Widespread sampling biases in herbaria revealed from large-scale digitization

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Summary

• Nonrandom 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 c. 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).
• We identified numerous shared and unique biases among these regions. Shared biases included specimens collected close to roads and herbaria; specimens collected more frequently during biological spring and summer; specimens of threatened species collected less frequently; and specimens of close relatives collected in similar numbers. Regional differences included overrepresentation of graminoids in SA and AU and of annuals in AU; and peak collection during the 1910s in NE, 1980s in SA, and 1990s in AU. Finally, in all regions, a disproportionately large percentage of specimens were collected by very few individuals. We hypothesize that these mega-collectors, with their associated preferences and idiosyncrasies, shaped patterns of collection bias via ‘founder effects’.
• Studies using herbarium collections should account for sampling biases, and future collecting efforts should avoid compounding these biases to the extent possible.

Introduction

Herbaria contain a wealth of information about the ecological and evolutionary history of living and extinct species (Funk, 2003). Despite the continuous decline in plant collecting and declining support for herbaria (Dalton, 2003; Prather et al., 2004a,b), there has been a recent surge of studies leveraging herbarium collections for diverse research projects not focused on systematics (Pyke & Ehrlich, 2010; Lees et al., 2011; Feeley, 2012; Lavoie, 2013; Hart et al., 2014). These studies include studies on plant demography, current and future species distributions, and temporal changes in phenology and morphology (e.g. Miller-Rushing et al., 2006; Newbold, 2010; Pyke & Ehrlich, 2010; Lavoie, 2013; Staats et al., 2013; Davis et al., 2015; Willis et al., 2017a,b).

Ideally, herbarium collections used for these studies would include statistically unbiased samples of plant diversity across space and time. However, as the majority of specimens were collected for qualitative taxonomic and/or systematic inquiries, they were usually collected nonrandomly and sampling designs were rarely quantified (Wolf et al., 2011; Schmidt-Lebuhn et al., 2013). Because nonrandom samples may be statistically biased, analyzing them without accounting for biases might lead to spurious results (Syfert et al., 2013).

Sampling biases fall into several broad categories. Taxonomic or phylogenetic bias is the unbalanced sampling of certain taxa or clades over others, typically resulting 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 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 year (Cottrell 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 Global

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Biodiversity Information Facility (GBIF; Edwards et al., 2000). Their analyses revealed large taxonomic gaps in global plant occurrence data (< 25% of species of land plants were sampled); extensive spatial gaps across regions that harbor high concentrations of plant diversity, especially in Asia, Central Africa, and Amazonia; and strong temporal discontinuities in occurrence records across decades, all of which can hamper inferences about the effects on plants of recent and future environmental change.

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 collections have not been digitized, and of those that have been, many are unavailable, in whole or in part, on GBIF. Thus, Meyer et al.’s (2016) assessment of biases may itself be biased, or may inaccurately reflect biases in more complete, regional botanical collections that have been more fully mobilized. Furthermore, over two-thirds of the plant 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 collection data, focused on specific regions that have been predominantly digitized and validated, may reveal clearer patterns of sampling biases between regions than the global trends identified by Meyer et al. (2016) (cf. Hijmans et al., 2000 for Bolivian potatoes; Solanum).

Expanding upon Meyer et al.’s work, we explored spatial, temporal, and taxonomic/phylogenetic sampling biases in collections from three of the most extensively collected, digitized, and mobilized regional floras in the world: South Africa (SA), Australia (AU), and the New England (NE) region of the USA. The SA flora is a compilation of digitized herbarium specimens from all major herbaria across the country available in a single online portal (South African National Biodiversity Institute (SANBI), 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 AU. The Consortium of the Northeast Herbaria database contains digitized specimens from 15 participating herbaria in the NE region of the USA (Schorn et al., 2016). We also examined trait bias – sampling bias attributable to intrinsic life-history characteristics, including life cycle (annual vs perennial), plant height, growth form (woody vs 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 within each of these regional floras. Our results revealed both commonalities and differences in regional collection biases and identified new sampling foci as collections grow in the future.

