

1  
2  
3  
4  
5  
6  
7  
8  
9  
10  
11  
12  
13  
14  
15  
16  
17  
18  
19  
20  
21  
22  
23  
24  
25

MR. BRIAN SALVIN MAITNER (Orcid ID : 0000-0002-2118-9880)

DR. NAIA MORUETA-HOLME (Orcid ID : 0000-0002-0776-4092)

DR. STEPHEN SMITH (Orcid ID : 0000-0003-2035-9531)

DR. CYRILLE VIOLLE (Orcid ID : 0000-0002-2471-9226)

Article type : Application

Handling Editor: Dr. Sean McMahon

Corresponding author email id: [bmaitner@gmail.com](mailto:bmaitner@gmail.com)

**Title:**

**The BIEN R package: A tool to access the Botanical Information and Ecology Network (BIEN) Database**

**Authors:**

Brian S. Maitner<sup>1</sup>, Brad Boyle<sup>1</sup>, Nathan Casler<sup>2</sup>, Rick Condit<sup>3</sup>, John Donoghue II<sup>1</sup>, Sandra M. Durán<sup>1</sup>, Daniel Guaderrama<sup>1</sup>, Cody E Hinchliff<sup>4</sup>, Peter M. Jørgensen<sup>5</sup>, Nathan J.B. Kraft<sup>6</sup>, Brian McGill<sup>7</sup>, Cory Merow<sup>8</sup>, Naia Morueta-Holme<sup>9</sup>, Robert K. Peet<sup>10</sup>, Brody Sandel<sup>11</sup>, Mark Schildhauer<sup>12</sup>, Stephen A. Smith<sup>4</sup>, Jens-Christian Svenning<sup>13</sup>, Barbara Thiers<sup>14</sup>, Cyrille Violle<sup>15</sup>, Susan Wisser<sup>16</sup> and Brian J. Enquist<sup>1,17</sup>

**Author Affiliations:**

This is the author manuscript accepted for publication and has undergone full peer review but has not been through the copyediting, typesetting, pagination and proofreading process, which may lead to differences between this version and the [Version of Record](#). Please cite this article as [doi: 10.1111/2041-210X.12861](https://doi.org/10.1111/2041-210X.12861)

This article is protected by copyright. All rights reserved

26 <sup>1</sup>Department of Ecology and Evolutionary Biology, University of Arizona,  
27 Tucson, AZ 85721, USA;  
28 <sup>2</sup>National Center for Supercomputing Applications, University of Illinois Urbana-Champaign,  
29 Urbana, IL 61801, USA;  
30 <sup>3</sup>Smithsonian Tropical Research Institute, Center for Tropical Forest Science, Global Forest  
31 Observatory Network, Unit 9100 Box 0948, DPO AA 34002, Panama;  
32 <sup>4</sup>Department of Ecology and Evolutionary Biology, University of Michigan, Ann Arbor, MI  
33 48109, USA;  
34 <sup>5</sup>Missouri Botanical Garden, P.O. Box 299, St. Louis, MO 63166-0299, USA;  
35 <sup>6</sup>Department of Ecology and Evolutionary Biology, University of Los Angeles, Los Angeles, CA  
36 90095, USA;  
37 <sup>7</sup>School of Biology and Ecology, University of Maine, Orono, ME 04469, USA;  
38 <sup>8</sup>Department of Ecology and Evolutionary Biology, Yale University, New Haven, CT 06520,  
39 USA;  
40 <sup>9</sup>Department of Integrative Biology, University of California, Berkeley, CA 94720, USA;  
41 <sup>10</sup>Department of Biology, University of North Carolina, Chapel Hill, NC 27599-3280, USA;  
42 <sup>11</sup>Department of Biology, Santa Clara University. Santa Clara, CA 95057, USA;  
43 <sup>12</sup>National Center for Ecological Analysis and Synthesis, Santa Barbara, CA, USA;  
44 <sup>13</sup>Section for Ecoinformatics & Biodiversity, Department of Bioscience, Aarhus University, Ny  
45 Munkegade 114, DK-8000 Aarhus C, Denmark;  
46 <sup>14</sup>William and Lynda Steere Herbarium at the New York Botanical Garden, Bronx, NY, USA;  
47 <sup>15</sup>Centre d'Ecologie Fonctionnelle et Evolutive (UMR 5175), CNRS – Université de Montpellier  
48 – Université Paul Valéry Montpellier, EPHE, Montpellier, France;  
49 <sup>16</sup>Landcare Research, Lincoln 7608, New Zealand;  
50 <sup>17</sup>The Santa Fe Institute, USA, 1399 Hyde Park Rd, Santa Fe, NM 87501, USA;

51

52 **Tweetable Abstract:**

53 BIEN for R: Global plant occurrences, trait data, plot data. Range maps and phylogenies for  
54 most New World plants.

