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Climate and socio-economic factors explain differences between observed and expected naturalization patterns of European plants around the world

Robin Pouteau, Wilfried Thuiller, Carsten Hobohm, Caroline Brunel, Barry J. Conn, Wayne Dawson, Michele de Sá Dechoum, Aleksandr L. Ebel, Franz Essl, Ori Fragman-Sapir, et al.

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1 **Research paper**

2 Climate and socio-economic factors explain differences between observed and expected
3 naturalization patterns of European plants around the world

4 **Authors**

5 Robin Pouteau*, Wilfried Thuiller, Carsten Hobohm, Caroline Brunel, Barry J. Conn, Wayne
6 Dawson, Michele de Sá Dechoum, Alexandr L. Ebel, Franz Essl, Ori Fragman-Sapir, Trevor
7 Fristoe, Nejc Jogan, Holger Kreft, Bernd Lenzner, Carsten Meyer, Jan Pergl, Petr Pyšek, Alla
8 Verkhovzina, Patrick Weigelt, Qiang Yang, Elena Zykova, Svetlana Aćić, Emiliano Agrillo,
9 Fabio Attorre, Ariel Bergamini, Christian Berg, Erwin Bergmeier, Idoia Biurrun, Steffen Boch,
10 Gianmaria Bonari, Zoltán Botta-Dukát, Helge Bruelheide, Juan Antonio Campos, Andraž
11 Čarni, Laura Casella, Maria Laura Carranza, Milan Chytrý, Renata Čušterevska, Michele De
12 Sanctis, Jürgen Dengler, Panayotis Dimopoulos, Rasmus Ejrnæs, Jörg Ewald, Federico
13 Fernández-González, Rosario G. Gavilán, Jean-Claude Gegout, Rense Haveman, Maike
14 Isermann, Ute Jandt, Florian Jansen, Borja Jiménez-Alfaro, Ali Kavgacı, Larisa Khanina,
15 Ilona Knollová, Anna Kuzemko, Maria Lebedeva, Jonathan Lenoir, Tatiana Lysenko, Corrado
16 Marcenò, Vasiliy Martynenko, Jesper Erenskjold Moeslund, Ricarda Pätsch, Remigiusz
17 Pielech, Valerijus Rašomavičius, Iris de Ronde, Eszter Ruprecht, Solvita Rūsiņa, Pavel
18 Shirokikh, Jozef Šibík, Urban Šilc, Angela Stanisci, Zvezdana Stančić, Jens-Christian
19 Svenning, Grzegorz Swacha, Pavel Dan Turtureanu, Milan Valachovič, Kiril Vassilev, Sergey
20 Yamalov, Mark van Kleunen

21 ***Corresponding author**

22 Zhejiang Provincial Key Laboratory of Plant Evolutionary Ecology and Conservation, Taizhou
23 University, Taizhou 318000, China¹

¹ Current address: AMAP, Univ Montpellier, CIRAD, CNRS, INRAE, IRD, Montpellier, France

24 E-mail: robin.pouteau@ird.fr

25 **Affiliations**

26 Wilfried Thuiller: Laboratoire d'Écologie Alpine (LECA), University Grenoble Alpes, CNRS,
27 University Savoie Mont Blanc, Grenoble, France – wilfried.thuiller@univ-grenoble-alpes.fr

28 Carsten Hobohm: Ecology and Environmental Education Working Group, University of
29 Flensburg (EUF), 24943 Flensburg, Germany – hobohm@uni-flensburg.de

30 Caroline Brunel: Zhejiang Provincial Key Laboratory of Plant Evolutionary Ecology and
31 Conservation, Taizhou University, Taizhou, China – caroline.brunel@ird.fr

32 Barry J. Conn (orcid: 0000-0002-7156-4802): School of Life and Environmental Sciences,
33 University of Sydney, Australia – barry.conn@sydney.edu.au

34 Wayne Dawson: Department of Biosciences, Durham University, Durham, UK –
35 wayne.dawson@durham.ac.uk

36 Michele de Sá Dechoum:

37 • Department of Ecology and Zoology, Graduate Course in Ecology, Federal
38 University of Santa Catarina, Florianopolis, SC, Brazil

39 • The Horus Institute for Environmental Conservation and Development.
40 Florianopolis, SC, Brazil – michele.dechoum@ufsc.br

41 Alexandr L. Ebel: Department of Botany, Tomsk State University, Tomsk, Russia – alex-
42 08@mail2000.ru

43 Franz Essl: Bioinvasions, Global Change, Macroecology-Group, Department of Botany and
44 Biodiversity Research, University of Vienna, Vienna, Austria – franz.essl@univie.ac.at

45 Ori Fragman-Sapir: Jerusalem Botanical Gardens, The Hebrew University of Jerusalem,
46 Jerusalem, Israel – ofragman@013.net

47 Trevor Fristoe: Ecology, Department of Biology, University of Konstanz, Konstanz, Germany
48 – trevor.fristoe@uni-konstanz.de

49 Nejc Jogan: Department of Biology, Biotechnical Faculty, University of Ljubljana, Ljubljana,
50 Slovenia – jernej.jogan@bf.uni-lj.si

51 Holger Kreft:

52 • Biodiversity, Macroecology & Biogeography, University of Göttingen, Göttingen,
53 Germany

54 • Centre of Biodiversity and Sustainable Land Use (CBL), University of Goettingen,
55 Germany – hkreft@uni-goettingen.de

56 Bernd Lenzner: Bioinvasions, Global Change, Macroecology-Group, Department of Botany
57 and Biodiversity Research, University of Vienna, Vienna, Austria –
58 bernd.lenzner@univie.ac.at

59 Carsten Meyer (orcid: 0000-0003-3927-5856):

60 • German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig,
61 Leipzig, Germany

62 • Institute of Biology, Leipzig University, Leipzig, Germany

63 • Institute for Geosciences and Geography, Martin Luther University Halle-
64 Wittenberg, Halle (Saale), Germany – carsten.meyer@idiv.de

65 Jan Pergl: Czech Academy of Sciences, Institute of Botany, Department of Invasion Ecology,
66 Průhonice, Czech Republic – jan.pergl@ibot.cas.cz

67 Petr Pyšek (orcid: 0000-0001-8500-442X):

68 • Czech Academy of Sciences, Institute of Botany, Department of Invasion Ecology,
69 Průhonice, Czech Republic – pysek@ibot.cas.cz

70 • Department of Ecology, Faculty of Science, Charles University, Prague, Czech
71 Republic

72 Alla Verkhozna: Siberian Institute of Plant Physiology & Biochemistry, Siberian Branch of
73 Russian Academy of Sciences, Irkutsk, Russia – allaverh@list.ru

74 Patrick Weigelt: Biodiversity, Macroecology & Biogeography, University of Göttingen,
75 Göttingen, Germany – patrick.weigelt@forst.uni-goettingen.de

76 Qiang Yang: Ecology, Department of Biology, University of Konstanz, Konstanz, Germany –
77 qiang.yang@uni-konstanz.de

78 Elena Zykova: Central Siberian Botanical Garden, Siberian Branch of Russian Academy of
79 Sciences, Novosibirsk, Russia – elena.yu.zykova@gmail.com

80 Svetlana Ačić (orcid: 0000-0001-6553-3797): University of Belgrade, Faculty of Agriculture,
81 Department of Botany, Belgrade, Serbia – acic@agrif.bg.ac.rs

82 Emiliano Agrillo (orcid: 0000-0003-2346-8346): Operational Center for Environmental
83 Monitoring - Institute for Environmental Protection and Research, Rome, Italy –
84 emiliano.agrillo@isprambiente.it

85 Fabio Attorre: Department of Environmental Biology, Sapienza University of Rome, Italy –
86 fabio.attorre@uniroma1.it

87 Ariel Bergamini (orcid: 0000-0001-8816-1420): WSL Swiss Federal Research Institute,
88 Zürcherstrasse 111, 8903 Birmensdorf, Switzerland – ariel.bergamini@wsl.ch

89 Christian Berg: Universität of Graz, Institute for Biology, Holteigasse 6, 8010 Graz, Austria –
90 christian.berg@uni-graz.at

91 Erwin Bergmeier (orcid: 0000-0002-6118-4611):

92 University of Göttingen, Department of Vegetation Science and Phytodiversity, 37073

93 Göttingen, Germany – erwin.bergmeier@bio.uni-goettingen.de

94 Centre of Biodiversity and Sustainable Land Use (CBL), University of Göttingen, Germany

95 Idoia Biurrun (orcid: 0000-0002-1454-0433): Department of Plant Biology and Ecology,

96 University of the Basque Country UPV/EHU, P.O. Box 644, 48080 Bilbao, Spain –

97 idoia.biurrun@ehu.es

98 Steffen Boch (orcid: 0000-0003-2814-5343) : WSL Swiss Federal Research Institute,

99 Zürcherstrasse 111, 8903 Birmensdorf, Switzerland – steffen.boch@wsl.ch

100 Gianmaria Bonari (orcid: 0000-0002-5574-6067):

101 • Department of Botany and Zoology, Faculty of Science, Masaryk University, Brno,

102 Czech Republic

103 • Faculty of Science and Technology, Free University of Bolzano, Bozen, Italy –

104 gianmaria.bonari@unibz.it

105 Zoltán Botta-Dukát (orcid: 0000-0002-9544-3474): Institute of Ecology and Botany, MTA

106 Centre for Ecological Research, Vácrátót, Hungary – botta-dukát.zoltan@okologia.mta.hu

107 Helge Bruelheide:

108 • Martin Luther University Halle-Wittenberg, Institute of Biology/Geobotany and

109 Botanical Garden, Am Kirchtor 1, 06108 Halle, Germany

110 • German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig,

111 Deutscher Platz 5e, 04103 Leipzig, Germany – helge.bruehlheide@botanik.uni-

112 halle.de

113 Juan Antonio Campos (orcid: 0000-0001-5992-2753): Department of Plant Biology and
114 Ecology, University of the Basque Country UPV/EHU, P.O. Box 644, 48080 Bilbao, Spain –
115 juanan.campos@ehu.eus

116 Andraž Čarni (orcid: 0000-0002-8909-4298) : Research Center of the Slovenian Academy of
117 Sciences and Arts, Novi trg 2, 1000 Ljubljana, Slovenia, University of Nova Gorica, Vipavska
118 13, 5000 Nova Gorica, Slovenia – carni@zrc-sazu.si

119 Laura Casella (orcid: 0000-0003-2550-3010): Italian National Institute for Environmental
120 Protection and Research (ISPRA), Department for Biodiversity Conservation and Monitoring,
121 Rome, Italy – laura.casella@isprambiente.it

122 Maria Laura Carranza (orcid: 0000-0001-5753-890X): Envixlab, Department of Biosciences
123 and Territory, University of Molise, C. da Fonte Lappone, 86090 Pesche, Italy –
124 carranza@unimol.it

125 Milan Chytrý (orcid: 0000-0002-8122-3075): Department of Botany and Zoology, Faculty of
126 Science, Masaryk University, Brno, Czech Republic – chytry@sci.muni.cz

