

Conservation applications of ecological niche modeling: non-native ferns may compete with native ferns for limited suitable habitat in Hawaiian dryland ecosystems

Krystalyn Edwards-Calma¹, Laura Jiménez^{1,2}, Rosana Zenil-Ferguson³, Karolina Heyduk^{1,4}, Miles K. Thomas⁵, and Carrie M. Tribble^{1*}

¹*School of Life Sciences, University of Hawai'i at Mānoa, Honolulu, HI, 96822, USA*

²*Centro de Modelamiento Matemático, Universidad de Chile, Santiago, Chile*

³*Department of Biology, University of Kentucky, Lexington KY 40506 USA*

⁴*Department of Ecology and Evolutionary Biology, University of Connecticut, Storrs CT 06269*

⁵*Herbarium Pacificum, Bernice Pauahi Bishop Museum, Honolulu, HI 96813*

*Corresponding author: ctribble09@gmail.com

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Abstract

Premise

Competition from non-native species and habitat loss are common threats to biodiversity and may act synergistically to increase competition for decreasing habitat availability. We develop a workflow using Hawaiian dryland ferns as a model for the interactions between land-use change and non-native competition in determining available habitat for native species.

Methods

We use fine-resolution climatic variables and carefully curated occurrence data from herbaria and community science repositories to estimate the distributions of Hawaiian dryland. We quantify the degree to which non-natives occupy similar regions of niche space and map the remaining available habitat given land-use change in Hawai'i.

Results

The estimated niches for the endemic species (*Doryopteris* spp.) are more narrow than for the indigenous *Pellaea ternifolia*. *Doryopteris angelica* has the lowest proportion of non-native occurrences in its estimated suitable area while *D. decora* has the highest. Both *D. decora* and *D. decipiens* have low proportions (<20%) of inferred suitable area covering native habitat.

Discussion

Areas characterized by shared environmental preferences of native and non-native ferns may also decrease due to human development and fallowed agricultural lands. Our study demonstrates the value of a novel ecological niche modeling approach for conservation risk assessment in a rapidly changing and urbanized island ecosystem.

Keywords: *climatic niche; conservation; ecological niche modeling; Hawaiian ferns, land use; Pteridaceae*

INTRODUCTION

1 One of the primary goals of conservation biology is to assess the amount and location of suit-
2 able habitat for species of concern. Ecological niche models (ENMs) use occurrence data (records
3 of where individuals of the species have been observed in the past) to estimate the climatically
4 suitable habitat of a particular species (Soberón, 2010). ENMs may allow conservation practi-
5 tioners to determine the extent of suitable habitat for threatened species, prioritize areas for land
6 conservation, identify regions where a rare species may exist but still be undetected, and many
7 more possible applications (Zhu et al., 2013). However, most traditional approaches to ecolog-
8 ical niche modeling confuse theoretical and practical differences between the fundamental and
9 realized niche, confounding the interpretation of such analyses and thus the proper application
10 of ENMs in conservation (Soberón, 2010; Zhu et al., 2013). The fundamental niche is defined as
11 the set of environmental conditions that a species could inhabit in the absence of biotic factors
12 (such as competition) and dispersal limitations, while the realized niche is the set of conditions
13 that a species currently inhabits as a result of the dynamics of biotic interactions and historical
14 dispersal patterns (Hutchinson, 1957; Pulliam, 2000; Soberón and Arroyo-Peña, 2017). For the vast
15 majority of niche modeling studies, occurrence data from biological collections describe where a
16 particular species has been observed and represent a sample from the realized niche, which may
17 under-represent the fundamental niche of a species (Jiménez et al., 2019). The dearth of appro-
18 priate statistical methods to estimate the fundamental niche from presence-only data had limited
19 the ability of conservationists to expand their efforts to preserve areas where vulnerable species
20 can live—or historically lived—because the right climatic conditions exist, even if the vulnerable
21 taxa currently do not occur in a given location due to biotic interactions. Therefore, it is essential
22 to infer fundamental niches for conservation purposes. New developments in ecological niche
23 modelling can allow conservation practitioners to preserve critical areas where species can sur-
24 vive. These new ENMs modeling approaches are improving our ability to infer the fundamental
25 niche by adding a hypothesis of the accessible area that the species is able to explore (Jiménez and
26 Soberón, 2022).

27 Additionally, factors beyond the abiotic variables used in ENMs may significantly affect the
28 available area for species of conservation concern. For example, land-use change may mean that

29 areas characterized by the appropriate climatic conditions for a species to thrive are instead un-
30 available via land conversion from native habitat to areas of intense human development, agricul-
31 ture, or invasive-dominant ecosystems. Additionally, the fundamental niche intentionally models
32 the range of an organism without considering the impact of biotic interactions such as competi-
33 tion. Thus, competition from non-native species with similar habitat preferences may also signifi-
34 cantly affect the amount of available area for native species of conservation concern. These factors
35 are typically not accounted for in traditional ENMs, so conservation applications of ENMs may
36 overestimate the available suitable area of species of interest. In this study, we describe a pipeline
37 for modeling the fundamental niche using the statistical approach in Jiménez and Soberón (2022),
38 while accounting for the impacts land use change and competition from non-native species on
39 available suitable habitat. We apply this methodology to the vulnerable native Hawaiian dryland
40 ferns in the family Pteridaceae and discuss the potential future applications of such approaches
41 for conservation.