55

56 **Abstract:**

57 **1.** There is an urgent need for large-scale botanical data to improve our understanding of  
58 community assembly, coexistence, biogeography, evolution, and many other fundamental  
59 biological processes. Understanding these processes is critical for predicting and handling  
60 human-biodiversity interactions and global change dynamics such as food and energy security,  
61 ecosystem services, climate change, and species invasions.  
62

63 **2.** The Botanical Information and Ecology Network (BIEN) database comprises an  
64 unprecedented wealth of cleaned and standardized botanical data, containing roughly 81 million  
65 occurrence records from ~375,000 species, ~915,000 trait observations across 28 traits from  
66 ~93,000 species, and co-occurrence records from 110,000 ecological plots globally, as well as  
67 100,000 range maps and 100 replicated phylogenies (each containing 81,274 species) for New  
68 World species. Here, we describe an R package that provides easy access to these data.  
69

70 **3.** The BIEN R package allows users to access the multiple types of data in the BIEN database.  
71 Functions in this package query the BIEN database by turning user inputs into optimized  
72 PostgreSQL functions. Function names follow a convention designed to make it easy to  
73 understand what each function does. We have also developed a protocol for providing  
74 customized citations and herbarium acknowledgements for data downloaded through the BIEN R  
75 package.  
76

77 **4.** The development of the BIEN database represents a significant achievement in biological  
78 data integration, cleaning and standardization. Likewise, the BIEN R package represents an  
79 important tool for open science that makes the BIEN database freely and easily accessible to  
80 everyone.

81  
82 **Keywords:** R, plants, range maps, presence, occurrence, phylogeny, community plot, functional  
83 traits, biodiversity, herbarium records, ecoinformatics  
84

## 85 **Introduction:**

86 Understanding the distribution and diversity of organisms relies on the availability of large  
87 quantities of standardized, robust, and comprehensive biological data. Furthermore, developing

88 ecological theories that are both general and predictive also necessitates high-quality data. There  
89 is an especially urgent need for large quantities of botanical data across differing spatial scales.

90 Despite the crucial importance of plants for ecosystem functioning and services, plants have  
91 lacked comprehensive data sets similar to those of less-speciose clades such as birds and  
92 mammals, which have trait (Myhrvold *et al.* 2015), phylogenetic (Bininda-Emonds *et al.* 2007;  
93 Jetz *et al.* 2012; Faurby & Svenning 2015) and range information (BirdLife International 2013;  
94 IUCN 2013) available for most species. Critical issues with economic implications, such as food  
95 and energy security, climate change, species invasions and ecosystem services, require an  
96 understanding of plant ecology and evolution globally, and an ability to predict how these factors  
97 will respond to global change.

98 In the era of Big Data and open science, the informatics challenge is to develop  
99 workflows that can standardize and provide access to disparate and heterogeneous data sources.  
100 Increasingly, the synthetic questions biologists are asking require computationally intensive  
101 integration between these numerous and diverse data sources. However, the existence and  
102 accessibility of data is only part of the issue. Before data can be integrated, they must be  
103 assessed, cleaned, and standardized.

104 The Botanical Information and Ecology Network (BIEN; <http://bien.nceas.ucsb.edu/bien/>  
105 ) database is an unprecedented collection of integrated and standardized New World botanical  
106 data (Enquist *et al.* 2016). This has resulted from the efforts of more than 50 scientists from 20  
107 institutions across the globe. The BIEN database is constructed from a workflow that performs  
108 three essential tasks. First, it integrates multiple data types (e.g. trait data, occurrence records,  
109 ecological plot data, Table 1) from numerous sources (Supplementary Tables S1-3). Second,  
110 utilizing open source tools, it assesses and corrects taxonomic errors, standardizes taxonomy,  
111 validates geographic information, and standardizes data formats. Third, it generates species-level  
112 distribution models and phylogenies. The current version of the BIEN database (BIEN 3.4 as of  
113 Jan 15th 2017) contains 81,108,996 occurrence records that stem from 378,502 species from  
114 across the globe. While the data are global in extent (Figures 1 & 2), data integration efforts have  
115 focused on the New World (North and South America, Table 2).

116 The BIEN database differs from other biodiversity databases in the breadth of data types  
117 it contains (Table 1), the integration of these data, the degree of validation and standardization  
118 the data goes through, and the amount of additional metadata added. Most other databases focus

119 on a single data type, such as TRY (Kattge *et al.* 2011) with trait data and GBIF ('Global  
120 Biodiversity Information Facility' 2012) with occurrence records, where BIEN integrates  
121 multiple data types, including occurrence, plot, trait, and phylogenetic data. Data within the  
122 BIEN database are also put through standardization and validation procedures, thereby  
123 increasing data quality and ease of use. For example, BIEN occurrence data undergo additional  
124 taxonomic and spatial validations that aren't used by GBIF ('Global Biodiversity Information  
125 Facility' 2012), and unlike Traitbank (Schulz 2016), BIEN trait units and names have been  
126 standardized.