127 Renata Čušterevska: Institute of Biology, Faculty of Natural Sciences and Mathematics,
128 University of Ss. Cyril and Methodius, 1000 Skopje, Republic of North Macedonia –
129 renatapmf@yahoo.com

130 Michele De Sanctis: Department of Environmental Biology, Sapienza University of Rome,
131 P.le Aldo Moro 5, 00185, Rome, Italy – michele.desanctis@uniroma1.it

132 Jürgen Dengler (orcid: 0000-0003-3221-660X):

- 133 • Vegetation Ecology, Institute of Natural Resource Sciences (IUNR), Zurich
- 134 University of Applied Sciences (ZHAW), Grüentalstr. 14, 8820 Wädenswil,
- 135 Switzerland

136 • Plant Ecology, Bayreuth Center of Ecology and Environmental Research
137 (BayCEER), University of Bayreuth, Universitätsstr. 30, 95447 Bayreuth,
138 Germany

139 • German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig,
140 Deutscher Platz 5e, 04103 Leipzig, Germany – dr.juergen.dengler@gmail.com

141 Panayotis Dimopoulos: University of Patras, Department of Biology, Laboratory of Botany,
142 26504 Patras, Greece – pdimopoulos@upatras.gr

143 Rasmus Ejrnæs: Section for Biodiversity, Department of Bioscience, Aarhus University,
144 Grenåvej 14, 8410 Rønne, Denmark – rasmus@bios.au.dk

145 Jörg Ewald (orcid: 0000-0002-2758-9324): Institute for Ecology and Landscape,
146 Weihenstephan-Triesdorf University of Applied Sciences, Hans-Carl-von-Carlowitz-Platz 3,
147 85354 Freising, Germany – joerg.ewald@hswt.de

148 Federico Fernández-González: Institute of Environmental Sciences, University of Castilla-La
149 Mancha, 45071 Toledo, Spain – Federico.Fdez@uclm.es

150 Rosario G. Gavilán (orcid: 0000-0002-1022-445X): Department of Pharmacology,
151 Pharmacognosy and Botany, Complutense University, E-28040 Madrid, Spain –
152 rgavilan@ucm.es

153 Giuliano Fanelli: Department of Environmental Biology, University of Rome Sapienza, Rome,
154 Italy – giuliano.fanelli@gmail.com

155 Jean-Claude Gegout: Université de Lorraine, AgroParisTech, INRAE, Silva, F-54000, Nancy,
156 France – jean-claude.gegout@agroparistech.fr

157 Rense Haveman: Central Government Real Estate Agency, Ministry of the Interior and
158 Kingdom Relations, Wageningen, the Netherlands – rens.haveman@wur.nl

159 Ute Jandt: (orcid: 0000-0002-3177-3669)

160 • Martin Luther University Halle-Wittenberg, Institute of Biology/Geobotany and
161 Botanical Garden, Am Kirchtor 1, 06108 Halle, Germany

162 • German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig,
163 Deutscher Platz 5e, 04103 Leipzig, Germany – ute.jandt@botanik.uni-halle.de

164 Maike Isermann:

165 • Institute of Ecology, Bremen University, FB 2, 28359 Bremen, Germany

166 • Lower Saxon Wadden Sea National Park Authority, Nature Conservation, 26382
167 Wilhelmshaven, Germany – maike.isermann@uni-bremen.de

168 Florian Jansen (orcid: 0000-0002-0331-5185): Faculty of Agricultural and Environmental
169 Sciences, University of Rostock, 18059 Rostock, Germany – florian.jansen@uni-rostock.de

170 Borja Jiménez-Alfaro (orcid: 0000-0001-6601-9597): Department of Organisms and Systems
171 Biology and Research Unit of Biodiversity (UMIB, UO-CSIC-PA), University of Oviedo,
172 Oviedo, Spain – jimenezalfaro.borja@gmail.com

173 Ali Kavgacı: Department of Forest Botany, Faculty of Forestry, University of Karabük,
174 Karabük, Turkey – alikavgaci1977@yahoo.com

175 Larisa Khanina: Institute of Mathematical Problems of Biology RAS, Branch of the M.V.
176 Keldysh Institute of Applied Mathematics of the Russian Academy of Sciences, Moscow,
177 Russia – khanina.larisa@gmail.com

178 Ilona Knollová: Department of Botany and Zoology, Faculty of Science, Masaryk University,
179 Brno, Czech Republic – ikuzel@sci.muni.cz

180 Anna Kuzemko (orcid: 0000-0002-9425-2756): Department of Geobotany and Ecology, M.G.
181 Kholodny Institute of Botany NAS of Ukraine; 2, Tereshchenkivska str., 01601, Kyiv, Ukraine
182 – anyameadow.ak@gmail.com

183 Maria Lebedeva: South-Ural botanical garden-institute, Ufa Federal Research Centre,
184 Russian Academy of science; Mendeleev str., 195/3, Ufa, Russia – lebedevamv@mail.ru

185 Jonathan Lenoir: UR “Ecologie et Dynamique des Systèmes Anthropisés” (EDYSAN, UMR
186 7058 CNRS), Université de Picardie Jules Verne, 1 Rue des Louvels, 80090 Amiens Cedex
187 1, France – jonathan.lenoir@u-picardie.fr

188 Tatiana Lysenko (orcid: 0000-0001-6688-1590):

189 • Komarov Botanical Institute of the Russian Academy of Sciences, Prof. Popov str.
190 2, 197376 Saint-Petersburg, Russia

191 • Institute of Ecology of the Volga River Basin of the Russian Academy of Sciences
192 – Branch of the Samara Scientific Center of the Russian Academy of Sciences,
193 Komzin str. 10, 445003 Togliatti, Russia – ltm2000@mail.ru

194 Corrado Marcenò (orcid: 0000-0003-4361-5200): Department of Botany and Zoology, Faculty
195 of Science, Masaryk University, Brno, Czech Republic – marcenocorrado@libero.it

196 Vasiliy Martynenko (orcid: 0000-0002-9071-3789): Laboratory of Geobotany and Plant
197 Resources, Ufa Institute of Biology of Ufa Federal Scientific Centre of Russian Academy of
198 Sciences, prospect Oktyabrya 69, 450054, Ufa, Republic of Bashkortostan, Russia –
199 vasmar@anrb.ru

200 Jesper Erenskjold Moeslund (orcid: 0000-0001-8591-7149): Section for Biodiversity,
201 Department of Bioscience, Aarhus University, Grenåvej 14, 8410 Rønne, Denmark –
202 jesper.moeslund@bios.au.dk

203 Ricarda Pätsch (orcid: 0000-0002-3349-0910): Wagner Ecology Lab, Department of
204 Biological Sciences, University of Alberta, Biological Sciences Building, Edmonton, Alberta,
205 T6G 2E9, Canada – ricarda.paetsch@gmail.com

206 Remigiusz Pielech (orcid: 0000-0001-8879-3305): Department of Forest Biodiversity, Faculty
207 of Forestry, University of Agriculture, al. 29 Listopada 46, 31-425 Kraków, Poland –
208 remekpielech@gmail.com

209 Valerijus Rašomavičius: Institute of Botany, Nature Research Centre, Vilnius, Lithuania –
210 valerijus.rasomavicius@botanika.lt

211 Iris de Ronde: Central Government Real Estate Agency, Ministry of the Interior and Kingdom
212 Relations, Wageningen, the Netherlands – iris.deronde@wur.nl

213 Eszter Ruprecht: Hungarian Department of Biology and Ecology, Babeş-Bolyai University,
214 Cluj-Napoca, Romania – eszter.ruprecht@ubbcluj.ro

215 Solvita Rūsiņa: Faculty of Geography and Earth Sciences, University of Latvia, Jelgavas iela
216 1, Rīga LV-1004, Latvia – Solvita.Rusina@lu.lv

217 Pavel Shirokikh (orcid: 0000-0003-1864-4878): Laboratory of Geobotany and Plant
218 Resources, Ufa Institute of Biology of Ufa Federal Scientific Centre of Russian Academy of
219 Sciences, prospect Oktyabrya 69, 450054, Ufa, Republic of Bashkortostan, Russia –
220 shirpa@mail.ru

221 Jozef Šibík (orcid: 0000-0002-5949-862X): Institute of Botany, Plant Science and Biodiversity
222 Center, Slovak Academy of Sciences, Dúbravská cesta 9, SK-84523 Bratislava, Slovakia –
223 jozef.sibik@savba.sk

224 Urban Šilc: Research Center of the Slovenian Academy of Sciences and Arts, Novi trg 2,
225 1000 Ljubljana, Slovenia – urban@zrc-sazu.si

226 Angela Stanisci (orcid: 0000-0002-5302-0932): Envixlab, Department of Biosciences and
227 Territory, University of Molise, Via Duca degli Abruzzi - 86039 Termoli (CB), Italy –
228 stanisci@unimol.it

229 Zvezdana Stančić (orcid: 0000-0002-6124-811X): Faculty of Geotechnical Engineering,
230 University of Zagreb, Hallerova aleja 7, 42000 Varaždin, Croatia – zvezdana.stancic@gfv.hr

231 Jens-Christian Svenning (orcid: 0000-0002-3415-0862): Center for Biodiversity Dynamics in
232 a Changing World (BIOCHANGE) and Section for Ecoinformatics and Biodiversity,
233 Department of Biology, Aarhus University, Ny Munkegade 114, DK-8000 Aarhus C, Denmark
234 – svenning@bio.au.dk

235 Grzegorz Swacha (orcid: 0000-0002-6380-2954): Botanical Garden, University of Wrocław,
236 Wrocław, Poland – grzegorz.swacha@uwr.edu.pl

237 Pavel Dan Turtureanu: A.Borza Botanical Garden, Center for Systematic Biology,
238 Biodiversity and Bioresources - 3B, Babes-Bolyai University, Cluj-Napoca, Romania –
239 pavel.turtureanu@ubbcluj.ro

240 Milan Valachovič (orcid: 0000-0001-5296-5240): Institute of Botany, Plant Science and
241 Biodiversity Center, Slovak Academy of Sciences, Dúbravská cesta 9, SK-84523 Bratislava,
242 Slovakia – milan.valachovic@savba.sk

243 Kiril Vassilev: Institute of Biodiversity and Ecosystem Research, Bulgarian Academy of
244 Sciences, Sofia, Bulgaria – kiril5914@abv.bg

245 Sergey Yamalov: South-Ural botanical garden-institute, Ufa Federal Research Centre,
246 Russian Academy of science; Mendeleev str., 195/3, Ufa, Russia – yamalovsm@mail.ru

247 Mark van Kleunen:

- 248 • Zhejiang Provincial Key Laboratory of Plant Evolutionary Ecology and
249 Conservation, Taizhou University, Taizhou, China
- 250 • Ecology, Department of Biology, University of Konstanz, Konstanz, Germany –
251 mark.vankleunen@uni-konstanz.de

252

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269 **Data availability statement**

270 The data used to build the species distribution models (occurrence records and
271 environmental variables) came from openly accessible repositories cited in the manuscript
272 (for details on the EVA database, see <http://euroveg.org/eva-database-obtaining-data>). All
273 1,485 generated habitat suitability raster files are available at
274 <https://dataverse.ird.fr/dataset.xhtml?persistentId=doi:10.23708/RNGS8Z>.