42 Hawai'i is widely known as a biodiversity hotspot, with textbook examples of adaptive radi-
43 ations following long-distance dispersal to the isolated volcanic archipelago (e.g., the lobelioids,
44 Givnish et al. 2009, silverswords, Baldwin and Sanderson 1998; Witter and Carr 1988, and song-
45 birds, Lovette et al. 2002). Despite the small area of available land, Hawai'i contains an extraor-
46 dinary diversity of biomes and environmental conditions, which has likely allowed organisms to
47 adapt to novel environments and diversify (Barton et al., 2021). Moreover, because of Hawai'i's
48 isolation and the rarity of long distance dispersal events, rates of endemism are particularly high:
49 90% of the approximately 1300 native Hawaiian vascular plant species are found nowhere else
50 (Rønsted et al., 2022). Sadly, Hawai'i's spectacular biodiversity is also under immense threat. A
51 recent report on the conservation status of approximately 1000 native Hawaiian vascular plants
52 showed that 75% of the flora is threatened or endangered (Rønsted et al., 2022), while only 33% of
53 that flora receives protection under the U.S. Endangered Species Act and the Hawai'i State Gov-
54 ernment, implying that much of the native flora is under threat yet unprotected.

55 Tropical dry forests are one of the most threatened biomes worldwide (Cordell and Sandquist,
56 2008; Quesada et al., 2009; Wilson et al., 1988). They once made up 42% of the world's tropical
57 regions but currently only 2% remain (Buzzard et al., 2016). Across the tropics, the dry forest
58 biome has attractive features for human use: they tend to be suitable areas for livestock and agri-

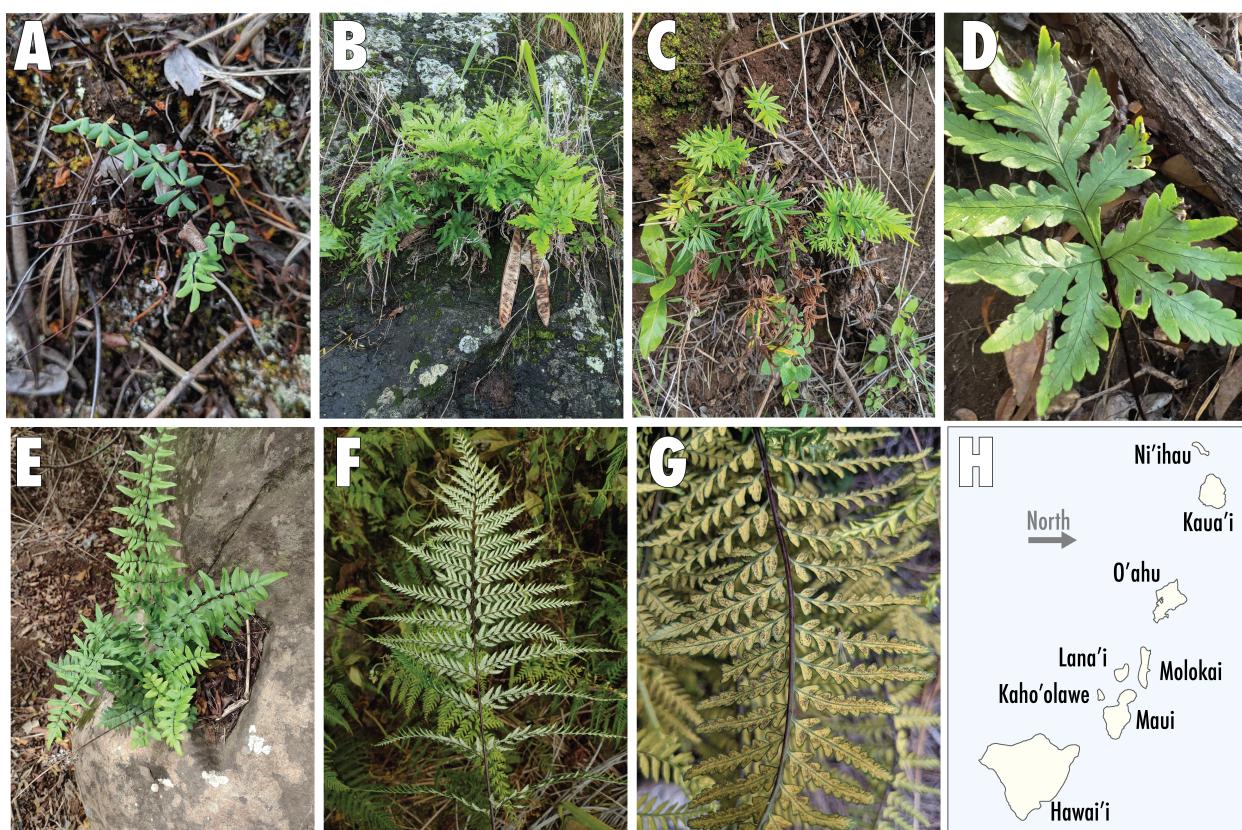


Figure 1: Native (top) and non-native (bottom) dryland ferns in Hawai'i. (A) *Pellaea ternifolia*, indigenous to Hawai'i (Hawai'i island, Maui, and Kaua'i) and the Americas, (B) *Doryopteris decipiens*, endemic to Hawai'i (all major islands), (C) *Doryopteris decora*, endemic to Hawai'i (all major islands), (D) *Doryopteris angelica*, endemic to Hawai'i (Kaua'i single-island endemic), (E) *Cheilanthes viridis*, non-native, (F) *Pityrogramma calomelanos*, non-native, (G) *Pityrogramma austroamericana*, non-native, and (H), map of the major Hawaiian islands. Pictures A, B, C, G by C. M. Tribble, picture D by Susan Fawcett, and pictures E and F by Kevin Faccenda.