Materials and Methods

Sources and description of data

We obtained 12 488 200 herbarium specimen records of vascular plants from AU (Australia Virtual Herbarium [AVH], 2016); 2049 905 herbarium specimen records from SA including Lesotho and Swaziland (South African National Biodiversity Institute (SANBI), 2016; and 879 388 herbarium specimen records from the NE (USA) flora (Consortium of Northeastern Herbaria (CNH), 2016). The records were cleaned in two steps (Supporting Information Fig. S1). First, we standardized the taxonomy of all species using the TAXONOMIC 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 Garden’s Tropicos (http://tropicos.org/) database or the PlantList (http://theplantlist.org/). Second, we removed specimens that were duplicates from the same collection locality and date; specimens with clearly erroneous locations (i.e. in oceans); specimens with zero coordinates and occurrences that fell outside the boundaries of our study; specimens missing exact collection date or georeferenced location data; and field observation records not tied to a physical specimen. Following this data cleaning, we retained 32% of the initial specimens for further analysis: 24% of the AU records (31 966 taxa; 2958 195 records); 49% (20 824 taxa; 1008 206 records) of the SA records; and 75% (3719 taxa; 661 370 records) of the NE records.

Analyses

Spatial biases First, we evaluated the density of sampling localities across the focal regions using Delaunay triangulation polygons, which measure the land area covered by each sampling locale (Fortune, 1992). Larger triangles indicate sparser collecting effort, whereas smaller triangles indicate more concentrated effort. Second, we examined infrastructure bias by calculating the minimum distance of each collection locality to the nearest major road (GADM, 2015) and herbarium (following Thiers, 2016). Our data set of roads derives from the publicly available Digital Chart of the World (http://maproom.psu.edu/dcw/), which was compiled by the US Defense Mapping Agency from 1 : 1000 000 scale paper maps (ESRI, 1992). All roads appearing at this scale were included in our analyses. Although this data set includes only larger roads and has not been updated since 1992, it probably represents the most comprehensive digital record of roads around the world. We then compared these distances to those generated by a null model (1000 iterations) in which the same number of sample points was randomly (Poisson) distributed across each geographic region. Third, we mapped geographic biases in sampling density, defined as areas of excessive (hotspots) or insufficient (coldspots) collection (Hijmans et al., 2000). Hotspots and coldspots were determined at a spatial grain of 0.25°×0.25° based on the number of specimens per grid cell, and identified using the 2.5% threshold (Orme et al., 2005; Ceballos & Ehrlich, 2006; Daru et al., 2015), based, respectively, on the 97.5th and 2.5th percentile values for the number of specimens collected per grid cell. Spatial distance calculations were computed with the functions DIST2LINE and SPDISTS in the R packages SP (Bivand et al., 2013) and GEOSPHERE (Hijmans, 2015), respectively. In our final predictive model of sampling density, we also included human population density (CIESIN, 2016), sampling localities, infrastructure (distance to herbaria and roads), number of specimens collected, and elevation.
Temporal bias For each regional flora, we explored bias at several temporal scales. Collection dates ranged from 20 May 1664 to 9 January 2016 (AU), 15 November 1656 to 6 June 2016 (SA), and 28 July 1687 to 4 May 2016 (NE). We hypothesized that collectors tended to avoid fieldwork in unfavorable conditions (e.g. winter and wartime) or on certain days of the week (e.g. weekdays for nonprofessional botanists). To test for temporal bias, we first recorded collection dates as days of the week (Sunday = 1, Monday = 2, etc.), and day of the year (DOY; where 1 January = 1 DOY and 31 December = 365 DOY, etc.). We then used a Rayleigh test of directional statistics in the R package circular (Agostinelli & Lund, 2013) to test whether each of these collection dates was randomly distributed against all dates spanning the entire duration of plant collection. If $P < \alpha = 0.05$, we rejected the null hypothesis of temporal uniformity at scales of weeks, days of the year, or decades.