275 **Research paper**

276 Climate and socio-economic factors explain differences between observed and expected
277 naturalization patterns of European plants around the world

278 **Running title:** Observed versus expected naturalizations

279 **Abstract**

280 *Aim* – The number of naturalized (i.e., established) alien species has increased rapidly over
281 the last centuries. Given the differences in environmental tolerances among species, little is
282 known about what factors determine the extent to which the observed size of the naturalized
283 range of a species and hence the extent to which the observed richness of naturalized
284 species of a region approach their full potential. Here, we asked which region- and species-
285 specific characteristics explain differences between observed and expected naturalizations.

286 *Location* – Global.

287 *Time period* – Present.

288 *Major taxa studied* – Vascular plants.

289 *Methods* – We determined the observed naturalized distribution outside Europe for 1,485
290 species endemic to Europe using the GloNAF database and their expected distributions
291 outside Europe using species distribution models. First, we investigated which of seven
292 socio-economic factors related to introduction pathways, anthropogenic pressures and
293 inventory effort best explained the differences between observed and expected naturalized
294 European floras. Second, we examined whether distributional features, economic use and
295 functional traits explain the extent to which species have filled their expected ranges outside
296 Europe.

297 *Results* – In terms of suitable area, **more than 95%** of expected naturalizations of European
298 plants were not yet observed. Species were naturalized in only 4.2% of their suitable regions

299 outside of Europe (range filling) and in 0.4% of their unsuitable regions (range expansion).
300 Anthropogenic habitat disturbance primarily explained the difference between observed and
301 expected naturalized European floras, as did the number of treaties relevant to invasive
302 species. Species of ornamental and economic value and with large specific leaf area
303 performed better at filling and expanding beyond their expected range.

304 *Main conclusions* – The naturalization of alien plant species is explained by climate matching
305 but also by the regional level of human development, the introduction pressure associated
306 with the ornamental and economic values of the species and their adaptation to disturbed
307 environments.

308

309 **Keywords** – Alien species, anthropogenic pressure, environmental driver, functional trait,
310 global change, introduction pathway, naturalization, ornamental plant, sampling bias, species
311 distribution model.

312 **Introduction**

313 Species that maintain self-sustaining populations outside their native range independently of
314 direct human intervention (i.e., naturalized species; Richardson *et al.*, 2000; Blackburn *et al.*,
315 2011) pose a serious concern to native biodiversity and ecosystem functioning and services
316 worldwide (IPBES, 2019). Their number has increased substantially over the last few
317 centuries and does not show any indication of saturation (Seebens *et al.*, 2017). Regions
318 differ considerably in their richness of naturalized species of vascular plants, and species
319 vary considerably in their extent of naturalization (Pyšek *et al.*, 2017, Essl *et al.*, 2019). Part
320 of this variation may result from regional differences in the size of the potential naturalized
321 alien pool and the size of the range potentially available for naturalized species. Little is
322 known about the relationship of the observed and potential richness of naturalized plant
323 species (i.e. the maximum numbers of alien species that these regions could host given
324 species' environmental tolerances; Stohlgren *et al.*, 2008).

325 The actual distribution of naturalized alien plant species results from [the number of](#)
326 [introductions by humans \(intentional or not\) followed by](#) the interplay of factors that
327 determine the environmental space available for a species and factors that prevent or
328 promote the extent to which an equilibrium with an environmental space is reached.
329 Environmental suitability (e.g., climate, habitat) determines the potential (expected)
330 distribution range by filtering the subset of introduced species that can maintain populations
331 over long periods by successfully reproducing, i.e., naturalized species (Richardson & Pyšek,
332 2012). Human activities can further influence the observed distribution range of species by
333 altering environmental conditions and dispersal processes that dictate opportunities for
334 introductions. For instance, human transport, trade and increasing urban development tend
335 to broaden observed ranges, while, conversely, biosecurity responses can contain this
336 expansion (Hulme, 2009).

337 An increasing body of literature is investigating what proportions of species' potential alien
338 ranges are actually occupied (Petitpierre *et al.*, 2012; Strubbe *et al.*, 2013, 2015; Hill *et al.*,
339 2017; Liu *et al.*, 2020). Indeed, most alien species are occupying only a subset of the niche
340 corresponding to suitable environments that are occupied in their native range ('niche filling').
341 Certain alien species are also able to colonize new environments that are not occupied in
342 their native range ('niche expansion'). Patterns of niche filling and niche expansion vary
343 highly depending on the type of organisms. For instance, the average niche filling was found
344 to be 80% for vascular plants considered invasive (i.e., spreading into areas away from sites
345 of introduction; Richardson *et al.*, 2000), 37% for invasive insects and 20% for invasive birds,
346 whereas the average niche expansion was 26% for birds, 20% for insects and 5% for plants
347 (Petitpierre *et al.*, 2012; Strubbe *et al.*, 2013; Hill *et al.*, 2017). Niche changes during
348 invasions appear to be primarily driven by introduction history (e.g., propagule pressure and
349 colonization processes) in birds, and by human disturbance in insects (Hill *et al.*, 2017).
350 However, little is known about what causes possible differences between observed and
351 expected ranges in naturalized plants.

352 Much of the research effort has been devoted to understanding the socio-economic drivers of
353 the number of naturalized alien plant species (Pyšek *et al.*, 2010; van Kleunen *et al.*, 2015;
354 Dawson *et al.*, 2017; Essl *et al.*, 2019), some of them focusing on the subset of invasive
355 species (Essl *et al.*, 2015; Bellard *et al.*, 2016). However, studies dealing with alien species
356 richness always face the challenge of disentangling the specific role of socio-economic
357 drivers from that of large-scale biogeographical factors also known to influence species
358 richness, irrespective of species origin (e.g., latitudinal gradient, resource availability).
359 Moreover, while it is recognized that environmental matching between native and alien
360 ranges largely controls naturalization (Richardson & Pyšek, 2012; Feng *et al.*, 2016),
361 focusing on alien species richness does not allow accounting for species-specific
362 environmental requirements. Hence, determining the expected distributional range of

363 naturalized species should be an important step in understanding the influence of socio-
364 economic factors on observed patterns of naturalization (Xu *et al.*, 2019).

365 Another advantage of analyzing species composition is that it allows for the identification of
366 the species that explain the pattern of naturalized species richness and their characteristics.
367 Many other studies have attempted to determine which species' performance or natural
368 distribution traits are associated with naturalization success by comparing naturalized and
369 non-naturalized plant species (e.g., Pyšek & Richardson, 2007; van Kleunen *et al.*, 2010;
370 Pyšek *et al.*, 2015). Conclusions from these studies delivered little consensus and often
371 appear to be context- and species-dependent. This might partly be because these studies
372 only investigated trait differences of native versus alien or non-invasive versus invasive
373 species (i.e., a binary variable), or tested for relationships between traits and the observed
374 naturalized range without considering the potential range. Therefore, we believe we can gain
375 additional insight into the drivers of naturalization patterns by searching for traits that explain
376 the extent to which species fill or expand beyond their suitable alien range (i.e., a continuous
377 variable that accounts for the potential range).

378 Europe is the second-most important donor continent of naturalized alien plant species, after
379 Asia. However, the European flora is by far the most successful in naturalizing. Of the
380 ~14,200 vascular plant species [native \(i.e. both endemic and non-endemic\)](#) to Europe, 3,383
381 have become naturalized somewhere in the world; this is three times more than expected by
382 chance (van Kleunen *et al.*, 2015). Furthermore, five of the 10 globally most widely
383 naturalized species are native to Europe (Pyšek *et al.*, 2017). The European flora is therefore
384 suitable for studying drivers of differences between observed and expected naturalized
385 ranges.

386 Current knowledge of which region- and plant-related characteristics influence the extent to
387 which species fill their expected naturalized ranges is poor and relies only on the search of
388 the socio-economic drivers of niche/range filling of a few invasive alien species (usually < 50;

389 e.g., Hill *et al.*, 2017; Petitpierre *et al.*, 2012; Strubbe *et al.*, 2013, 2015). However, the
390 drivers of niche/range filling have never been explored for naturalized species (a much
391 broader group of species than invasives; Richardson *et al.*, 2000); yet the processes driving
392 the ability to survive and reproduce are likely to differ from the processes driving the spread
393 and impact of invasive species (e.g., Abellán *et al.*, 2017). Moreover, the drivers of the
394 difference between observed and expected naturalized regional floras (and not between
395 observed and expected distributions of a few individual species) have never been explored.

396 In this study, we focused on 1,485 plant species endemic to Europe, an unprecedented
397 number for this type of study, which allowed us to analyse in more detail what explains
398 variation in the differences between observed and expected values. We determined both
399 their current observed distribution in 931 non-European regions using the 'Global Naturalized
400 Alien Flora' database (GloNAF; van Kleunen *et al.*, 2019) and their expected distribution
401 using biogeoclimatic ensemble species distribution models (SDMs). We also explored which
402 socio-economic factors explain why some regions have a recorded naturalized flora of
403 European origin close to their expected naturalized flora, while other regions have not.
404 Finally, we examined whether species distribution patterns, economic uses and functional
405 traits explain why some species fill a larger portion of their expected range than others do.

406 **Materials and methods**

407 *Expected naturalized ranges*

408 Species selection

409 Many plant species native to Europe are also native to other continents such as Asia and
410 Africa, for which there is usually a lower density (temporal and spatial) of species occurrence
411 records (Meyer *et al.*, 2016). Therefore, we focused on vascular plant species whose native
412 range is restricted to Europe (European endemics), as this allows us to better capture the full
413 realized niche of the species in their native range. The physiography of the European
414 continent was defined as bordered by the Arctic Ocean to the north (Iceland and Norwegian

415 Islands were included, Greenland was excluded), the Atlantic Ocean to the west (the British
416 and Irish Isles were included and the Macaronesian archipelagos were excluded), the Ural
417 Mountains, the Ural River, and the Caspian Sea to the east, and the Caucasus and the
418 Mediterranean Sea to the south (Mediterranean islands were included, Anatolia was
419 excluded).

420 The database 'Endemic vascular plants in Europe' (EvaplantE; Hobohm 2014), which
421 comprises a list of > 6,200 endemic taxa, was used as a baseline for species selection.
422 Scientific names were standardized based on a working list of all plant species (The Plant
423 List; TPL; <http://www.theplantlist.org/> version 12-10-2018). This taxonomic standardization
424 was done with the R package 'Taxonstand' (Cayuela *et al.*, 2017). Standardized infraspecific
425 taxa were excluded from the list, resulting in 4,965 species (Figure 1).