59 culture because their soils are often fertile and they possess a marked rainfall season, and they
60 are favored for human development because the dry environment may reduce the risk of dis-
61 ease transmission (Murphy and Lugo, 1986; Portillo-Quintero et al., 2015). These global patterns
62 hold true in Hawai'i; most original tropical dry forests in Hawai'i have been destroyed (Cordell
63 and Sandquist, 2008) and they are continuously exposed to various threats, mainly due to human
64 activity (Miles et al., 2006). In Hawai'i, the leading causes for deforestation of drylands include
65 conversion of former dry forests to agricultural- or pastureland, human development, invasion by
66 nonnative species, and wildfires (D'Antonio and Vitousek, 1992).

67 While ferns are typically found in moist ecosystems, some ferns have unique adaptations that
68 allow them to thrive in drylands (Sharpe, 2019). In particular, the Cheilanthoideae subfamily of
69 Pteridaceae consists of primarily dry-adapted species worldwide (Gaston and Rollo, 1995; Hevly,

70 1963). In Hawai'i, this subfamily is represented by five native species (*Doryopteris* J.Sm. and *Pel-*
71 *laea ternifolia* (Cav.) Link). Taxonomic classification of *Doryopteris* (like much of Cheilanthoideae)
72 has been a challenge for many years because of the lack of clear morphological characters to iden-
73 tify species (see panels B-D in Fig. 1), causing taxonomic uncertainty within the genus (Bouma
74 et al., 2010; Gastony and Rollo, 1995; Ranker et al., 2019a,b; Rothfels et al., 2008; Schuettpelz et al.,
75 2007; Tryon, 1942, 1944; University Herbarium, 1939; Vernon and Ranker, 2013; Windham et al.,
76 2009; Yesilyurt, 2003; Yesilyurt and Schneider, 2010; Yesilyurt et al., 2015; Zhang et al., 2008, 2007).
77 Despite this uncertainty, to date there are four recognized species of *Doryopteris* that are endemic
78 to the Hawaiian islands: *Doryopteris decora* Brack., *Doryopteris decipiens* (Hook.) J.Sm., *Doryopteris*
79 *angelica* K.Wood & W.H.Wagner, and *Doryopteris takeuchii* (W.H.Wagner) W.H.Wagner (University
80 Herbarium, 1939; Vernon and Ranker, 2013; Yesilyurt, 2005; Yesilyurt et al., 2015), along with one
81 named hybrid *Doryopteris subdecipiens* W.H.Wagner (Palmer, 2003). Members of this genus reside
82 in warm and dry shrubland, grassland, and forest habitats (Palmer, 2003), and are known to favor
83 rocky substrates within those habitats (Tryon, 1942). *Pellaea ternifolia* is another native Hawai-
84 ian Cheilanthoid species (see panel A in Fig. 1) that can also be found in dryland environments
85 (Gastony and Rollo, 1995; Valier, 1995). Unlike the Hawaiian *Doryopteris*, *P. ternifolia* is not en-
86 demic to Hawai'i; it also occurs in Central and South America. In Hawai'i, *P. ternifolia* tends to
87 occur in high elevation montane environments such as Mauna Kea, Mauna Loa, and Kilaeua on
88 Hawai'i island and Haleakalā on Maui.

89 Almost 200 native fern species inhabit the Hawaiian Islands, of which 77% are endemic Palmer
90 (2003). Ranges of ferns are considered to be strongly determined by habitat availability (Smith,
91 1993; Tryon, 1986); moreover, endemic ferns are generally found at high elevations (Kessler et al.,
92 2010; Kluge and Kessler, 2006, though Hawaiian lowland endemics *Doryopteris* spp. are an excep-
93 tion to this trend). These features make native fern species particularly susceptible to habitat loss,
94 human-induced habitat disturbance, and competition for space against alien species. For example,
95 non-native relatives to the Hawaiian members of Cheilanthoideae may be competing for the same
96 dryland habitat that the native species prefer. *Pityrogramma calomelanos* (L.) Link—which is native
97 to North and South America and the Caribbean (POWO, 2023)—and *P. austroamericana* Domin—
98 which is native to Central and South America (POWO, 2023)—both arrived to Hawai'i and are
99 now widespread across the islands (see panels F and G in Fig. 1). Both *Pityrogramma* species are

100 also part of the Pteridaceae family and share morphological characteristics typical of the Cheilan-
101 thoideae subfamily (such as farina, a waxy substance on the abaxial surface of the fronds thought
102 to protect the fronds from damage when desiccated, Kao et al. 2019), although they belong to the
103 subfamily Pteridoideae (Zhang et al., 2017). *Cheilanthes viridis* (Forssk.) Sw. (Cheilanthoideae,
104 Pteridaceae) is another non-native dryland fern found in Hawai'i (see panel E in Fig. 1) whose
105 native distribution includes most of eastern and southern Africa and the Middle East (POWO,
106 2023).

107 Land use has drastically changed over the years in Hawai'i. In 1848, The Great Māhele (great
108 land division) was one of the first changes in land usage (Chinen, 2020) and resulted in the parti-
109 tioning of land in the Hawaiian Islands, changing the feudal system to permit land privatization
110 (Kamakau, 1961; Linnekin, 1983). Following the Great Māhele, Hawai'i became attractive to for-
111 eign investors who sought to claim lands in an effort to further develop agricultural plantations.
112 In 1835, the first sugarcane plantation was established in Koloa, Kaua'i (Association, 1992) and,
113 by 1890, foreigners and foreign corporations owned three out of four acres of private lands in
114 Hawai'i (Takaki, 1984). Problems of managing plantations arose soon after the Great Māhele.
115 Sugar, pineapple, and macadamia plantations—often established in former tropical dry forest
116 environments—caused shortages of water in West Maui (Association, 1992). Since the early 1900s,
117 agricultural corporations and landholders have diverted natural water resources to plantations in
118 former native dryland habitat, resulting in over 90 billion gallons of water being diverted from
119 natural island streams (John A. Engott, 2007). In more recent years, due to changes in the econ-
120 omy, plantations have largely been abandoned (Zou and Bashkin, 1998), and former plantation
121 land is now primarily dominated by invasive grasses such as guinea grass (*Megathyrsus maximus*,
122 Farrant et al., 2023). These land-use changes are a significant conservation concern because alien
123 species often invade native communities after some type of disturbance (Smith, 1985). In partic-
124 ular, abandoned drylands create potential hazards to native dryland ferns as non-native species
125 invade, changing the biotic and abiotic environment (non-native grasses may suppress native
126 species success) and potentially enhancing competition for limited available suitable habitat (non-
127 native closely related ferns may share similar habitat preferences D'Antonio and Vitousek, 1992).