Trait bias 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 regional databases (all accessed in June 2016), including: New South Wales Flora Online (http://plantnet.rbgsyd.nsw.gov.au); JSTOR Global Plants (https://plants.jstor.org); Atlas of Living Australia (http://b i.eala.org.au); Plants of Southwestern Australia (http://keys.lucidc ntrl.org); the African Plant Database (http://ville-ge.ch); Plants of Southern Africa (http://plantzafrica.com); Plant Resources of Tropical Africa (http://www.prota4u.org); Flora of North America (http://efloras.org); and the USDA Plants Database (http://plants.usda.gov). We then manually checked these data for inconsistencies in terminologies for defining certain traits. For example, ‘vines’ vs ‘lianas’ for climbers, ‘forbs’ vs ‘herbs’ for herbaceous life forms, and ‘biennial’ for perennial growth duration. Extinction risk assessments for each species were obtained from the International Union for the Conservation of Nature (IUCN) Red List database (www.iucnredlist.org, accessed August 2016), which uses the following categories: Data Deficient (DD), Least Concern (LC), Lower Risk/Conservation Dependent (LR/CD), Near Threatened (NT), Vulnerable (VU), Endangered (EN), Critically Endangered (CR), and Extinct (EX). We grouped these narrow categories into two broader threat categories, threatened (EX + CR + EN + VU) and 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 observed specimens collected per species with the abundance of a species if specimen collection was equal across all species for each trait category. Because of dramatically unequal sampling effort in some species – for example Senna artemisioides with 10 167 specimens vs Eucalyptus cordonata with only one – and the low coverage of taxa with available trait data, we randomly sampled 50 specimens from each available species with trait data using 1000 randomizations. Species with < 50 specimens were excluded from this analysis.

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

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 3700 from NE.

We estimated phylogenetic signal using three common metrics: Abouheif’s $C_{\text{mean}}$ statistic (Abouheif, 1999), Blomberg’s $K$ (Blomberg et al., 2003), and Pagel’s lambda ($\lambda$) (Pagel, 1999). Significance was assessed by comparing observed values to a null distribution created by shuffling the trait values across the tips of the phylogeny 1000 times. Pagel’s $\lambda$ uses a maximum-likelihood method with branch-length transformation to estimate the 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 phylogenetic signal is indicated by a $K$ value > 1) and Pagel’s $\lambda$ were calculated using the R package phytools (Revell, 2012). Abouheif’s $C_{\text{mean}}$ was calculated using adephylo (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 NE: Asteraceae, Brassicaceae, Cyperaceae, Ericaceae, Fabaceae, Lamiaceae, Poaceae, Ranunculaceae, and Rosaceae.

In addition to phylogenetic signal, we also used phylogenetic generalized least squares (PGLS) regressions 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 endangerment’ (EDGE; Isaac et al., 2007). Species ages were measured as the length of terminal branches (BL) linking species on a phylogenetic tree. ED measures the degree of 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 endangerment (GE) from IUCN conservation categories: $GE = \ln(1 + ED) + GE \times \ln(2)$, where GE represents expected probability of species extinction over a 100-yr period (Redding & Mooers, 2006) categorized as follows: Least Concern $= 0.001$, Near Threatened and
Conservation Dependent = 0.01, Vulnerable = 0.1, Endangered = 0.67, and Critically Endangered = 0.999.

Last, we examined the phylogenetic structure of collecting efforts across decades 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 clustering.

**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 number of specimens collected per collector and the number of species collected per collector.

**Computation and availability of data and code**

All statistical analyses were conducted using the Research Computing Clusters of Harvard University (https://rc.fas.harvard.edu/). Data files and custom R scripts are available from the Harvard Forest Data Archive, dataset HF296 (http://harvardforest.fas.harvard.edu/data-archive).

**Results**

**Spatial bias**

High sampling density was observed in southeastern and southwestern AU, the Cape region of 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 specimens, we found a mismatch between hotspots (top 2.5% quantiles) and coldspots (lowest 2.5% quantiles) of herbarium sampling within quarter degree grids for (d) Australia, (e) South Africa and (f) New England. The hotspots and coldspots are the top and lowest 2.5% quantiles, respectively, of the number of specimens per locale.

![Fig. 1 Spatial bias in herbarium collections. (a–c) The geographic distribution of herbarium collecting activity depicting the spatial variation in sampling effort using Delaunay polygon tiles for (a) Australia (n = 857 245 locales), (b) South Africa (n = 61 130 locales), and (c) New England (n = 130 374 locales). (d–e) Hotspots (red) and coldspots (blue) of herbarium sampling within quarter degree grids for (d) Australia, (e) South Africa and (f) New England. The hotspots and coldspots are the top and lowest 2.5% quantiles, respectively, of the number of specimens per locale.](image-url)
quantiles) of sampling intensity (Fig. 1d–f), suggesting that hotspots and coldspots were not randomly distributed. Hotspots of collecting tended to cluster around coasts in AU and SA, whereas coldspots were abundant in interior areas. In NE, hotspots were concentrated in the south and coldspots occurred in the north.