426 Compilation of species occurrence records

427 To comprehensively describe the distribution of the species in Europe (which in addition to
428 the native range might for some species also include part of their alien range), we combined
429 occurrence records from six sources (see Supporting Information Appendix S1 for details
430 and references): the 'Global Biodiversity Information Facility' (GBIF), the 'European
431 Vegetation Archive' (EVA; Chytrý *et al.*, 2016), the 'EU-Forest' dataset, the 'Atlas Florae
432 Europaeae', the 'Plant Functional Diversity of Grasslands' network (DIVGRASS) and the
433 digital atlas of the German flora.

434 When several occurrence records from these different sources were duplicated on the same
435 cell, only one occurrence record per species was kept to avoid pseudoreplication. Cell size
436 was set to $0.42^\circ \times 0.42^\circ$ (~50 km \times 50 km at the equator) to approach the resolution of the
437 source of occurrence records having the coarsest resolution (Atlas Florae Europaeae). Using
438 a higher resolution (i.e., a smaller cell size) would have required to downscale this source of
439 occurrence records, which typically introduces spatial biases and uncertainty in the model
440 predictions (Bonbi & D'Amen, 2012). Moreover, this resolution appeared to be a good

441 compromise to account for the fact that the selected environmental variables (climate, land
442 use and soil type) determine invasion potential at different scales (10,000 to 200 km, 2,000 to
443 10 km, and 10 km to 10 m, respectively; Milbau *et al.*, 2009).

444 Species with fewer than 10 deduplicated occurrence records were not further considered
445 since the resulting SDM might be inaccurate. The final dataset comprised 135,189
446 occurrence records for a total of 1,485 European plant species, belonging to 327 genera and
447 67 families (Table S1), i.e. on average 91 occurrence records per species with a maximum of
448 1,382 occurrence records for *Achillea ptarmica* (Asteraceae). We had enough occurrence
449 records to build an SDM (i.e., 10 occurrences) for 272 European species among the 407
450 already naturalized species and for 1,213 species currently not known to be naturalized
451 anywhere.

452 Environmental predictors of expected ranges

453 We defined six environmental variables to model and project species expected ranges.
454 These variables were related to climate, land use, and soil physico-chemical properties,
455 which are commonly recognized to shape the distribution of plants (Gurevitch *et al.*, 2006).
456 Annual mean temperature (°C), annual precipitation (mm) and precipitation seasonality
457 (yearly coefficient of variation) representing the period 1979-2013 were provided at a 30
458 arcsec resolution by the CHELSA climate database (Karger *et al.*, 2017). Worldwide
459 correlations between these variables and the others described below did not exceed the
460 threshold of $|r| = 0.70$ (Supporting Information Appendix Figure S1) beyond which collinearity
461 begins to severely distort model estimations and subsequent predictions (Dormann *et al.*,
462 2013). However, these variables reflecting trends in average climate conditions were
463 significantly correlated with climate extremes to which plants are recognized to be highly
464 responsive (Zimmermann *et al.*, 2009). The percentage of each grid cell with primary land
465 cover based on the Harmonized Global Land Use models was also used (Chini *et al.*, 2014).
466 Organic carbon content (g per kg) and pH in the first 15 cm of soil were extracted at a 1 km

467 resolution from the global gridded soil information database SoilGrids (Hengl *et al.*, 2014).
468 Environmental variables were aggregated (using the mean value) to the resolution of 0.42° ×
469 0.42°.

470 Species distribution modelling

471 Comparing the realized niche within the native versus alien ranges can lead to three cases:
472 1) in the alien range, the species uses a similar or smaller realized niche than in the native
473 range, e.g., because of new competitors, herbivores or pathogens, or dispersal limitation; 2)
474 the species occupies a realized niche very different from the one in the native area but within
475 the species initial fundamental niche, e.g., because of new biotic interactions, multiple sites
476 of introduction, niche differentiation or different environmental conditions; or 3) the realized
477 niche extends outside of the species initial fundamental niche, e.g., because of rapid niche
478 evolution, enemy-release or new positive interactions (see Gallien *et al.*, 2010 for further
479 details). As options 2 and 3 remain impossible to predict for a large number of species, we
480 only considered option 1 in this study.

481 The expected distribution of the 1,485 European plant species was modelled by statistically
482 relating the environmental predictors to the distribution data in Europe. Six species
483 distribution modelling (SDM) methods including generalized additive models, generalized
484 linear models, generalized boosting trees, maximum entropy, multivariate adaptive
485 regression splines and random forest were used. All of these methods require presence and
486 absence or pseudo-absence/background data (a random subset of the available
487 environmental conditions in the area, i.e., Europe in our case), whose selection can
488 significantly affect predictions if not made adequately. Consequently, we performed a
489 preliminary analysis to identify the most appropriate set of pseudo-absences for each SDM
490 method (Appendix S2). The predictive performance of SDM in Europe was assessed by
491 measuring the area under the receiver operating characteristic (ROC) curve (AUC) and the
492 true skill statistics (TSS; Allouche *et al.*, 2006). AUC values can range from 0 to 1, an AUC of

493 1 indicating a perfect fit, an AUC of 0.5 meaning that predictions from SDM do not differ from
494 random, and an AUC of 0 meaning the SDM is always incorrect. TSS ranges from -1 to 1,
495 where 1 indicates perfect agreement and 0 indicates a random prediction.

496 To combine the predictive capability of the six SDMs, their projections were subsequently
497 aggregated into an average or consensus projection. To ensure the quality of the ensemble
498 SDM, we only kept the projections for which the accuracy estimated by AUC and TSS were
499 higher than 0.8 and 0.6, respectively, and assembled the selected SDMs using a committee-
500 average approach with each SDM was weighted proportional to its TSS evaluation.

501 Probability maps obtained from the ensemble forecasting projections were then transformed
502 into binary suitable/non-suitable maps using the threshold maximizing the TSS to ensure the
503 most accurate predictions, since it is based on both sensitivity and specificity. The entire
504 species distribution modelling workflow was performed within the 'biomod2' R platform
505 (Thuiller *et al.*, 2009).

506 *Recorded naturalized ranges*

507 The current observed naturalized range of the European flora was obtained through the
508 GloNAF database version 1.2, a recently compiled database of ~14,000 naturalized alien
509 plant species covering 1,029 regions worldwide, corresponding to administrative regions
510 such as countries, states and provinces, and also including 381 islands (van Kleunen *et al.*,
511 2019). The size of a region ranged from 0.03 km² (Tauna islet of the Gambier archipelago,
512 French Polynesia) to 2,486,952 km² (the Republic of the Sudan plus South Sudan). A total of
513 407 European [endemic](#) species were found as naturalized in at least one of the non-
514 European GloNAF regions.

515 We explored various definitions of suitability by analyzing what proportion of a region should
516 be suitable for a species to become naturalized in this region. The following definitions were
517 used: at least one suitable cell, 1%, 5%, 20%, 50%, or 100% suitable cells, respectively. For
518 each region, species were classified into: true positives (TP), i.e., species that are both

519 observed and predicted as being successfully naturalized in the region; false positives (FP),
520 i.e., species that have not become naturalized, but are predicted as successfully naturalized;
521 false negatives (FN), i.e., species that are observed as successfully naturalized but not
522 predicted to be naturalized; and true negatives (TN), i.e., species that are neither predicted
523 nor observed to be naturalized in the region. Then, the predicted naturalized species pool
524 was compared to the observed naturalized pool by calculating: 1) the naturalization debt in
525 the broad sense hereafter referred to as 'naturalization debt' ($FP/(TP+FP)$), i.e., the
526 proportion of predicted naturalizations that are not yet observed either because species are
527 not yet introduced to the focal area (introduction debt) or because they were introduced, but
528 are not naturalized yet (naturalization debt in the strict sense, i.e., *sensu Rouget et al., 2016*),
529 which can take several decades or centuries for some taxa; 2) the assemblage sensitivity
530 ($TP/(TP+FN)$), i.e., the proportion of observed naturalizations that are correctly predicted ; 3)
531 the negative predictive value ($TN/(TN+FN)$), i.e., the proportion of species that were not
532 predicted to be naturalized and have not become naturalized in the region.

533 *Socio-economic drivers of naturalization patterns*

534 Seven factors were extracted to explain naturalization debt, assemblage sensitivity and the
535 negative predictive value (Table 1). These factors were selected because they provided
536 insights into introduction pathways, human pressures and inventory effort: airport density
537 ('Airports'), seaport density ('Seaports'), number of international treaties relevant to invasive
538 alien species ('Treaties'), human population density ('Population'), proportion of croplands
539 ('Cropland'), human development index ('Development') and species inventory effort
540 ('Inventory'; for details, see Appendix S3). These socio-economic factors were independent
541 of the surface area of the regions. They were available for 526 non-European regions
542 including 17 islands or archipelagos and 509 mainland areas. Regional factors were tested
543 for multicollinearity by computing a correlation matrix based on Pearson's r . Cross-
544 correlations did not exceed $r = 0.51$ (between airport and seaport density), which is below the
545 threshold of 0.70 (Figure S3; Dormann *et al.*, 2013).

546 Identification of the drivers of naturalization patterns was based on boosted regression trees
547 (BRTs), a machine-learning method that combines a large number of relatively simple tree
548 models to optimize predictive performance (Elith *et al.*, 2008). The quality of BRT fits was
549 controlled with the coefficient of determination of the regression between estimated and
550 observed values (r^2) and the standard error (se). Results were interpreted by looking at the
551 relative influence of the seven regional socio-economic factors to the predictive models
552 (calculated based on the number of times a factor is selected in the model, weighted by its
553 improvement to the overall model) and by considering the partial dependence of the
554 predictions on each factor after accounting for the average effect of the other factors.

555 *Plant features explaining naturalization patterns*

556 To assess which species-level features affect naturalization success, we first compared
557 European species recorded as naturalized outside Europe ('naturalized') and those not
558 recorded as naturalized outside Europe ('non-naturalized'). Eight features including two
559 distributional indices, the economic use of the species and four functional traits were
560 selected: 1) species prevalence in Europe measured as the number of $0.42^\circ \times 0.42^\circ$ cells
561 occupied; 2) the surface area of the species' expected range outside Europe. These
562 distributional indices were used because more widespread species are usually hypothesized
563 to have a greater potential to spread and a higher invasion success (Gallien *et al.*, 2019); 3)
564 whether the species is used as an ornamental plant or 4) has other economic relevance
565 (e.g., food plant, medicines, materials) according to the World Economic Plants (WEP)
566 database (Wiersema & León, 2013) since the ornamental and economic relevance of a
567 species is thought to be a key predictor of its probability to be introduced and its
568 naturalization success (van Kleunen *et al.*, 2018, 2020); 5) the growth form (i.e., graminoid,
569 non-graminoid herb, shrub or tree), a major determinant of invasiveness (Pyšek &
570 Richardson, 2007); 6) mean specific leaf area (SLA), i.e., the ratio of leaf area to leaf dry
571 mass (expressed in $\text{m}^2\cdot\text{g}$); 7) mean plant height (in m); and 8) mean seed mass (in g). These
572 last three functional traits represent key axes of plant ecological strategies following the leaf–

573 height–seed (LHS) scheme of Westoby (1998). Correlation between LHS traits was $|r| <$
574 0.27. Functional traits were extracted from the TRY database (Kattge *et al.*, 2020). Linear
575 mixed-effect models (LMMs) were fitted using a phylogenetic generalized least square
576 approach (PGLS) to compare features of naturalized and non-naturalized species while
577 controlling for between-species phylogenetic distances (for details on the phylogeny used,
578 see Appendix S4). As we did not have data on all features for each species (see Figure 6),
579 we ran separate LMMs for each of the eight features. LMMs were performed using the *gls()*
580 function of the R package ‘nlme’ (Pinheiro *et al.*, 2019).