128 To better understand the possible adverse effects of (i) the presence alien species in the focal
129 species ranges and (ii) land-use changes on the survival of native fern species that have become

130 adapted to particular environmental niches in the Hawaiian islands, we developed an integrated
131 ENM pipeline. We first carefully curated occurrence data for seven fern species from herbaria
132 and open-access community science databases and integrated these presence-only data with fine-
133 resolution climatic variables of the main Hawaiian Islands to estimate the fundamental niches of
134 the native ferns *Doryopteris angelica*, *Doryopteris decipiens*, *Doryopteris decora*, and *Pellaea ternifolia*.

135 We used the novel ENM approach proposed by Jiménez and Soberón (2022). For every recorded
136 location of a given species, there are a series of corresponding climatic values (i.e., annual average
137 temperature). We can plot the occurrences of species in multidimensional environmental space.
138 In this environmental space, the Jiménez and Soberón (2022) model leverages random sampling
139 along with the climatic variables to infer a multivariate normal ellipsoid that represents the prob-
140 ability of a species tolerating combinations of environmental variables. With the multivariate
141 normal probability at hand, we can trace back to the the latitude and longitude in geographical
142 space, but now those localities in the map are associated with a probability of environmental tol-
143 erance, and when plotted back in geographical space they create suitability maps for our species.
144 In summary, this methodology takes the geographical occurrence data to climatic space, it calcu-
145 late probabilities of climatic tolerance, and uses those probabilities to return to geography, where
146 we have now information about the chances for suitable space. Following similar protocols, we
147 curated data of three non-native fern species, *Cheilanthes viridis*, *Pityrogramma calomelanos*, and *P.*
148 *austroamericana*, and quantified the degree to which non-natives occupy the potential ranges of
149 the native ferns. Finally, using a land-use map of the Hawaiian islands, we identified regions with
150 different degrees of disturbance that hold suitable climatic conditions for the native fern species
151 to evaluate the impact of habitat change.

152 METHODS

153 Species occurrence data

154 We obtained occurrence records of the native and non-native species (Fig. 2 step (1)) from digitized
155 museum collections and human observations via the Pteridophyte Collections Consortium (PCC
156 Consortium, 2022), the Global Biodiversity Information Facility (GBIF GBIF, 2022), the iNatural-
157 ist website<https://www.naturalist.org/>), the Herbarium Pacificum (BISH) at Bishop Museum,

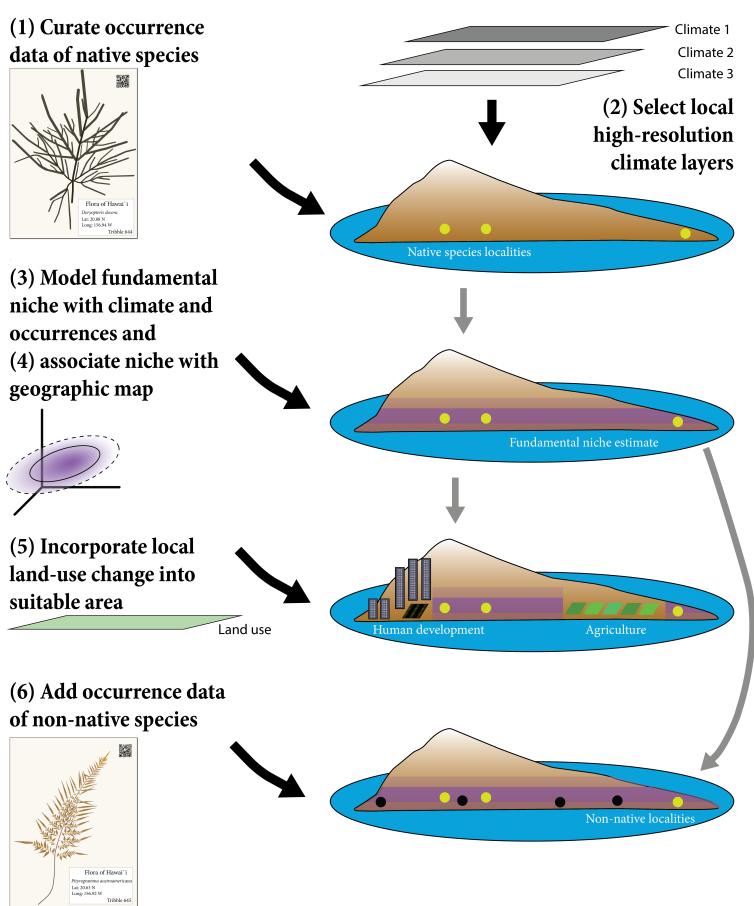


Figure 2: Schematic representation of the analytical workflow. (1) Curate occurrence data of data species. Data from verified physical vouchers is preferred, but iNaturalist or other participatory science databases can supplement if necessary. Curation should include both expert verification of species identifications and careful data cleaning to remove untrustworthy or error-prone GPS data. (2) Gather and transform high-resolution climatic variables, ensuring the variables are not highly correlated and appear sufficient to describe important aspects of the species ecology. (3) Use the data from steps (1) and (2) to estimate the fundamental niche of the native species and (4) identify their potential distributions. The suitability maps created in step (3) were used to: (5) quantify how much suitable area is no longer available to native ferns due to land-use and (6) calculate the percentage of non-native fern occurrences inside the suitable areas.