127 The wealth of information contained in the BIEN database is housed at the National  
128 Center for Ecological Analysis and Synthesis (NCEAS) in Santa Barbara, CA. The database is  
129 open-access and can either be accessed by a graphical user interface (GUI) at BIENdata.org or  
130 through the newly developed BIEN package for R that we introduce here (R Core Development  
131 Team 2015). The BIENdata.org website offers a user-friendly way to visualize species  
132 occurrence data and range maps and download range maps, occurrences and trait data for a  
133 single species at a time. Users who are interested in downloading data for many taxa or  
134 geographic locations and accessing differing combinations of data will find the BIEN R package  
135 a more flexible approach.

136

### 137 **Data Cleaning and Integration Workflow**

138 As described by Enquist *et al.* (in prep., 2016), the generation of the BIEN database  
139 consists of a linked workflow that (i) standardizes taxonomy by correcting spelling of species  
140 names and updating synonyms to currently accepted names via the Taxonomic Name Resolution  
141 Service or TNRS (Boyle *et al.* 2013); (ii) detects and flags observations with erroneous  
142 geographic coordinates; and (iii) flags cultivars and non-native species via the Native Species  
143 Resolver (<http://bien.nceas.ucsb.edu/bien/tools/nsr/>). Coordinates were flagged as erroneous if  
144 they fell outside the specified political regions, if the latitude was exactly 0 or 90 degrees, the  
145 longitude was exactly 0 or 180 degrees, or if the point fell in the ocean. The detection of  
146 cultivars and non-native records relies on native species lists, which are not available throughout  
147 the New World, so this filtration is imperfect.

148 Range maps were built for each species using a method determined by the number of  
149 observations of that species. A species with a single record was assigned a range that included

150 only the 100km<sup>2</sup> cell where it was found. Ranges for species with 2-3 records were rectangular  
151 bounding boxes with the limits set by the minimum and maximum latitude and longitude of all  
152 occurrences. Ranges for species with 4-9 records were built with convex hulls (the minimum-  
153 fitting polygon that encompasses all occurrences of that species). For species with >9 records,  
154 we built species distribution models using the Maxent algorithm (Phillips *et al.* 2006). Only one  
155 occurrence record per cell (in cases of multiple records) was used for Maxent model building.  
156 Maxent model building generally followed the recommendations outlined in (Merow *et al.*  
157 2013, 2014). Model settings were chosen to balance overfitting, which underestimates range  
158 sizes, with underfitting, which results in excessively smooth models that over predict range size.  
159 Only linear, quadratic, and product features were used and regularization was set at the default  
160 value.

161 Environmental predictors for SDMs were obtained from the WorldClim current (1960-  
162 1990) climate data at 10 arc-minute resolution (Hijmans *et al.* 2005) and resampled to a 10 km  
163 resolution. Predictors included mean annual temperature, mean diurnal temperature range,  
164 annual precipitation, precipitation seasonality, precipitation in warmest quarter/ (precipitation in  
165 warmest quarter + precipitation in coldest quarter), and five spatial eigenvectors (De Marco *et al.*  
166 2008). The spatial eigenvectors essentially captured large-scale regional differences in  
167 occurrence and primarily served as broad-scale dispersal limitation of species ranges, limiting  
168 predictions far in geographic space from presence locations (Blach-Overgaard *et al.* 2010).

169 Maxent's continuous predictions were converted to binary presence/absence predictions  
170 by choosing a threshold based on the 75th percentile of the cumulative output.

171 We constructed a phylogeny of 18,641 species based on a standardized list of New World  
172 species and the gene regions *atpB-rbcL*, *ndhF*, *psbA*, *psbA-psbH*, *rbcL*, and *trnT-trnL-trnF*  
173 marker regions using the software PHLAWD (Smith *et al.* 2009). The phylogeny was  
174 constructed with RAxML (7.3.0; Stamatakis 2006) with an unconstrained ML search and  
175 divergence times were estimated using penalized likelihood and the treePL software package  
176 (Smith & O'Meara 2012). Additional details on the methodology used to extract these data from  
177 GenBank and align them are presented in Hinchliff and Smith (2014). Onto this phylogeny, we  
178 grafted the additional taxa from the BIEN dataset using taxonomy (genus membership) as a  
179 guide for the remaining ~72,000 species. This grafting was repeated to create a set of 100  
180 phylogenies to account for uncertainty in placement of species without genetic information.

181 Additional information on the BIEN phylogenies is available online  
182 (<http://bien.nceas.ucsb.edu/bien/biendata/bien-2/phylogeny/> ).

183

#### 184 **BIEN R package**

185 A common hurdle for researchers is accessing the specific subset of data they need for  
186 their study. Particularly for large databases with multiple data types, data access can involve  
187 complex queries in database languages that are unfamiliar to many researchers. The BIEN  
188 package for R (stable version available on CRAN and development version at  
189 <https://github.com/bmaitner/RBIEN>) provides a set of tools to access and work with the BIEN  
190 database that allows users without a knowledge of PostgreSQL or the BIEN database structure to  
191 easily access the data within BIEN.