Herbarium specimens tended to be collected closer than expected to roads and herbaria ($P<0.01$; Fig. 2a,b). More than 50% of herbarium specimens were collected within 2 km of road-sides in all three floras ($P<0.01$; Fig. 2a). Moreover, distance to herbaria explained 45% of the variance in collecting effort in AU, 29% in SA and 12.3% in NE, with a higher density of specimens closer to herbaria (Table 1). Despite substantial gradients in altitudes in each region (−15 to 2022 m above sea level (asl) in AU; 1 to 3254 m asl in SA; and −3 to 1485 m asl in NE), most specimens were collected below 500 m asl in AU and NE (81, 44, and 93% of specimens in AU, SA, and NE, respectively; Fig. 2c). We also found a negative correlation between collecting effort and altitude in AU and SA, suggesting a tendency for specimens to be collected at lower elevations. However, the opposite was true for NE, where more specimens tended to be collected at higher elevations than expected by chance (Table 1).

Temporal bias

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). 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 occurred c. 100 yr after peak collection activity in NE. Seasonally, specimen collections were biased toward spring and summer for all three floras, with peak collections ranging from September to December in AU and SA (Rayleigh $Z=0.189$ and $Z=0.251$, respectively; both $P<0.001$), and May to September in NE (Rayleigh $Z=0.718$; $P<0.001$; Fig. 4a). There was a significant trend toward collection on weekends (Saturdays and Sundays) in NE (Rayleigh test $Z=1.0$; $P<0.001$) and midweek in SA and AU (Rayleigh test $Z=0.105$ and $Z=1.0$, respectively; both $P<0.001$; Fig. 4a).

Trait bias

Perennials were more frequently collected than annuals in terms of specimens per species in SA and NE; the opposite was true for AU where there was a greater tendency for annuals to be collected (Fig. 5a). Similarly, graminoid specimens per species were overrepresented relative to other habits in AU and SA, whereas herbs and trees were overrepresented in NE (Fig. 5b). Relatively short plants were more frequently represented than taller plants in all three floras: 79.3, 89.3 and 84.9% of the plants collected in AU, SA and NE, respectively, were < 5 m in height (Fig. 5c).

Threatened species were collected significantly less often than nontreated plants across all three floras (all $P<0.001$; Fig. 5d).

Phylogenetic bias

It is possible that closely related species were collected more similarly (either undercollected or overcollected) than expected by chance. We tested this assumption phylogenetically and found a significant, but weak phylogenetic signal in the abundance of specimens per species across all three floras (Table 2). Specifically, closely related species tended to have a more similar number of specimens than expected (Table 2; Fig. 6). This phylogenetic bias was strongest in SA (Abuohieif’s $C_{\text{mean}}=0.15$ and $\lambda=0.32$; both $P<0.01$, but $K=0.0013$ (ns)). For instance, in SA, collections from the genus Protea averaged 115 specimens per species whereas only two specimens were collected for species in the genus Rytigynia on average. Most Agoseris in NE were represented by < 10 specimens per species, whereas many fern genera were represented by high specimen numbers (e.g. Onoclea with 845 specimens/species). Australian collections showed the weakest phylogenetic bias (Abuohieif’s $C_{\text{mean}}=0.12$ 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 higher collection numbers than 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% (SA). In general, EDGE species (distinctive, endangered species) were generally undercollected 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).

Collector bias

The number of specimens per collector was highly skewed (Fig. 8). In AU, > 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, > 50% of the specimens were amassed by 9.5% of collectors, including J. P. H. Acoks (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 AU, 0.95 in SA and 0.84 in NE; all $P<0.01$; Fig. S2).
Fig. 2  Comparison of geographic sampling bias of herbarium records in relation to (a) the minimum distance to roads; (b) the minimum distance to herbaria; and (c) regional altitudes at sampling locales. Black lines in (a) and (b) correspond to sampling locales and red indicates an equal number of random points generated 1000 times. Dark gray shading in (c) corresponds to sampling locales and red shading corresponds to regional altitudes, that is, all other altitudes, for all three floras: Australia (left), South Africa (middle) and New England (right). The dotted line in (c) indicates an altitude of 500 m above sea level.
Table 1 Model coefficients for multiple regressions of collecting effort in terms of the number of specimens collected per locality

<table>
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<th>Predictors (log_{10}-transformed)</th>
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<th>P-values</th>
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Discussion