158 and the Joseph F. Rock Herbarium (HAW) at the University of Hawai'i at Mānoa. We were unable
159 to recover sufficient accurate occurrences to include *Doryopteris takeuchii* or *Doryopteris subdecipliens*
160 in the study, so all downstream analyses focused on the remaining three *Doryopteris* and *P. ternifo-*
161 *lia*. Coordinates for *D. angelica* were obscured on iNaturalist because it is a protected species, so we
162 obtained coordinates for *D. angelica* by reaching out to individual observers. In efforts to obtain
163 more occurrences for *D. angelica*, *D. decora*, *P. calomelanos*, and *P. austroamericana*, we received occur-
164 rences from the herbaria BISH and HAW. HAW provided access to digitized collections through
165 the database of the Consortium of Pacific Herbaria, and we requested occurrence records from

166 BISH.

167 We combined all occurrence data and removed duplicate records. We cleaned the combined
168 dataset to reduce inaccuracies in species occurrences. We excluded records with large coordi-
169 nate uncertainty as they lacked precision. We prioritized recent occurrences—after 1950—because
170 older records are more susceptible to GPS error and there could be a mismatch between the mod-
171 ern climate data (described below) and the older observations. In one case—*D. decora*—we in-
172 cluded older observations due to low sample size. Additionally, we included some localities with
173 unknown years to increase our sample size. We determined the accuracy of the localities based
174 on expected species ranges. We plotted the occurrences over the Hawaiian Islands. We discarded
175 points located in the ocean, outside of Hawai'i, or markedly outside of the species expected range.
176 Lastly, we verified species' identifications by consulting taxonomic experts, including the authors
177 of this study.

178 **Climatic data**

179 To model the fundamental niche of the four focal native species in our study, we selected cli-
180 matic variables that describe factors we believe are likely to influence species ranges (described
181 in Fig. 2 step (2)). We gathered seven environmental variables from the Climate of Hawai'i web-
182 site (Giambelluca et al., 2014), which includes detailed breakdowns of climatic and environmen-
183 tal variables: diffuse radiation (W/m^2), leaf area index, relative humidity (%), soil evaporation
184 (mm), solar radiation (W/m^2), available soil moisture, actual evapotranspiration (mm). However,
185 the most up-to-date data on Hawaiian temperature and precipitation is available at the Hawai'i
186 Climate Data Portal (<https://www.hawaii.edu/climate-data-portal/>). We obtained monthly
187 temperature minimum and maximum values and monthly precipitation values from the Data Port-
188 tal and transformed those data into five additional variables: mean minimum temperature of the
189 coldest four months (cold mean; $^{\circ}\text{C}$), mean maximum temperature of the hottest four months (hot
190 mean; $^{\circ}\text{C}$), mean precipitation of the wettest three months (wet mean; mm), mean precipitation of
191 the driest three months (dry mean; mm), and precipitation seasonality (standard deviation of the
192 monthly precipitation values).

193 We ensured all climatic variables were the same resolution (250m) and projection (WGS84).
194 The raster layers cover all major Hawaiian Islands except for Ni'iha, which was excluded from

195 our analyses. As many of the 12 variables may be highly correlated, we estimated the Pearson cor-
196 relation coefficient (r) for all pairwise comparisons and retained six variables such that all pairwise
197 r values satisfied $-0.7 < r < 0.7$.

198 **Climatic niche modeling of native species**

199 We applied a methodology that enables the identification of the potential distribution of a species
200 (i.e., the set of sites with suitable climatic conditions where the species can inhabit) by modeling
201 its fundamental Grinellian niche (*sensu* Soberón, 2007) in a multivariate space defined by climatic
202 variables relevant for the survival of the species. In particular, we used the model proposed by
203 Jiménez and Soberón (2022) to estimate the fundamental niches of *Doryopters decipiens*, *D. decora*,
204 *D. angelica*, and *Pellaea ternifolia* with occurrence data and information regarding the dispersal lim-
205 itations and climate availability within the area of study. For each native fern species, we used
206 the curated sample of occurrences (see step 1 in Fig. 2) and extracted the climatic values from the
207 raster layers of the six variables described before. We worked under the assumption that the dis-
208 persal abilities of these species allow them to reach any of the main islands within the Hawaiian
209 archipelago, while they cannot disperse to other land areas by natural means. In other words, the
210 main Hawaiian islands (Kaua'i, O'ahu, Moloka'i, Lāna'i, Kaho'olawe, Maui, and Hawai'i) repre-
211 sent our hypothesis about the dispersal abilities of the species (known as the 'M' hypothesis; see
212 Soberón, 2010). We obtained a random sample of eight thousand points within the main Hawaiian
213 islands, extracted their climatic conditions, and use them to estimate the parameters of interest.

