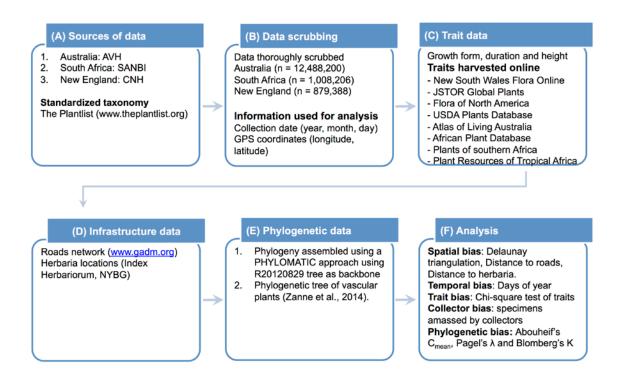
The following Supporting Information is available for this article:

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

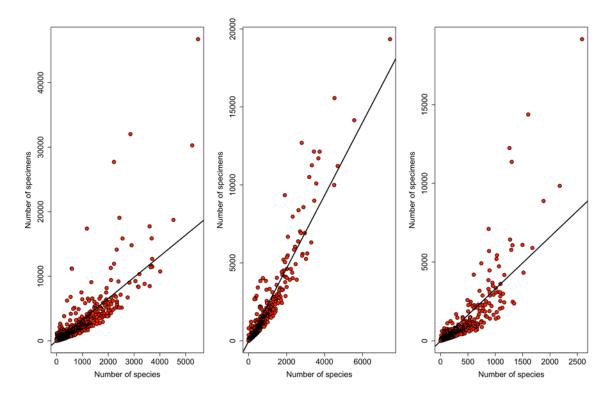
**Fig. S2** Relationships of the number of specimens collected per species with number of species collected in each flora for Australia, South Africa, and New England of the USA.

**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  $\lambda$ ) 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 <sup>NS</sup>	0.016 <sup>NS</sup>	0.15*	0.028 <sup>NS</sup>	-0.014 <sup>NS</sup>	0.0026 <sup>NS</sup>	-0.04 <sup>NS</sup>	0.098*
C <sub>mean</sub>									
Blomberg's	0.11 <sup>NS</sup>	0.57 <sup>NS</sup>	0.55 <sup>NS</sup>	0.42 <sup>NS</sup>	0.092 <sup>NS</sup>	0.56 <sup>NS</sup>	0.20*	0.093 <sup>NS</sup>	0.17 <sup>NS</sup>
К									
Pagel's	0.14**	0.00006 <sup>NS</sup>	0.053*	0.27 <sup>NS</sup>	0.23*	0.02 <sup>NS</sup>	0.00008 <sup>NS</sup>	0.00007 <sup>NS</sup>	0.19 <sup>NS</sup>
lambda									

References to Supporting Information

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