Historically, the primary function of herbaria has been to serve as a resource for botanists carrying out taxonomic and systematic research, allowing users to construct classifications of plants, verify identifications, determine the ranges and morphological 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 ways that collectors rarely imagined (Pyke & Ehrlich, 2010; Lavoie, 2013; Nualart et al., 2017; Rudin et al., 2017). Among these, the most comprehensive investigation is by Meyer et al. (2016), who proposed an important conceptual framework for analyzing gaps and biases along taxonomic, geographic, and temporal dimensions. Although Meyer et al. (2016) focused more on observational records than herbarium collections, they uncovered numerous biases in ‘digitally accessible information’ regarding plants and provided an important baseline for evaluating and improving global floristic coverage in collection data. However, collection biases in large geographic areas are difficult to categorize because the collections’ data are not yet digitized, and this may skew the global patterns of bias noted by Meyer et al. (2016). By focusing on three of the most well-collected and 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 regional collections.

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 probably increased as botanists tended to travel and collect using automobiles (Everill et al., 2014). Because roads are known to alter local environmental conditions and facilitate biological invasions (e.g. Forman & Alexander, 1998; Hui et al., 2003; Griffith 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 distributions. Specifically, species collected along roadsides are likely to be overrepresented by species that thrive with disturbance, and underrepresented by forest interior and wetland species that are harmed by disturbance (Gutzwiller & Flather, 2011; Rivers-Moore & Cowden, 2012). As the road network continues to expand and as people become ever more concentrated in cities, this bias toward collecting near roads might become stronger in coming decades.

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

Although we realize that patterns of species richness may not be randomly distributed across the landscape, accounting for underlying patterns of richness or abundance is difficult because our knowledge of such patterns often derives from (and is thus not independent of) these same (biased) collections. By comparing locations of samples (collections) against a Poisson set of points and specimens per species, and not total collection numbers, we tested only for the nonrandom distribution of collection...
Fig. 3 Timeline of herbarium specimen collection density in relation to major historical events in time (indicated in red text) for the three floras: Australia, South Africa and New England. Analysis of phylogenetic structure through time by binning sequences of collection dates into decades and testing for overdispersion vs clustering is indicated in black font. The red trend line indicates the gross domestic product of each region. NRI, net relatedness index; NTI, nearest taxon index. **, P < 0.001; *, P < 0.01.
locations on a landscape. And, indeed, we found that the collection locations were not spatially distributed randomly (Poisson) on the landscape. It is also possible that georeferencing might have introduced additional bias in some specimens. While ascertaining the degree of accuracy of georeferenced records might be challenging because such information is often unavailable, our cumulative curves are probably less affected.

Temporal and seasonal bias

Collections in AU and SA increased through time until a few decades ago, but those in NE peaked much earlier in the early 1900s. These differences between regional collection activities may parallel broader societal factors influencing plant collection. In 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 botanical expeditions in the region coinciding with the ‘Golden Age’ of plant collecting in Europe and North America (Whittle, 1970; Musgrave et al., 1999). In SA, collection efforts began much later, peaking during the Apartheid Era (1948–1994), and declined thereafter under the New Democratic Rule, concomitant with the general economic decline of the country and concern for public safety (Ferreira & Harmse, 2000; Lemanski, 2004). In AU, the mass immigration of Europeans in 1948 after World War II included numerous highly skilled professionals (Price, 1998; Leuner, 2007) and coincided with an enormous increase in botanical collecting. Botanical collecting may have declined more recently as a result of legislation in AU and SA to regulate collection activities, especially those designed to protect rare and endangered species.
Collecting efforts within a season revealed common patterns of bias: specimens in all three regions were collected overwhelmingly in biological spring and summer. Sampling during these time periods probably reflects efforts to collect plants in good flowering and fruiting condition. However, this seasonal bias probably overlooks key developmental transitions (e.g. Poethig, 2017).
Abouheif’s Blomberg’s K and Pagel’s λ  

<table>
<thead>
<tr>
<th></th>
<th>Australia (n = 5814 species)</th>
<th>South Africa (n = 3568 species)</th>
<th>New England (n = 3700 species)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Abouheif’s C_{\text{mean}}</td>
<td>0.12**</td>
<td>0.15**</td>
<td>0.12**</td>
</tr>
<tr>
<td>Blomberg’s K</td>
<td>0.000085ns</td>
<td>0.0013ns</td>
<td>0.0030ns</td>
</tr>
<tr>
<td>Pagel’s λ</td>
<td>0.18**</td>
<td>0.32**</td>
<td>0.29**</td>
</tr>
</tbody>
</table>

Phylogenetic data are derived from Zanne et al. (2014). All tests are based on 1000 randomizations. **, $P < 0.001$; ns, $P > 0.05$.