214 Jiménez and Soberón (2022) modeling approach assumes that the fundamental niche of the
215 species has an ellipsoidal shape given by a multivariate normal distribution in environmental
216 space which is described through two parameters of interest: a vector that represents both the
217 optimal environmental conditions for the species and the center of the ellipsoid, and a variance-
218 covariance matrix that defines the shape and size of the niche. We supplied the likelihood func-
219 tion of the model with the occurrence data and the set points representing the available climatic
220 conditions within the study area, and estimated the parameters of interest through a maximum
221 likelihood approach (step 3 in Fig. 2). The Jiménez and Soberón (2022) estimates ellipsoids in
222 multidimensional space where the number of dimensions is equal to the number of climatic lay-
223 ers provided to the model. Estimating the parameters of a model with more than three layers

224 requires an advanced optimization algorithm that is currently not implemented. To address this
225 computational limitation, we created two different models using two sets of three climatic vari-
226 ables each, called Model 1 and Model 2. Model 1 uses the climatic variables diffuse radiation,
227 relative humidity, and precipitation seasonality, while Model 2 uses leaf area index (LAI), hot
228 mean, and dry mean.

229 After getting the parameter estimates for each model and species, we created different maps
230 to visualize and compare the outputs of Model 1 and Model 2 as follows (step 4 in Fig. 2). First,
231 we used the estimated parameters to create suitability maps by calculating an index, for each
232 cell, whose values range from zero to one (see Jiménez and Soberón, 2022). A suitability index
233 close to 1 means that the environmental conditions associated to that cell are close to the center of
234 the estimated fundamental niche, while a suitability index close to 0 means that the environmental
235 conditions at that cell are either close to the border of the niche or outside of the niche. In each case,
236 we defined the border of the estimated fundamental niche through the 95% confidence ellipsoid
237 of the underlying multivariate normal model. Then, we transformed the suitability maps into
238 binary maps where we assigned a positive value to cells that fall inside the fundamental niche,
239 and a value of zero to cells outside the niche and whose environmental conditions are considered
240 as unsuitable for the survival of the species. For each species, we combined the binary maps for
241 Model 1 and Model 2 by adding the corresponding raster files. With this, we obtained maps whose
242 cells were grouped into four categories: (i) *unsuitable*, these correspond to sites with environmental
243 conditions that are outside of the estimated fundamental niches under both Model 1 and Model
244 2; (ii) *suitable under Model 1*, containing sites inside the fundamental niche estimated under Model
245 1; (iii) *suitable under Model 2*, sites inside the fundamental niche estimated under Model 2; and
246 (vi) *suitable under Models 1-2*, representing sites predicted as part of the fundamental niche under
247 both Model 1 and Model 2. For further analyses, we only kept regions that were part of the last
248 category.

249 **Impact of non-native species on projected species ranges**

250 We created new maps to evaluate to what extent the non-native ferns, *Pityrogramma calomelanos*,
251 *P. austroamericana*, and *Cheilanthes viridis*, occupy regions identified as the predicted potential dis-
252 tribution of the native species. For each native species, we filtered out the regions identified as

253 suitable under both Model 1 and Model 2 in the previous part of the analyses. We plotted those
254 regions into a new map and added the occurrence data of the non-native ferns on top (step 6 in
255 Fig. 2). Additionally, we calculated the proportion of occurrences of non-native ferns that overlap
256 with the suitable regions of each native species.

257 **Impact of land-use change on projected species ranges**

258 We downloaded a raster file with a land-cover map provided by the U.S. Geological Survey (Jacobi
259 et al., 2017), which was developed as part of a comprehensive assessment of carbon sequestration
260 potential by natural ecosystems in the State of Hawaii (Selmants et al., 2017). This raster layer
261 depicts the land cover and degree of human disturbance to plant communities on the seven main
262 Hawaiian Islands. Each cell in the raster was classified in a hierarchical fashion within 48 detailed
263 land-cover units, which were also grouped into 27 general land-cover units, 13 biome units, and 7
264 major land-cover units. We reclassified the raster and created two new land-cover classifications
265 (see Table S3 in Supplemental Materials), one where we recombined the biome and general land-
266 cover units into 18 new classes. For the second classification, we considered four habitat statuses:
267 native, mixed (native-alien), alien, and developed, which refer to the degree of disturbance to
268 vegetation.

269 We calculated the intersection between each of the new land-cover classifications and the bi-
270 nary maps indicating which cells were predicted as part of the fundamental niche of each species
271 under Models 1 and 2 (step 5 in Fig. 2). For this, we first projected and cropped the land-cover
272 rasters so the extent and resolution match the binary maps. We then combined the land-cover and
273 binary rasters into a single raster layer. With the resulting raster, we calculated the proportion
274 of area (relative to the total area predicted as suitable by Models 1-2) covered by the intersection
275 of the suitable areas and each of the land-cover classes. In this way, we were able to quantify
276 how much land in the main Hawaiian islands is climatically suitable for each native fern but is no
277 longer available due land-use changes (Fig. 5).

278 **Code availability**

279 All the analyses described above were performed using R code (R Core Team, 2020), which we
280 share through this GitHub repository: <https://github.com/LauraJim/Hawaiian-Ferns>. More