581 We used LMMs accounting for phylogeny to test whether plant-feature values explain: 1) the
582 relative extent of range filling, i.e., the proportion of suitable regions where a species has
583 become naturalized; and 2) the relative extent of range expansion, i.e., the proportion of
584 unsuitable regions where a species has nevertheless become naturalized. The indices of
585 range filling and expansion that we used are similar to the indices used by Petitpierre *et al.*
586 (2012). However, we calculated them in the geographical space instead of in the
587 environmental space because occurrence data on naturalized plants are not evenly
588 comprehensive across regions (which would have been needed to offer a reliable fit of the
589 niches), and when an alien species is reported, we do not always know whether or not it
590 reproduces in the wild (Figure 1).

591 **Results**

592 *Socio-economic drivers of naturalization patterns*

593 European endemic vascular plant species have naturalized in 319 non-European regions.
594 The naturalization debt (i.e., the proportion of species which are not yet naturalized in a
595 particular region but likely to become so, *if introduced*, given their environmental envelope)
596 *ranged from 95 to 100% (mean = 99%) among the 931 non-European regions according to*
597 *the threshold of one pixel and from 71 to 100% (mean = 99% as well) according to the*
598 *threshold of 100% of suitable cells (Figure 2). This means that less than 5% and less than*

599 29% of species for which a region is suitable according to the lowest and highest thresholds,
600 respectively, have actually been recorded there. When restricting the analysis to 272
601 European species already naturalized outside of Europe, the naturalization debt decreased
602 but remained relatively high with values in the range 81-100% (mean = 96%) with the former
603 threshold and 52-100% (95%) with the latter threshold.

604 The proportion of observed naturalizations that are correctly predicted based on
605 environmental matching (sensitivity) ranged from 0 to 100% for all thresholds and both
606 species sets. However, the average value of sensitivity was two percentage points lower for
607 the full set of species (19-64%) than for the subset of already naturalized species (21-66%).
608 The average proportion of unsuccessful naturalizations in unsuitable environments (negative
609 predictive value) was between 98.8% and 100% (mean = 99.9%) with the lowest threshold,
610 and between 97.0% and 100% (mean = 99.7%) with the highest threshold for the full set of
611 species. When considering already naturalized species only, the negative predictive value
612 decreased substantially to the range 80.0-100% (mean = 99.4%) in the former case and
613 83.4-100% (mean = 98.8%) in the latter case. The threshold of 1% of suitable cells offered
614 the best solution for maximizing sensitivity while still having a high negative predictive value
615 when comparing observed and expected naturalized floras. Therefore, we used predictions
616 from this threshold in the subsequent analyses.

617 Naturalization debt was lower in more developed regions (human development index > 0.7)
618 (Figures 3 and 4) with moist and cool climates, including coastal North America, Japan,
619 temperate Australia and New Zealand (Figure 5). Sensitivity increased progressively with the
620 level of development and inventory effort (Figures 3 and 4), and reached maximum values in
621 North-East America, the South American Atlantic Forest and the southern Andes, Turkey, the
622 East-Central African mountains, South Africa, Japan, temperate Australia and New Zealand
623 (Figure 5). The negative predictive value was lower in more developed regions (human
624 development index > 0.7), with more treaties relevant to biological invasions (number of
625 treaties > 20; Figures 3 and 4), including regions in North America, the southern tip of South

626 America and Africa, most of northern Asia excluding Mongolia, temperate Australia and New
627 Zealand (Figure 5).

628 *Plant features explaining naturalization patterns*

629 Compared to plant species that are currently not known to be naturalized somewhere else
630 (1,213), those that are naturalized (272) tend to be more widespread in Europe, to have a
631 wider expected range outside Europe, and to have an ornamental or other economic use
632 (Figure 6). There was no significant difference in functional traits between naturalized and
633 non-naturalized species.

634 On average, species were found to be naturalized in 4.2% of their potentially suitable regions
635 outside Europe (Figure S4). Thirty of the species occupied less than 0.1% of suitable regions
636 and 31 species occupied more than 10% of suitable regions. The species with the most
637 complete range filling was *Syringa vulgaris* (Oleaceae), native to the Balkan Peninsula and
638 Romania, being currently naturalized in 70% of its suitable regions (Table S2). The average
639 range expansion into regions estimated to be unsuitable was 0.4%. 118 species (43%)
640 exhibited range expansion above 0.1%, but only one species had a range expansion above
641 10%: *Verbascum virgatum* (Scrophulariaceae), native to south-western Europe and southern
642 England, and currently naturalized in 10.5% of the regions that were predicted to be
643 unsuitable for it. We found that species performing best at filling their expected range have
644 an ornamental or other economic relevance, a higher SLA, and a narrower expected range
645 than species with a less complete range filling (Table 2). Furthermore, species observed as
646 naturalized beyond their expected range tend to have an ornamental or other economic use
647 and to be less widespread in their native range than species with no range expansion (Table
648 2).

649 **Discussion**

650 The global geographic pattern of richness in naturalized plant species [from Europe](#) is non-
651 random as supported by our results, originating from a complex interplay of at least four

652 mechanisms: 1) environmental matching between the native range of individual species and
653 their expected alien range, which largely controls unsuccessful naturalization; 2) the
654 alteration of the environment in the alien ranges by socio-economic activities (after
655 accounting for sampling effort); 3) introduction pressure associated with the economic use of
656 the species; and 4) differences in functional traits (e.g., SLA), which cause some species to
657 more completely fill their expected alien ranges than others.

658 *Accounting for environmental matching to decipher patterns of naturalization*

659 We modelled environmental tolerance of species and then the influence of extrinsic socio-
660 economic drivers (Figure 1), instead of both together as done in previous studies (Essl *et al.*,
661 2015; Dawson *et al.*, 2017). This allowed us to separate the effect of environmental matching
662 in explaining the global pattern of richness in naturalized plant species [from Europe](#). We
663 elucidated that the environment alone is able to correctly predict up to two thirds of
664 successful naturalizations (i.e., sensitivity) and 99.9% of unsuccessful naturalizations (i.e.,
665 negative predictive value), depending on the method used to compare local expected
666 distributions with regional observed distributions ('upscaling method'; Figure 2). Based on a
667 similar approach, Bellard *et al.* (2013) showed that even the so-called "world's 100 worst
668 invasive alien species" have a quite restricted expected range, covering mainly Europe and
669 the areas along the Atlantic coast of North America. Thus, it clearly appears that knowing the
670 available suitable environmental space of species is critical to assess the factors that
671 determine their naturalization success.

672 Our approach indicated that non-European regions are currently occupied [by less than 5%](#) of
673 the endemic European plants for which the area, or parts of it, would be suitable (Figure 2).
674 Therefore, all regions have an enormous naturalization debt. This debt suggests that many
675 endemic European plants have not been introduced outside of Europe yet or that they still
676 have not overcome the biotic barriers provided by herbivores, pathogens and native
677 competitors or the lack of important mutualists. Nevertheless, the large environmentally

678 suitable ranges outside of Europe confirm that European plants have a considerable
679 potential of naturalization outside their native range (van Kleunen *et al.*, 2015; Pyšek *et al.*,
680 2017), once they have overcome dispersal barriers and biotic resistance. Our results
681 therefore indicate that it is unlikely that the number of new naturalizations will soon have
682 reached its peak.

683 SDMs are increasingly used as a basis to implement biosecurity policies, e.g., by drawing up
684 watch lists or lists of undesirable species (e.g., Padayachee *et al.*, 2019). As biosecurity
685 measures are usually taken at a regional scale while SDMs predict expected distributions at
686 smaller scales, it is critical to determine which portion of a region should be suitable for an
687 alien species to become naturalized, and hence, of concern. Our retrospective approach
688 comparing SDM outputs with the current naturalized flora [from Europe](#) in 931 non-European
689 regions showed that a threshold in the order of 1% of a focal region offers a good trade-off
690 between maximizing correctly predicted successful and unsuccessful naturalizations (Figure
691 2). Nevertheless, it may be preferable to choose a lower threshold with the precautionary
692 principle, and this threshold might change when using a different spatial resolution.

693 *Socio-economic drivers explaining differences from expected patterns of naturalization*

694 Our results revealed that naturalization debt mostly decreases with the level of development
695 (Figures 3 and 4), which means that a higher proportion of species with the ability to become
696 naturalized are observed as such in more developed regions. In these regions, the
697 distribution of naturalized species is typically closer to an equilibrium with the environment.
698 This result is in line with Pyšek *et al.* (2010) and Dawson *et al.* (2017), who found that human
699 impact (represented by proxies like wealth and per capita GDP, respectively) play a leading
700 role in driving the global patterns of animal and plant naturalizations. One explanation is that
701 economic wealth is often associated with alien species introductions and with profound
702 alteration of natural habitats, which benefit alien species establishment and growth
703 (MacDougall & Turkington, 2005; Seebens *et al.*, 2015).

704 Surprisingly, unlike other studies (Dawson *et al.*, 2017; Tingley *et al.*, 2018), proxies of
705 international transport (i.e., density of airports and seaports in our study) did not explain
706 naturalization debt (Figures 3 and 4). Nevertheless, we found that regions with the lowest
707 naturalization debt are all either coastal or island regions (coastal North America, Algeria,
708 Japan, temperate Australia and New Zealand; Figure 5). As a result, although naturalization
709 debt seems little influenced by the density of seaports, it is likely that the volume of maritime
710 traffic, or at least the presence of seaports, does play a significant role by increasing the
711 rates of species introductions. One might argue that this pattern also reflects that coastal
712 regions are usually more strongly developed than more inland regions as a consequence of
713 the increasing demand for infrastructures to sustain residential, commercial and tourist
714 activities (Figure S2), which offers more opportunities for introduction and naturalization of
715 alien plants in coastal regions (Gallardo *et al.*, 2015). We found that naturalization debt was
716 higher for the full set of 1,485 European species (either already naturalized or not yet
717 naturalized) than for the subset of 272 species already naturalized, which are more likely to
718 have overcome dispersal barriers (Figure 2). This suggests that introduction pathways
719 influence naturalization debt although, unfortunately, it remains difficult to assess by how
720 much.