192

193 The functions within the BIEN package fall into nine main categories:

194

- 195 I. Species lists,
- 196 II. Occurrence records (species observations from specimens and plots),
- 197 III. Plot data (species occurrence and abundance from plots),
- 198 IV. Stem data (stem counts and measurements from plots),
- 199 V. Trait data,
- 200 VI. Species range maps,
- 201 VII. Taxonomic information,
- 202 VIII. Phylogenetic information,
- 203 IX. Associated metadata

204

205 We follow a function naming convention where each function begins with the prefix “BIEN\_”  
206 followed by a designation of which data type it accesses (eg “`trait_`” or “`ranges_`”) followed  
207 by a description of what the specific function does. For example, the function  
208 `BIEN_occurrence_species()` is used to download geographic information system (GIS)  
209 occurrence data for a given species. Detailed descriptions of each function family, and examples  
210 of common usages can be found in Supplementary Document S4. A complete list of functions  
211 with abbreviated descriptions is available in tabular format with `vignette("BIEN")`, and a

212 tutorial is available with `vignette("BIEN_tutorial")`. Data usage caveats can be found in  
213 Supplementary Table S5.

214

### 215 **Data license**

216 The data and derived products are distributed via CC-BY-NC-ND Creative Commons  
217 license (<https://creativecommons.org/licenses/by-nc-nd/4.0/>). This allows redistribution and re-  
218 use of a licensed work on the conditions that the creator is appropriately credited, that use is  
219 noncommercial, and that users do not distribute derivative works. Users interested in uses of  
220 these data and products outside the scope of this license are encouraged to contact the BIEN  
221 working group ([bien.working.group@gmail.com](mailto:bien.working.group@gmail.com)) to discuss alternative licensing options.

222

### 223 **Authorship guidelines**

224 Any publications using data obtained from BIEN should cite the original publications and  
225 acknowledge data providers when possible. Studies utilizing large amounts of BIEN data should  
226 seek solutions that ultimately lead to an entry in the Web of Science so that data providers and  
227 original publications are given credit. Normal ethics apply to co-authorship of scientific  
228 publications. Users of data stored in BIEN should consider inviting the original data contributor  
229 to be a co-author(s) of any resultant publications if that contributor's data are a major portion of  
230 the dataset analyzed, or if a data contributor makes a significant contribution to the analysis of  
231 the data or to the interpretation of results. Similarly, studies utilizing significant fractions of  
232 BIEN data and derived data products (species lists, geographic ranges, etc.) for their research are  
233 encouraged (but NOT required) to contact the BIEN working group, who may be able to provide  
234 insight into the use of BIEN data and expected updates to the database. In no circumstances  
235 should authorship be attributed to data contributors, individually or collectively, without their  
236 explicit consent.

237

### 238 **Future directions**

239 The BIEN database represents an unprecedented source of global botanical data that  
240 currently focuses on the New World. The BIEN database, workflow, and R package are under  
241 continuous development. We plan on expanding the BIEN database to increase coverage of  
242 other continents as well as incorporating additional types of trait data. Another priority is to



243 develop an API to allow easier access to the BIEN database by other programs. As development  
244 proceeds, we will add functionality to allow users to contribute their own data and range maps.  
245 Accessing these data will allow studies at an unprecedented scale with more complete, high  
246 quality data sets than smaller scale initiatives might achieve on their own. This is an important  
247 step toward improving our understanding of global patterns of biodiversity and developing  
248 predictive ecological theory at the largest of scales.

249

#### 250 **Authors' contributions:**

251 B.S.M, C.M, B.B. and B.J.E. wrote the manuscript. B.S.M. led development of the R package  
252 with assistance by B.B., C.M., D.G. and N.C. All authors worked on development of the BIEN  
253 database and workflow, contributed feedback on the manuscript, and approved the publication.

254

#### 255 **Acknowledgements:**

256 We thank Irena Šimová for input on the manuscript, Scott Chamberlain for valuable comments  
257 on R programming, our beta-testers for their input, and our CRAN reviewers for their cheerful  
258 comments and assistance. We thank Scott Chamberlain, Robert Freckleton, and an anonymous  
259 reviewer for their helpful comments on both the manuscript and the R package. N.M.-H.  
260 acknowledges support by a Villum Foundation Postdoctoral Fellowship. We gratefully  
261 acknowledge the Canada Foundation for Innovation for funding the infrastructure for  
262 Canadensys. This work was conducted as a part of the Botanical Information and Ecology  
263 Network (BIEN) Working Group (PIs BJE, Richard Condit, BB, SD, RKP) supported by the  
264 National Center for Ecological Analysis and Synthesis (funded by NSF Grant #EF-0553768), the  
265 University of California, Santa Barbara, and the State of California. The BIEN Working Group  
266 was also supported by iPlant (NSF #DBI-0735191). CV was supported by the French Foundation  
267 for Research on Biodiversity (FRB; [www.fondationbiodiversite.fr](http://www.fondationbiodiversite.fr)) in the context of the CESAB  
268 project “Assembling, analysing and sharing data on plant functional diversity to understand the  
269 effects of biodiversity on ecosystem functioning: a case study with French Permanent  
270 Grasslands” (DIVGRASS), by a Marie Curie International Outgoing Fellowship within the 7th  
271 European Community Framework Program (DiversiTraits project, no. 221060) and by the  
272 European Research Council (ERC) Starting Grant Project “Ecophysiological and biophysical  
273 constraints on domestication in crop plants” (Grant ERC-StG-2014-639706- CONSTRAINTS).