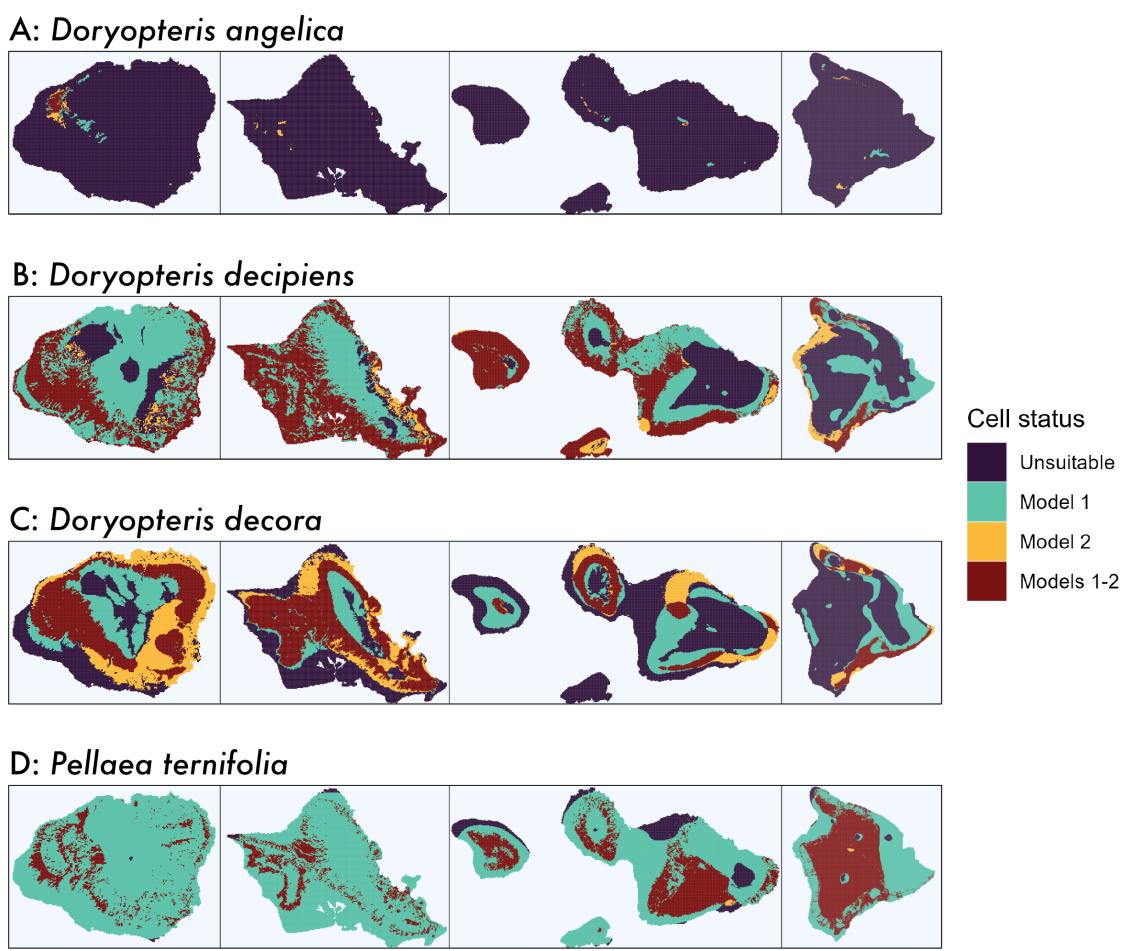


Figure 3: Suitability under Model 1 (teal), Model 2 (yellow), and the overlap of 1 and 2 (burgundy) for native Hawaiian dryland ferns: (A) *Doryopteris angelica*, (B) *Doryopteris decipiens*, (C) *Doryopteris decora*, and (D) *Pellaea ternifolia*.

281 information regarding the packages and functions we used is provided in the GitHub repository
282 and the Supplemental Material.

283 RESULTS

284 Climatic niche of native species

285 We obtained a total of 25,244 occurrences for the seven (native and non-native) species of interest
286 from all the browsed databases. After data cleaning and filtering, we retained 342 occurrences
287 (shown in Fig. 4). This final dataset includes 55 occurrences of *P. calomelanos*, 82 of *P. austroameri-*
288 *cana*, 53 of *C. viridis*, 79 of *P. ternifolia*, 29 of *D. angelica*, 45 of *D. decipiens*, and 15 of *D. decora*. These

Table 1: Percent of occurrences of the listed conspecific native species (e.g., *D. decora* for row *D. decora*) and for all non-native species that fall within the predicted ranges for native species.

Species	% occurrences in predicted range	
	conspecific	all non-native
<i>Doryopteris angelica</i>	82.7	3.2
<i>Doryopteris decipiens</i>	90.1	35.6
<i>Doryopteris decora</i>	90	54.8
<i>Pellaea ternifolia</i>	92.9	33.5

289 data include occurrences amalgamated by GBIF (2022a,b,c,d,e,f,g,h).

290 We performed a correlation analysis with the 12 climatic variables we initially selected, re-
291 taining six variables such that all pairwise correlation coefficients (r) satisfied $-0.7 > r > 0.7$:
292 diffuse radiation (W/m^2), leaf area index, relative humidity (%), maximum mean temperature
293 of the hottest four months (called hot mean; $^{\circ}\text{C}$), mean precipitation of the driest three months
294 (called dry mean; mm), and precipitation seasonality (standard deviation of the monthly precip-
295 itation values). Figure S1 shows the pairwise correlations between the original 12 variables and
296 Figure S2 shows the pairwise correlations for the selected subset of climatic variables.

297 For each native fern species, we used the cleaned set of occurrences together with the set of
298 random points—which represent the available climate in the main Hawaiian islands—to estimate
299 the parameters of the models that represent the species' fundamental niche. We fit two 3D models:
300 in Model 1, the niche axes are relative humidity, temperature seasonality, and diffuse radiation.
301 In Model 2, the axes are hot mean, leaf area index, and the dry mean. Parameter estimates are
302 provided in Table S1 at the Supplemental Material. The estimated ellipsoids of *D. angelica* were
303 the smallest and, under Model 1, its niche is nested inside ellipsoids representing the niches of *D.*
304 *decipiens* and *D. decora* (see Fig. S5 at the Supplemental Material). Under both models, the esti-
305 mated niches of the ferns in *Doryopteris* are similar to each other—their centers are close and they
306 intersect (see Figures S5 and S6). For *P. ternifolia*, the estimated optimal conditions (ellipsoid's cen-
307 ter) were very different from the ones estimated for *Doryopteris* spp. under Model 2. This species
308 seems to be adapted to colder temperatures and presents highest leaf area indices. However, un-
309 der Model 1, *P. ternifolia*'s niche is quite broad and contains the ellipsoids that represent the niches
310 of the other native species. In the Supplemental Material, we also provide the suitability maps,
311 for each species and model, that were calculated with the parameter estimates (see Fig. S7).