721 In addition to the influence of development and potentially of introduction pathways on
722 naturalized range filling, sensitivity was also found to be affected by the magnitude of
723 inventory effort (Figures 3 and 4). Specifically, regions with the highest quality record of
724 native plant richness also tend to have more records of alien plants. This is also corroborated
725 by Dawson *et al.* (2017), and indicates that the geographic coverage of reported plant
726 naturalizations remains unevenly exhaustive across regions so that our findings might be
727 significantly influenced by biases in recording effort. In contrast, the negative predictive value
728 decreased in regions with more treaties relevant to invasive alien species (Figures 3 and 4).
729 This most likely does not indicate that treaties promote invasions, but that it is more likely
730 that a region adopts a treaty when it has many invasive species. The effectiveness of these
731 relatively recent treaties in preventing new invasions can only be assessed in the future.

732 Unexpectedly, human population density and the proportion of cropland had a relatively low
733 contribution in explaining the global pattern of naturalization. Contrary to this, high human
734 population densities are thought to influence the likelihood of an introduced species
735 becoming naturalized and spreading through human impacts on the environment (Essl *et al.*,
736 2011; Pyšek *et al.*, 2010). Similarly, intensive agriculture (associated with historical sowing or
737 planting of now naturalized species and associated weed-seed contaminants) is largely
738 recognized to be a major cause of new introductions of naturalized and invasive species
739 (Bellard *et al.*, 2016; Perrings *et al.*, 2005; Seebens *et al.*, 2015). A potential explanation for
740 the low contributions of population density and the proportion of cropland in driving patterns
741 of naturalization could be its redundancy with the proportion of primary land cover used as
742 environmental variable in SDMs since urbanization and agriculture are largely responsible for
743 land conversion (Chini *et al.*, 2014). Although this redundancy makes it difficult to isolate the
744 effect of human population density and agriculture, we believe the introduction of primary
745 land cover in SDMs was critical in this study to distinguish plants that are drivers and
746 passengers of human-induced disturbances (MacDougall & Turkington, 2005). Another
747 possible cause for the low contribution of agriculture is the ancient human colonization
748 history of Europeans who used to transport cultivated plants across continents while other
749 regions only recently opened up to international movements of plants (di Castri, 1989). In this
750 perspective, most of the plants used in agriculture and associated weeds endemic to Europe
751 with the potential to become naturalized might have already largely spread worldwide
752 (Monnet *et al.*, 2020).

753 *Plant features explaining differences from expected patterns of naturalization*

754 The average score of alien geographic range filling of naturalized plants from Europe was
755 4.2%. This might appear to be very low compared to percentages of environmental niche
756 filling reported in other studies (Hill *et al.*, 2017; Petitpierre *et al.*, 2012; Strubbe *et al.*, 2013,
757 2015; Liu *et al.*, 2020). However, it should be noted that it is easier to fill a niche than to fill all
758 locations that have that niche. None of our species had a range filling above the commonly

759 used threshold of 90%. Similarly, no invasive birds showed more than 90% of their native
760 niche filled in the invasive range (Strubbe *et al.*, 2013). However, no less than 52% of
761 invasive plants (Petitpierre *et al.*, 2012), 32% of 22 insects (Hill *et al.*, 2017) and 10% of
762 vertebrates (Strubbe *et al.*, 2015) had a niche filling above 90%. Moreover, only one
763 naturalized plant species from Europe out of 272 had a range expansion above the
764 commonly used threshold of 10%, while it was the case for the niche expansion of 55% of
765 invasive insects (Hill *et al.*, 2017), 29% of birds (Strubbe *et al.*, 2013), 17% of vertebrates
766 (Strubbe *et al.*, 2015) and 14% of invasive plants (Petitpierre *et al.*, 2012).

767 We found that SLA was positively correlated with range filling of naturalized plants from
768 Europe (Table 2). According to the LHS scheme of Westoby (1998), SLA is a proxy for rapid
769 growth and an acquisitive resource-use strategy, which would be typical of resource-rich
770 and/or disturbed environments, both of which have become more common in the last
771 centuries due to human activities. In the competitor/stress-tolerator/ruderal (CSR) framework
772 of the universal adaptive strategy theory (Grime, 1977), rapid growth is a key characteristic of
773 strong competitors, suggesting that strong competitors for light and nutrients are also filling
774 their naturalized ranges faster. Indeed, Guo *et al.* (2018) recently showed that species with
775 high scores on the competitor axis were more likely to naturalize. Thus, it seems very
776 consistent that having a high SLA represents a key strategy in regions subject to intensive
777 development associated with rapid, deep and widespread anthropogenic perturbations. SLA
778 should therefore also indicate potential for faster expansion towards the equilibrium range in
779 alien regions. This corroborates previous results showing that SLA is higher for naturalized
780 and invasive species compared to non-naturalized or native species (van Kleunen *et al.*,
781 2010; Pyšek & Richardson, 2007; Pyšek *et al.*, 2015).

782 Interestingly, seed mass was not significantly related to range filling (Table 2). This trait can
783 have contrasting effects on patterns of naturalization since large seeds have greater nutrient
784 reserves, which increases the chances to become naturalized, but small seeds are beneficial
785 for long-distance dispersal and therefore favour spread (Crawley *et al.*, 1996; Moodley *et al.*,

786 2013). This suggests that it is not a species' dispersal ability or the ability to settle in new
787 regions that mainly drive naturalization but the ability to be transported by humans for
788 aesthetic or economic values. Our results showed that species used by humans are indeed
789 more frequently naturalized (Figure 6), and perform better at filling (e.g., *Syringa vulgaris*)
790 and expanding (e.g., *Verbascum virgatum*) their potential range than species not used by
791 humans (Table 2). Ornamental horticulture, agriculture and forestry are recognized as major
792 pathways of alien plant introduction. For instance, it has been estimated that at least 75%
793 and 93%, respectively, of the naturalized alien plants worldwide are grown in domestic and
794 botanical gardens (van Kleunen *et al.*, 2018), and plants with a known economic importance
795 are 18 times more likely to be naturalized somewhere in the world (van Kleunen *et al.*, 2020).
796 Possible reasons include: economic importance increases the number of releases and/or the
797 number of individuals released (i.e., propagule pressure), while human selection favours
798 traits related to invasiveness (e.g., ease to grow and reproduce, early and long flowering
799 period, low susceptibility to insect pests or pathogens).

800 Moreover, we found that the same set of distributional features explains whether or not
801 species are naturalized and the extent of naturalized range filling and geographic expansion
802 of species range (range filling and expansion were significantly correlated; $r = 0.61$, P -value
803 < 0.001), but in opposite directions. Indeed, naturalized species tend to be more widespread
804 in their native range and to have wider expected alien ranges than non-naturalized species,
805 i.e., they are typically generalist species (Gallien *et al.*, 2019). In contrast, naturalized
806 species with the most complete range filling and broadest range extension tend to have more
807 limited expected alien ranges and to be less widespread in their native range, respectively.

808 This result needs to be interpreted with caution since it can simply arise from a numerical
809 artefact as it may be easier to fill a small expected range than a large one. Another possible
810 cause could be that SDMs tend to be more accurate for specialist species than for
811 generalists (Connor *et al.*, 2018). Furthermore, range filling depends on opportunities to
812 disperse, so species with a wider expected alien range may basically have less chance to fill

813 it. Finally, more narrowly distributed endemics can surprisingly have a greater range
814 expansion if their observed distribution in the native range results from biogeographical (e.g.,
815 the Mediterranean Basin, the Alps in Europe) or from biological interactions rather than from
816 climatic barriers. In that case SDMs would fail to fit their fundamental climatic envelope due
817 to niche truncation (Bush *et al.*, 2018). It could also be that their current distribution is
818 matching their climatic niche, but that this distribution is not correctly reflected in data
819 sources, maybe because they mostly inhabit under-sampled regions. Consequently, the
820 projected expected alien ranges may underestimate the true potential ranges.

821 **Conclusions**

822 Our findings indicate that the environment largely controls the expected naturalized range of
823 European plants. We demonstrated that increased anthropogenic disturbance associated
824 with human development as well as increased probability of introductions associated with
825 international exchange primarily explain the extent to which species spread within their
826 expected naturalized range. We showed that plants selected for ornamental and other
827 economic purposes perform better at filling and expanding their range, most likely because of
828 increased introduction pressure and a preference for cultivating plants with traits that also
829 make them more likely to naturalize. We revealed that species with functional traits indicative
830 of rapid growth and acquisitive resource use tend to fill their range more completely than
831 species with slower responses. These findings give a new overall picture of the drivers of
832 naturalization that can help plan future studies on the macroecology of alien species as well
833 as in designing future biosecurity plans. [As harmful invasive species emerge from the pool of
834 naturalized species, the analysis of which species could naturalize is a first step in identifying
835 species that could ultimately become invasive in a region.](#)

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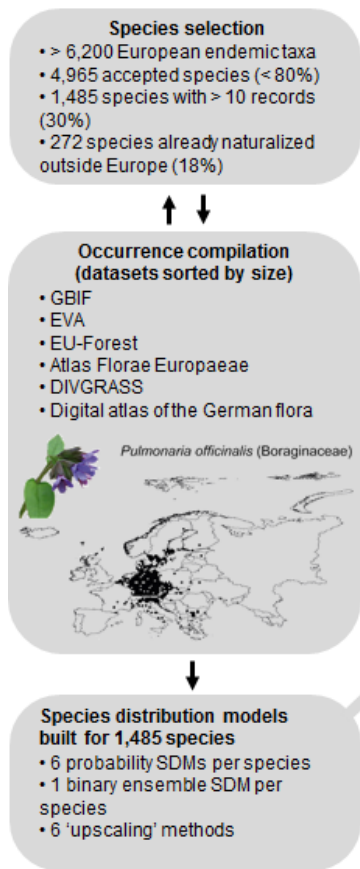
Table 1 Summary of the seven socio-economic factors used to assess the drivers of naturalization success of the European vascular plants including the year and spatial resolution at which the original data was collected/calculated, the range of values for the regions used in analyses, and which transformation was applied to these values prior to analyses. Minimum and maximum values are those, after transformation, over the 526 non-European regions for which all socio-economic factors were available.

Variable	Year	Resolution	Unit	Min	Max	Transformation
<i>Introduction pathways</i>						
Airports	2019	-	100,000 km ²	0	11	log
Seaports	2014	-	100,000 km ²	0	15	log
Treaties	2016	-	#	6	30	-
<i>Human pressures</i>						
Population	2000	1 km	Inhabitant.km ⁻²	0	10	log
Cropland	< 2007	5 arcmin	%	0	95	-
Development	2015	5 arcmin	Dimensionless	.27	.93	-
<i>Sampling effort</i>						
Inventory	2015-16	110 km	%	0	249	-

Table 2 Regression coefficients of the relationship between distributional, economic and functional features of naturalized plants from Europe and their range filling (proportion of suitable regions where species are observed as naturalized) and range expansion (proportion of unsuitable regions where species are observed as naturalized). Continuous variables were standardized to allow comparison. Significance levels of the relationships fitted with linear mixed-effect models accounting for between-species phylogenetic distances are indicated by asterisks ($.05 < P\text{-value} \leq .01$: *, $.01 < P\text{-value} \leq .001$: **, $P\text{-value} < .001$: ***).