274 JCS was supported by the European Research Council (ERC-2012-StG-310886-HISTFUNC).  
275 SKW was supported by Core funding for Crown Research Institutes from the New Zealand  
276 Ministry of Business, Innovation and Employment's Science and Innovation Group. We would  
277 like to acknowledge the herbaria that contributed data to this work: A, AAH, AAS, AAU, ABH,  
278 ACAD, ACOR, AD, ADW, AFS, AHUC, AIMS, AJOU, AK, AKPM, ALCB, ALT, ALTA,  
279 ALU, AMD, AMES, AMNH, AMO, ANA, ANGU, ANSM, ANSP, ANUC, ARAN, ARC,  
280 ARIZ, ARM, AS, ASDM, ASU, ATCC, AUG, AUT, B, BA, BAA, BAB, BACP, BAF, BAFC,  
281 BAI, BAJ, BAL, BARC, BAS, BBB, BBS, BC, BCF, BCMEX, BCN, BCRU, BEREA, BG, BH,  
282 BHCB, BHO, BILAS, BIO, BISH, BLA, BM, BO, BOCH, BOG, BOL, BOLV, BONN, BOUM,  
283 BPI, BR, BRA, BREM, BRI, BRIT, BRIU, BRLU, BRM, BSB, BSIP, BSN, BTN, BUL, BULU,  
284 BUT, C, CALI, CAMU, CAN, CANB, CANL, CAS, CAY, CBG, CBM, CBS, CEN, CEPEC,  
285 CESJ, CGE, CHAM, CHAP, CHAPA, CHI, CHL, CHR, CHRB, CIB, CICY, CIIDIR, CIMI,  
286 CINC, CIQR, CLEMS, CLF, CM, CMC, CMMEX, CNHM, CNPO, CO, COA, COAH, COCA,  
287 CODAGEM, COFC, COL, COLO, CONC, CORD, CP, CPAP, CPUN, CR, CRAI, CRP, CS,  
288 CSU, CTES, CTESN, CU, CUVC, CUZ, CVRD, CWU, DAO, DAOM, DAV, DBN, DES, DLF,  
289 DMNH, DMU, DNA, DR, DS, DUKE, DUSS, E, EA, EAC, EBH, EBUM, ECH, ECU, EIF,  
290 EIU, EKY, EM, EMMA, ENCB, ENS, ERA, ESA, ESS, F, FAA, FAU, FB, FBCS, FCME,  
291 FCO, FCQ, FEN, FH, FHO, FI, FLAS, FLOR, FM, FR, FRP, FTG, FUEL, FURB, G, GB, GDA,  
292 GDAC, GE, GENT, GEO, GES, GH, GI, GJO, GLM, GMNHJ, GOET, GUA, GZU, H, HA,  
293 HAC, HAJB, HAL, HAM, HAMAB, HAO, HAS, HAST, HASU, HAW, HB, HBG, HBR,  
294 HCIB, HEID, HGI, HIP, HKU, HNHM, HNT, HO, HPL, HRCB, HRP, HSS, HSU, HU, HUA,  
295 HUAA, HUAL, HUAZ, HUEFS, HUEM, HUFU, HUSA, HUT, HXBH, HYO, IA, IAA, IAC,  
296 IAL, IAN, IB, IBGE, IBUG, ICEL, ICN, IEB, IFO, ILL, ILLS, IMSSM, INB, INEGI, INIF,  
297 INM, INPA, IPA, IPRN, ISC, ISL, ISTC, ISU, ITCV, ITMH, IZAC, IZTA, JACA, JBAG, JE,  
298 JEPS, JOE, JUA, JYV, K, KANU, KIEL, KMN, KMNH, KOELN, KOR, KPM, KSC, KSTC,  
299 KSU, KTU, KU, KUO, KYO, L, LA, LAE, LAF, LAM, LCR, LD, LE, LEB, LEMA, LG, LI,  
300 LIL, LINN, LISE, LISI, LISU, LKHD, LL, LM, LOJA, LOMA, LP, LPAG, LPB, LPD, LPS,  
301 LSU, LTR, LY, LYJB, LZ, M, MA, MAF, MAIC, MAK, MAN, MARY, MASS, MB, MBK,  
302 MBM, MBML, MCM, MCN, MCNS, MEL, MEN, MERL, MEXU, MFA, MFU, MG, MGC,  
303 MICH, MIL, MIN, MISS, MJG, MMMN, MNHM, MNHN, MO, MOL, MOR, MOSS, MPU,  
304 MPUC, MRSN, MSB, MSC, MSE, MSTR, MSUN, MT, MTMG, MU, MUB, MUCV, MVFA,