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

Annuals were overrepresented relative to perennials in herbarium collections in AU; the opposite was observed in SA and NE. There was also a high representation of 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 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 nonnative perennial grasses interspersed with forbs and woody plants (Bond & Parr, 2010). New England, in contrast, is generally forested and has an abundance of shade-tolerant shrubs and perennial herbs. Graminoids are also considered harder to identify and may be avoided by nonspecialists. Lianas and vines simultaneously represent the smallest proportion of growth forms and comprise the least number of species per species in all three floras. 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...
Fig. 7 Phylogenetic bias in collection frequency for exemplar families in New England flora. Phylogenetic bias is indicated by significant phylogenetic signal in at least one of three metrics (Abouheif’s $C_{\text{mean}}$, Blomberg’s $K$ and Pagel’s $\lambda$). The color bar illustrates values within families: log$_{10}$ numbers of specimens per species and ranges from red (low number of specimens per species) to blue (high number of specimens per species). **, $P < 0.001$; *, $P < 0.01$; ns, $P > 0.05$. 

Asteraceae
$C_{\text{mean}} = 0.19^{**}$
$\lambda = 0.22^{**}$
$K = 0.02^{ns}$

Brassicaceae
$C_{\text{mean}} = 0.02^{ns}$
$\lambda = 0.00007^{ns}$
$K = 0.037^{ns}$

Cyperaceae
$C_{\text{mean}} = 0.07^{*}$
$\lambda = 0.035^{ns}$
$K = 0.0077^{ns}$

Ericaceae
$C_{\text{mean}} = 0.06^{**}$
$\lambda = 0.16^{ns}$
$K = 0.051^{*}$

Fabaceae
$C_{\text{mean}} = 0.02^{ns}$
$\lambda = 0.09^{ns}$
$K = 0.013^{ns}$

Lamiaceae
$C_{\text{mean}} = 0.07^{ns}$
$\lambda = 0.21^{ns}$
$K = 0.093^{ns}$

Poaceae
$C_{\text{mean}} = 0.06^{*}$
$\lambda = 0.046^{*}$
$K = 0.03^{ns}$

Ranunculaceae
$C_{\text{mean}} = 0.009^{ns}$
$\lambda = 0.00007^{ns}$
$K = 0.03^{ns}$

Rosaceae
$C_{\text{mean}} = 0.07^{*}$
$\lambda = 0.00007^{ns}$
$K = 0.009^{ns}$
among the three floras in the traits of botanical collections almost certainly are reflective of genuine differences in species abundance and diversity.

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., 2007; Meyer et al., 2016) did not use the full complement of modern phylogenetic methods that included patterns of evolutionary relatedness, and so were limited in their

Table 3 Multiple regressions of phylogenetic generalized least squares of collecting effort (frequency) of herbarium specimens with phylogenetic metrics of species uniqueness

<table>
<thead>
<tr>
<th></th>
<th>Predictors (log10-transformed)</th>
<th>Percentage of variance explained (%)</th>
<th>P-values</th>
<th>Model adjusted $R^2$</th>
<th>Model slope</th>
<th>Model intercept</th>
</tr>
</thead>
<tbody>
<tr>
<td>Australia</td>
<td>BL</td>
<td>1.36</td>
<td>0.7</td>
<td>0.049</td>
<td>0.035</td>
<td>4.37</td>
</tr>
<tr>
<td></td>
<td>ED</td>
<td>0.2</td>
<td>0.008</td>
<td>&lt; 0.001</td>
<td>0.44</td>
<td></td>
</tr>
<tr>
<td></td>
<td>EDGE</td>
<td>3.75</td>
<td>&lt; 0.001</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>South Africa</td>
<td>BL</td>
<td>0.47</td>
<td>0.3</td>
<td>0.09</td>
<td>-0.63</td>
<td>3.63</td>
</tr>
<tr>
<td></td>
<td>ED</td>
<td>0.000015</td>
<td>0.001</td>
<td></td>
<td>0.63</td>
<td></td>
</tr>
<tr>
<td></td>
<td>EDGE</td>
<td>8.89</td>
<td>&lt; 0.001</td>
<td></td>
<td>-1.3</td>
<td></td>
</tr>
<tr>
<td>New England</td>
<td>BL</td>
<td>0.09</td>
<td>0.94</td>
<td>1.70E-02</td>
<td>-0.0052</td>
<td>3.89</td>
</tr>
<tr>
<td></td>
<td>ED</td>
<td>0.054</td>
<td>0.0045</td>
<td></td>
<td>0.79</td>
<td></td>
</tr>
<tr>
<td></td>
<td>EDGE</td>
<td>1.87</td>
<td>&lt; 0.001</td>
<td></td>
<td>-2.28</td>
<td></td>
</tr>
</tbody>
</table>