	Range filling	Range expansion	<i>n</i>
Prevalence in Europe	0.13	-0.40*	272
Potential range outside Europe	-0.25**	0.08	272
Ornamental	0.22***	0.14*	272
Other use	0.21***	0.22***	272
Growth form	0.02	0.01	193
Specific leaf area (SLA)	0.27*	-0.01	117
Height	0.05	0.01	175
Seed mass	0.02	-0.01	156

Species distribution models – Identifying non-European regions with matching environments



A posteriori analysis – Assessing the role of climate, species traits and socio-economic factors in explaining differences between observed and expected degrees of naturalization

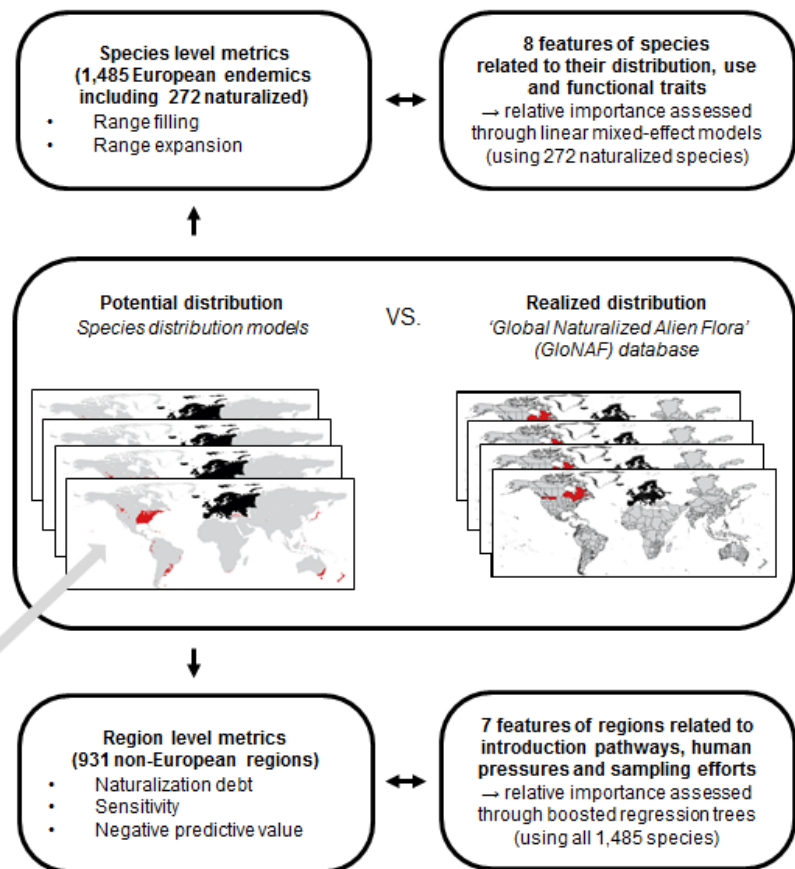


Figure 1 Workflow of this study used to determine which features of both plants and recipient regions explain possible discrepancies between observed and expected extents of naturalization.

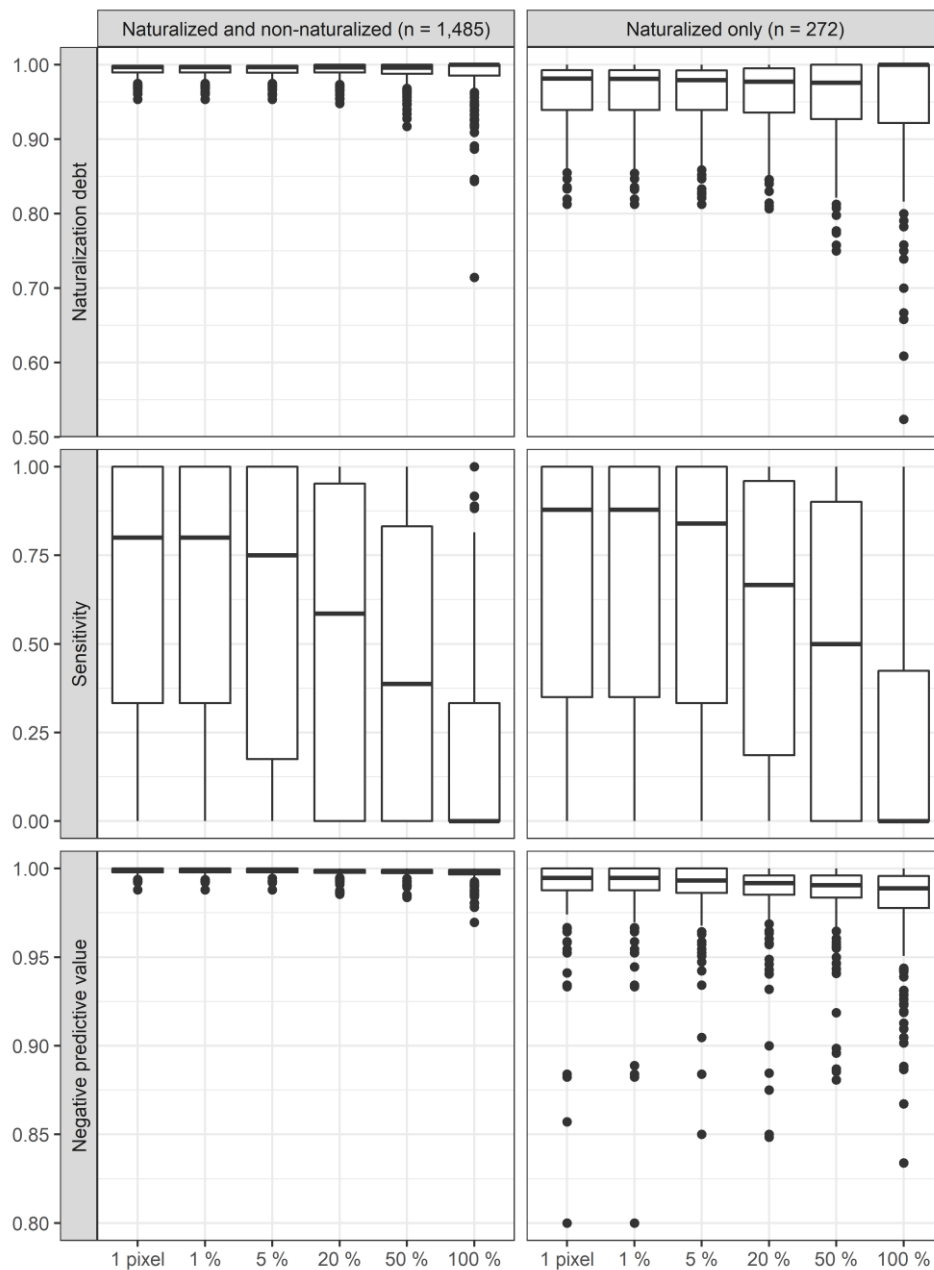


Figure 2 Comparison of predicted and observed naturalized floras in 931 non-European regions based on various upscaling methods used to transform local naturalization predictions into regional naturalization predictions. The left column comprises our full set of European plants (either already naturalized or not yet naturalized outside of Europe) and the right column comprises only the subset of those plants already naturalized outside of Europe. Naturalization debt refers to the proportion of predicted naturalizations that are not yet observed, sensitivity to the proportion of correctly predicted successful naturalizations and

negative predictive value to the proportion of correctly predicted unsuccessful naturalizations.

A region was considered suitable if it contained at least one suitable pixel, 1% suitable cells, 5%, 20%, 50%, and 100%.

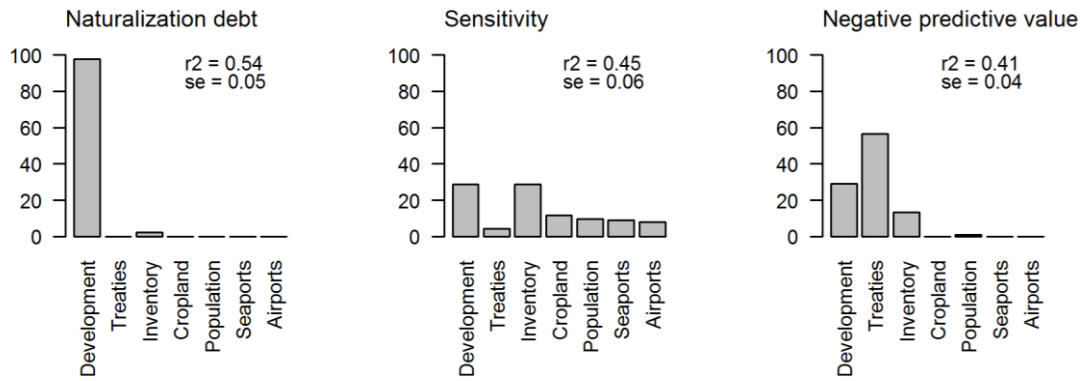


Figure 3 Relative influence of seven socio-economic factors in explaining differences between observed and expected European naturalized floras calculated by boosted regression trees. Naturalization debt refers to the proportion of predicted naturalizations that are not yet observed, sensitivity to the proportion of correctly predicted successful naturalizations and negative predictive value to the proportion of correctly predicted unsuccessful naturalizations. Socio-economic factors are sorted by mean relative influence over the three metrics. These results are for the upscaling method ‘1 %’ i.e., a region was considered suitable for a species if it contained at least 1% suitable cells.

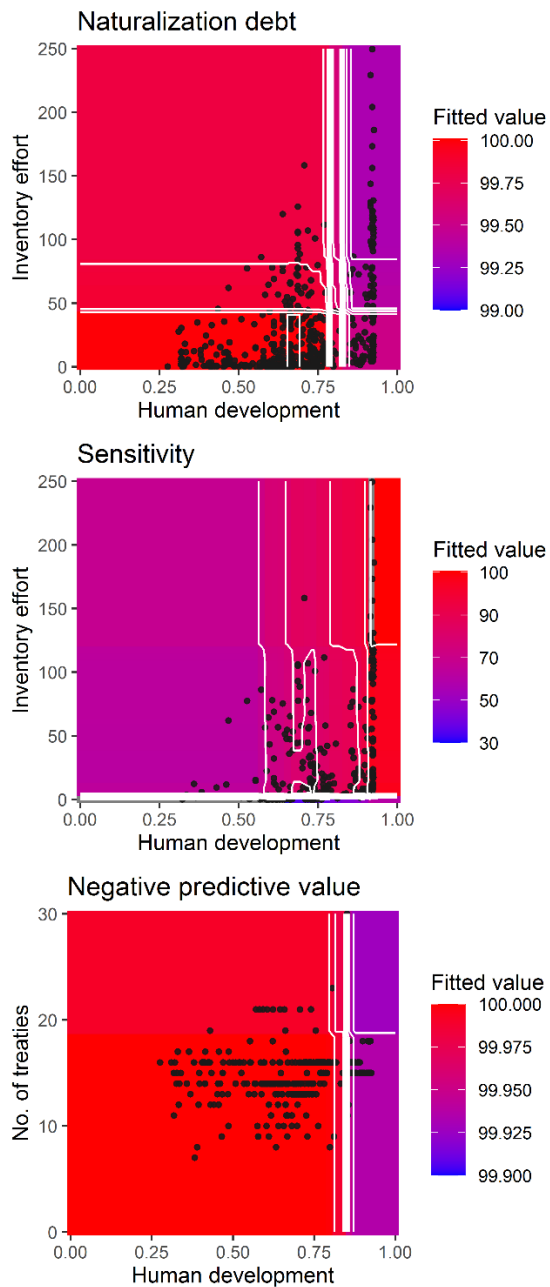


Figure 4 Joint partial dependence plots of interactions fitted by boosted regression trees between the most contributing regional socio-economic factors and differences between observed and expected European naturalized floras. Naturalization debt refers to the proportion of predicted naturalizations that are not yet observed, sensitivity to the proportion of correctly predicted successful naturalizations and negative predictive value to the proportion of correctly predicted unsuccessful naturalizations. Black dots represent empirical data points. These results are for the upscaling method '1 %' i.e., a region was considered suitable for a species if it contained at least 1% suitable cells.