305 MVFQ, MVJB, MVM, MY, N, NA, NCSC, NCU, ND, NE, NEB, NHM, NHMC, NHT, NLH,  
306 NLU, NMB, NMC, NMCR, NMNL, NMR, NMSU, NMW, NO, NOU, NRCC, NSPM, NSW,  
307 NT, NUM, NWOSU, NY, O, OC, OCLA, ODU, OHN, OKL, OKLA, OMA, OS, OSA, OSC,  
308 OSH, OSN, OULU, OWU, OXF, P, PACA, PAR, PE, PEL, PENN, PERTH, PEUFR, PFC, PH,  
309 PI, PKDC, PLAT, PMA, PMNH, PNH, POLL, POM, PORT, PR, PRC, PRE, PTBG, PVNH,  
310 PY, QCA, QCNE, QFA, QM, QMEX, QRS, QUE, R, RAS, RB, RBR, REG, RENO, RFA,  
311 RIOG, RM, RNG, ROST, RPM, RSA, RYU, S, SALA, SAM, SAN, SANT, SAPS, SASK, SBT,  
312 SD, SEL, SEV, SF, SFSU, SGO, SI, SIM, SING, SIU, SJRP, SLPM, SMB, SMDB, SMF, SNM,  
313 SOM, SP, SPF, SPSF, SQF, SRFA, STL, STU, SUVA, SVG, SZU, TAES, TAI, TAIF, TAMU,  
314 TAN, TEF, TENN, TEPB, TEX, TFC, TFM, TI, TKPM, TNS, TO, TRA, TRH, TROM, TRT,  
315 TRTC, TRTE, TRTS, TS, TSM, TTRS, TU, TULS, TUR, U, UADY, UAM, UAMIZ, UARK,  
316 UAS, UAT, UB, UBA, UBC, UC, UCAM, UCBG, UCR, UEC, UESC, UFG, UFMA, UFMT,  
317 UFP, UFRJ, UFRN, UFS, UGDA, UH, UI, UJAT, ULM, ULS, UME, UMO, UNA, UNB,  
318 UNCC, UNEX, UNL, UNM, UNR, UNSL, UPGB, UPEI, UPNA, UPNG, UPS, US, USAS,  
319 USJ, USM, USNC, USON, USP, USZ, UT, UTC, UTEP, UTMC, UV, UVIC, UVSC, UWO, V,  
320 VA, VAL, VALD, VDB, VEN, VM, VMSL, VT, W, WAG, WAT, WELT, WFU, WII, WIN,  
321 WIS, WMNH, WOH, WRSL, WS, WTU, WU, XAL, Y, YA, YAM, YU, Z, ZMT, ZSS, ZT.

322

### 323 **Data Accessibility:**

324 The BIEN package is available for download from CRAN at: [https://CRAN.R-](https://CRAN.R-project.org/package=BIEN)  
325 [project.org/package=BIEN](https://CRAN.R-project.org/package=BIEN) . The development version of the package is available at:  
326 <https://github.com/bmaitner/RBIEN> .

327

### 328 **References**

329 Bininda-Emonds, O.R.P., Cardillo, M., Jones, K.E., MacPhee, R.D.E., Beck, R.M.D., Grenyer,  
330 R., Price, S. a., Vos, R. a., Gittleman, J.L. & Purvis, A. (2007). The delayed rise of present-  
331 day mammals. *Nature*, **446**, 507–512.