BL, terminal branch length; ED, evolutionary distinctiveness; EDGE, evolutionary distinctiveness and global endangerment.

Fig. 8 Collector bias in herbarium collections. The number of herbarium specimens amassed per collector for three regional floras in (a) Australia, (b) South Africa, and (c) New England is shown. The top five collectors in each flora are highlighted in red. Numbers within parentheses correspond to lifespans of the collectors, with collectors that have died highlighted in red and currently living ones in black.
ability to detect details of taxonomic bias. By contrast, our phylogenetic approach not only captured taxonomic bias in favor of certain entire families (e.g., Asteraceae, Cyperaceae, Poaceae, and Rosaceae in NE), but revealed that evolutionarily distinct and globally endangered species are underrepresented in herbarium records relative to more common species. Such evolutionarily distinct species, which are threatened with extinction, represent important targets for future documentation or prioritization for conservation (Isaac et al., 2007). However, collecting threatened taxa requires specialized training, compliance with regulations, and awareness of actual collection needs (Minteer et al., 2014). Increasingly, DNA barcoding approaches, using small samples from living tissues, combined with GPS-referenced digital photography might be an avenue to document such species.

Collector bias

In all three regions, a large percentage of specimens was gathered by only a few collectors (Fig. 8). Thus, the habits and preferences of a few individuals probably shaped the establishment and formation of these herbarium collections. These ‘founder effects’ propagate across all the dimensions of collection bias examined above, and help us to understand past collection behavior (Bepper et al., 2012). For example, certain collectors may focus on geographically circumscribed floristic zones, often near their place of residence, 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 botanists may tend to collect specimens on weekdays at any time of the year, whereas amateurs and faculty with teaching responsibilities may focus their efforts on weekends and vacation months. Those interested in function and physiology may only collect plants of certain habits or life histories (e.g., carnivorous, aquatic or succulent plants). These effects would probably be compounded when associated with mega-collectors. For instance, the Harvard University Herbaria’s collection of Asian, especially woody plants was largely built by a few collectors and dates to the early establishment of 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 these early pioneers is a ripe area for future research.

Future collecting

To ensure that herbaria continue to be vital centers for research beyond their importance to taxonomy and systematics, herbarium directors and collectors should account for and, whenever possible, reduce biases in plant collections. Biases can be accounted for to a 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 collections from herbaria, roads, or other infrastructure (McCarthy et al., 2012), using rarefaction methods to predict abundances (Schmidt-Lebuhn et al., 2013), or including 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 expeditions should focus on ‘coldspots’ of collection intensity (Hijmans et al., 2000), that is, places that are underrepresented in collections. Although some of the coldspots we identified probably represent more inaccessible environments, they often correspond to 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 indicate areas where herbarium specimens have yet to be mobilized, providing additional focus for efforts to make collection data widely available. Equally important is the need 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 nonnative invasive species that have rapidly expanding distributions and vulnerable native species that have ranges that are collapsing.

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

We acknowledge that some of the biases also may be attributed to longstanding curation practices at the herbariums themselves. As herbarium collections were amassed for qualitative floristic, taxonomic, and systematic research, duplicate specimens of common species and nonreproductive material have sometimes been discarded, sent elsewhere, or not accepted in the first place. This trend is becoming even more 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 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 through continued collecting of specimens to overcome past biases. And, most importantly, researchers analyzing herbarium specimens in a widening array of studies needed to be aware of the biases in these collections, and apply appropriate statistical techniques.

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Author contributions

References


Supporting Information

Additional Supporting Information may be found online in the Supporting Information tab for this article:

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

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

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

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