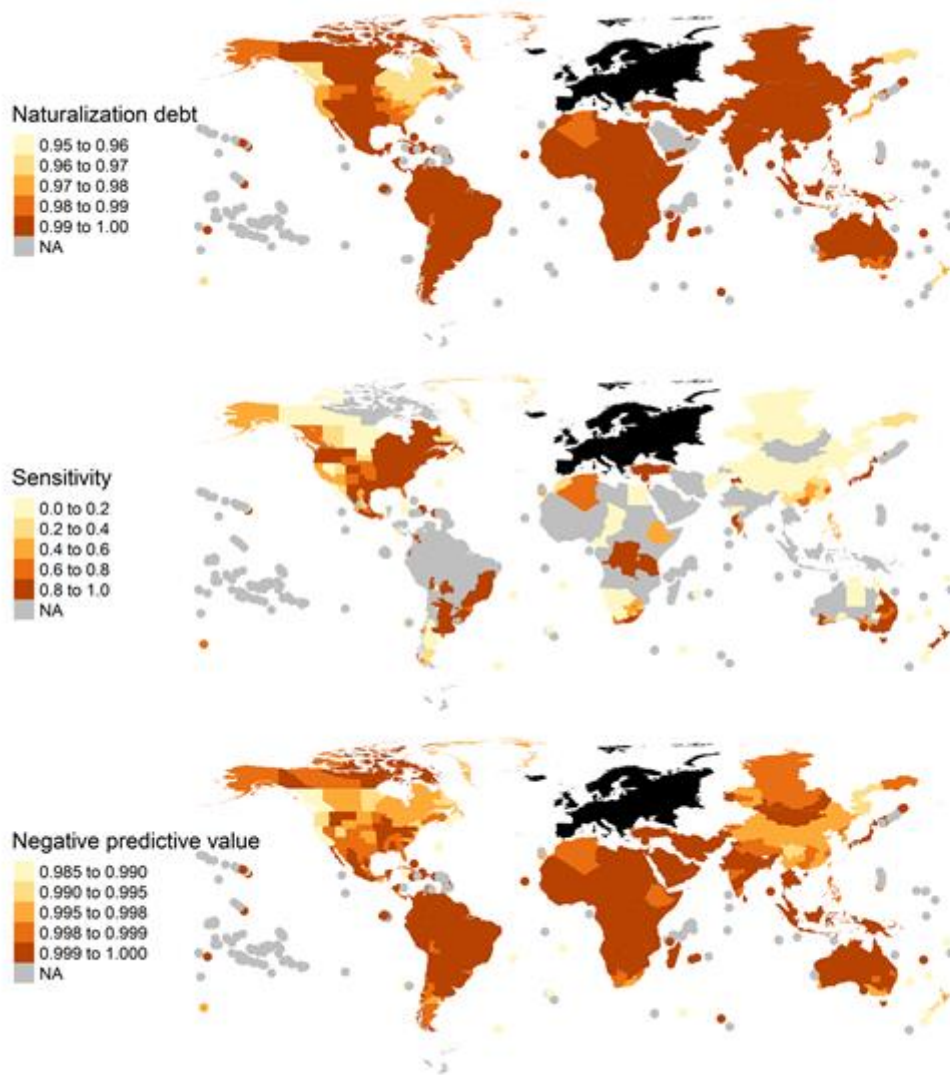


Figure 5 Global maps of the differences between observed and expected European naturalized floras. Regions with no GloNAF data are displayed in white. Naturalization debt refers to the proportion of predicted naturalizations that are not yet observed, sensitivity to the proportion of correctly predicted successful naturalizations (there are many NA values as many regions have no observed naturalizations by European endemic plants) and negative predictive value to the proportion of correctly predicted unsuccessful naturalizations. These results are for the upscaling method '1 %' i.e., a region was considered suitable for a species if it contained at least 1% suitable cells.

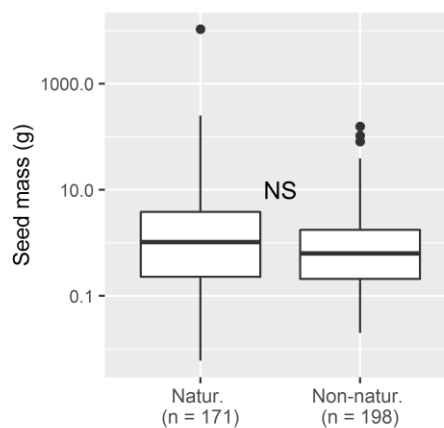
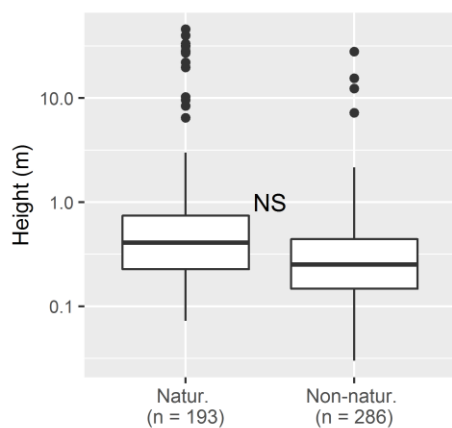
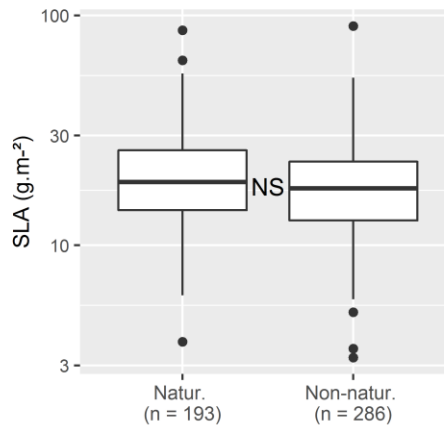
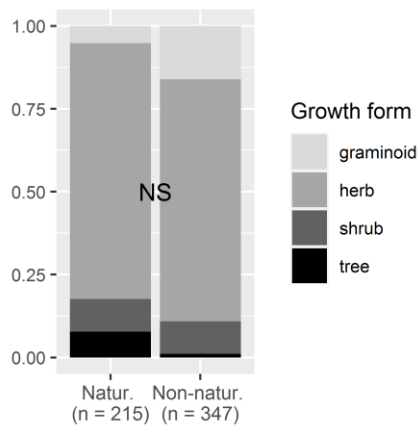
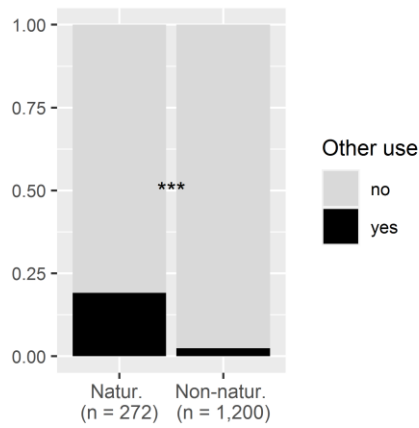
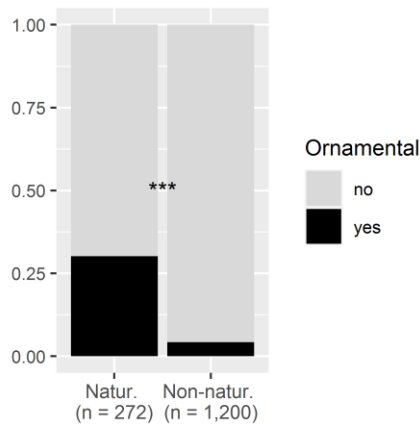
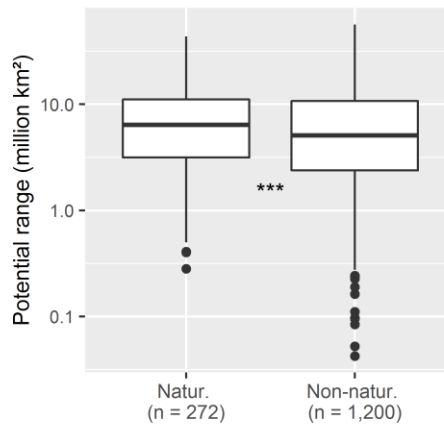
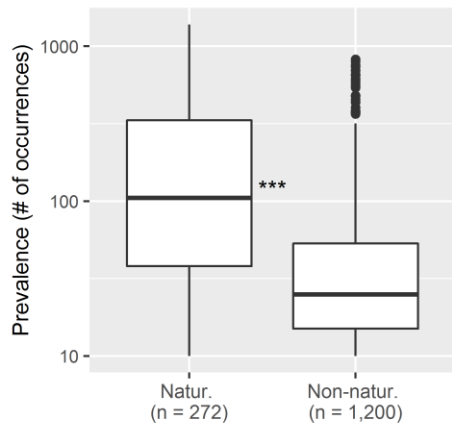


Figure 6 Difference in distributional, economic and functional features between European vascular plants already recorded as naturalized outside Europe ('Natur.') and those not recorded as naturalized outside Europe ('Non-natur.'). 1,472 species in total. 'Prevalence' refers to the number of $0.42^\circ \times 0.42^\circ$ cells in Europe occupied, 'Potential range' to the surface area of the species' expected range outside Europe, 'Ornamental' and 'Other use' indicate whether species are used as an ornamental plant or has other economic relevance (e.g., food plant, medicines, materials), respectively, 'Growth form' reflects the morphology of a plant, especially its physiological adaptation to the environment, 'SLA' for 'specific leaf area' is the ratio of leaf area to leaf dry mass, 'Height' and 'Seed mass' are the plant height and seed mass. Significance levels of the relationships fitted with linear mixed-effect models accounting for between-species phylogenetic distances are indicated by asterisks ($.05 < P\text{-value} \leq .01$: *; $.01 < P\text{-value} \leq .001$: **, $P\text{-value} < .001$: ***).