332 BirdLife International. (2013). IUCN Red List for birds. <http://www.birdlife.org>.

- 333 Blach-Overgaard, A., Svenning, J.-C., Dransfield, J., Greve, M. & Balslev, H. (2010).  
334 Determinants of palm species distributions across Africa: the relative roles of climate, non-  
335 climatic environmental factors, and spatial constraints. *Ecography*, **33**, 380–391.
- 336 Boyle, B., Brad, B., Nicole, H., Zhenyuan, L., Garay, J.A.R., Dmitry, M., Tony, R., Naim, M.,  
337 Narro, M.L., Piel, W.H., Mckay, S.J., Sonya, L., Chris, F., Peet, R.K. & Enquist, B.J.  
338 (2013). The taxonomic name resolution service: an online tool for automated  
339 standardization of plant names. *BMC Bioinformatics*, **14**, 16.
- 340 De Marco, P., Diniz-Filho, J.A.F. & Bini, L.M. (2008). Spatial analysis improves species  
341 distribution modelling during range expansion. *Biology Letters*, **4**, 577–580.
- 342 Enquist, B.J., Condit, R., Peet, R.K., Schildhauer, M. & Thiers, B.M. (2016).  
343 *Cyberinfrastructure for an integrated botanical information network to investigate the*  
344 *ecological impacts of global climate change on plant biodiversity*. PeerJ Preprints.
- 345 Enquist, B.J., Sandel, B., Boyle, B., Svenning, J.-C., McGill, B.J., Donoghue, J.C., Hinchliff,  
346 C.E., Jørgensen, P.M., Kraft, N.J.B., Marcuse-Kubitza, A., Merow, C., Morueta-Holme, N.,  
347 Peet, R.K., Schildhauer, M., Spencer, N., Regetz, J., Šímová, I., Smith, S.A., Thiers, B.,  
348 Violle, C., Wisser, S.K., Anelman, S., Casler, N., Condit, R., Dolins, S., Guaderrama, D.,  
349 Maitner, B.S., Narro, M.L., Ott, J.E., Phillips, O., Sloat, L.L. & ter Steege, H. (in prep.).  
350 Botanical big data shows that plant diversity in the New World is driven by climatic-linked  
351 differences in evolutionary rates and biotic exclusion.
- 352 Faurby, S. & Svenning, J.-C. (2015). Historic and prehistoric human-driven extinctions have  
353 reshaped global mammal diversity patterns. *Diversity and Distributions*, **21**, 1155–1166.
- 354 Global Biodiversity Information Facility. (2012). URL <http://www.gbif.org/> [accessed 11  
355 December 2012]
- 356 Hijmans, R.J., Cameron, S., Parra, J., Jones, P.G. & Jarvis, A. (2005). WorldClim, version 1.3.  
357 *University of California, Berkeley*.
- 358 Hinchliff, C.E. & Smith, S.A. (2014). Some limitations of public sequence data for phylogenetic  
359 inference (in plants). *PloS one*, **9**, e98986.

- 360 IUCN. (2013). IUCN Red List of Threatened Species. *Version 2013.2*. URL [IUCNredlist.org](http://IUCNredlist.org)
- 361 Jetz, W., Thomas, G.H.H., Joy, J.B.B., Hartmann, K. & Mooers, a. O.O. (2012). The global  
362 diversity of birds in space and tim. *Nature*, **491**, 1–5.
- 363 Kattge, J., Diaz, S., Lavorel, S., Prentice, I.C., Leadley, P., Bönisch, G., Garnier, E., Westoby,  
364 M., Reich, P.B., Wright, I.J. & Others. (2011). TRY--a global database of plant traits.  
365 *Global change biology*, **17**, 2905–2935.
- 366 Merow, C., Cory, M. & Silander, J.A. (2014). A comparison of Maxlike and Maxent for  
367 modelling species distributions. *Methods in Ecology and Evolution*, **5**, 215–225.
- 368 Merow, C., Cory, M., Smith, M.J. & Silander, J.A. (2013). A practical guide to MaxEnt for  
369 modeling species' distributions: what it does, and why inputs and settings matter.  
370 *Ecography*, **36**, 1058–1069.
- 371 Myhrvold, N.P., Baldrige, E., Chan, B., Sivam, D., Freeman, D.L. & Ernest, S.K.M. (2015). An  
372 amniote life-history database to perform comparative analyses with birds, mammals, and  
373 reptiles. *Ecology*, **96**, 3109–3109.
- 374 Phillips, S.J., Anderson, R.P. & Schapire, R.E. (2006). Maximum entropy modeling of species  
375 geographic distributions. *Ecological Modelling*, **190**, 231–259.
- 376 R Core Development Team. (2015). R: A language and environment for statistical computing. R  
377 Foundation for Statistical Computing. URL <https://www.R-project.org/>
- 378 Schulz, K. (2016). TraitBank: An open digital repository for organism traits. *2016 International*  
379 *Congress of Entomology*. Entomological Society of America.
- 380 Smith, S.A., Beaulieu, J.M. & Donoghue, M.J. (2009). Mega-phylogeny approach for  
381 comparative biology: an alternative to supertree and supermatrix approaches. *BMC*  
382 *evolutionary biology*, **9**, 37.
- 383 Smith, S.A. & O'Meara, B.C. (2012). treePL: divergence time estimation using penalized  
384 likelihood for large phylogenies. *Bioinformatics* , **28**, 2689–2690.

385 Stamatakis, A. (2006). RAxML-VI-HPC: maximum likelihood-based phylogenetic analyses with  
386 thousands of taxa and mixed models. *Bioinformatics* , **22**, 2688–2690.

387  
388  
389  
390  
391  
392  
393  
394  
395  
396  
397  
398  
399  
400  
401  
402  
403

**Table 1. Summary of Data Types in BIEN 3.4**

	<b>Count</b>	<b>Number of Species</b>
<b>Occurrence Records</b>	81,108,996	378,502
<b>Ecological Plots Observations</b>	17,244,737	16,216
<b>Trait Records</b>	916,988	93,251
<b>Taxonomic Records</b>	539,781	378,577
<b>Range Maps</b> (see text for details)	98,829	98,829

<b>Phylogenies</b> (see text for details)	100	81,274
--	-----	--------

404  
405  
406  
407  
408  
409  
410  
411  
412  
413  
414  
415  
416  
417  
418  
419  
420  
421  
422  
423  
424  
425  
426  
427  
428  
429  
430  
431  
432

Author Manuscript

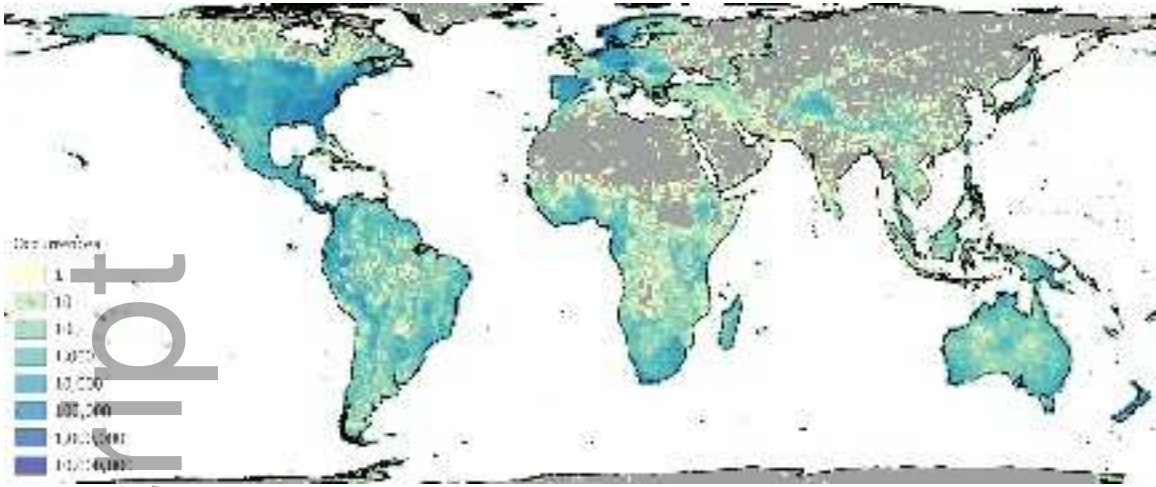
433 **Table 2. Summary of the number of occurrences within BIEN 3.4**

434

	Number observed within BIEN 3.4 (Global)	Number observed within BIEN 3.4 (New World)
Total Observations	81,108,996	28,311,777
Geographically validated Observations	46,432,247	22,462,543
Specimens	63,864,259	13,365,277
Plot observations	17,244,737	14,946,500
Plots	113,146	107,619
Species	378,502	117,808

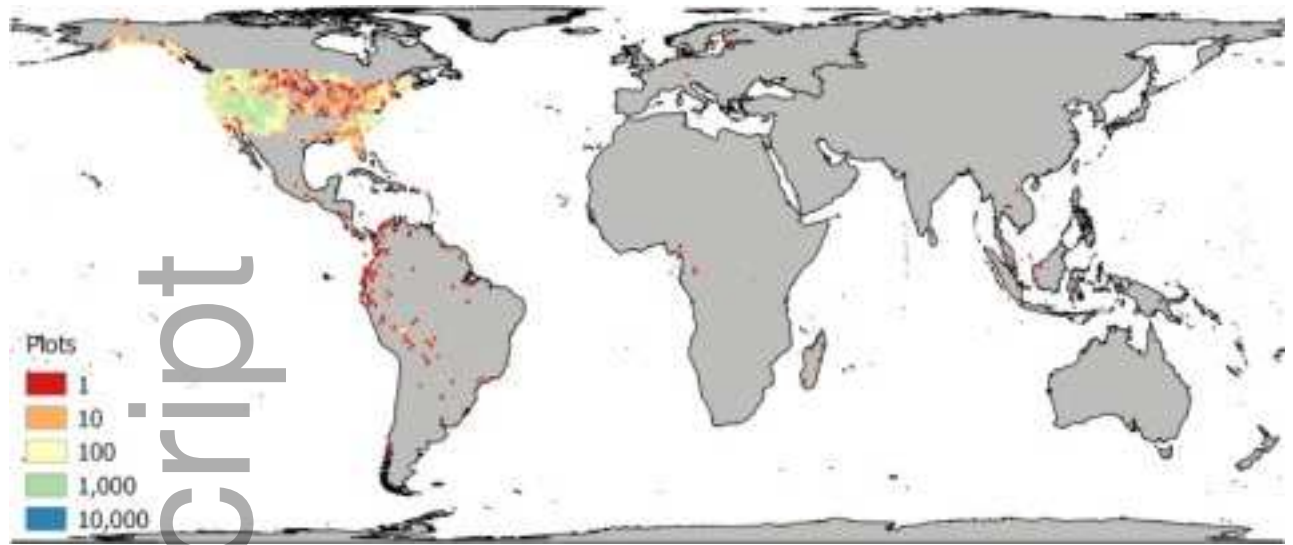
435





mee3\_12861\_f1.png

Author Manuscript



mee3\_12861\_f2.png

Author Manuscript