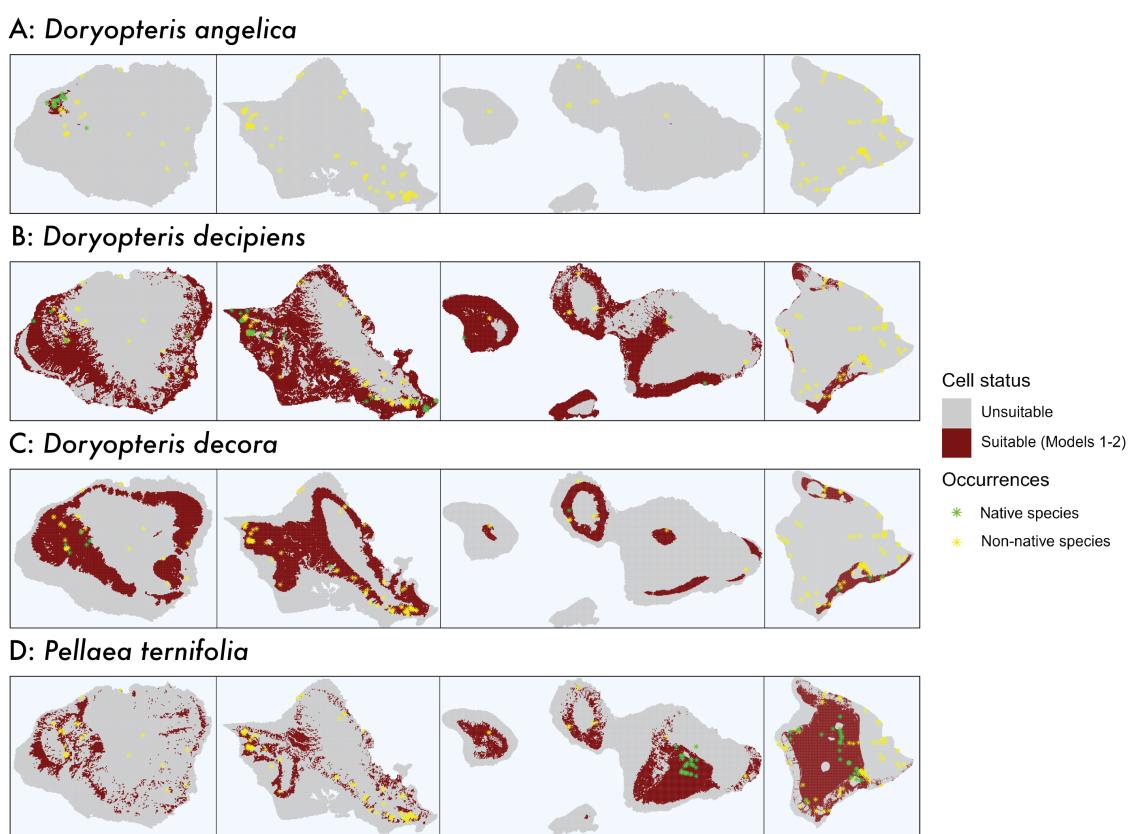


Figure 4: Combined suitability from models 1 and 2 (burgundy) for native Hawaiian dryland ferns: (A) *Doryopteris angelica*, (B) *Doryopteris decipiens*, (C) *Doryopteris decora*, and (D) *Pellaea ternifolia*. Green asterisks represent occurrences of the respective native species, while yellow asterisks represent occurrences of all studied non-natives.

312 After identifying the 95% confidence ellipsoids of the underlying model for the niche, we ob-
313 tained binary maps for Model 1 and Model 2, separately. Each pair of binary maps was then
314 combined to obtain the maps shown in Figure 3. All the main Hawaiian islands hold climatically
315 suitable areas for the four native species (except Kaho'olawe with *D. decora*). For *D. angelica*, only
316 Kaua'i contains regions identified as suitable under both Model 1 and Model 2. Under Model 1,
317 *P. ternifolia* has the largest potential range, however, after intersecting it with the predicted poten-
318 tial range under Model 2, the final potential range was drastically reduced in Kaua'i and O'ahu.
319 For further analysis, we only kept the regions holding climatically suitable conditions under both
320 models.

321 Impact of non-native species on projected species ranges

322 Figure 4 shows the binary maps obtained for each native species. The regions colored in bur-
323 gundy represent cells that were identified as part of the fundamental niche of the species under
324 both Model 1 and Model 2, which we call the predicted potential distribution for the species. *D.*
325 *angelica* has the smallest predicted potential distribution within the main Hawaiian islands, where
326 only Kaua'i (left panel in Fig. 4) holds sites with suitable conditions for this species. The pre-
327 dicted potential distributions of *D. decipiens*, *D. decora*, and *P. ternifolia* are wider and all the main
328 islands have regions with suitable conditions (except Kaho'olawe with *D. decora*). We added the
329 occurrence points of both native (green points) and non-native ferns on top of these maps (yel-
330 low points) to evaluate two questions: (1) Does the predicted potential distribution include sites
331 where the species was already identified as present? and (2) To what degree do the non-native
332 ferns occupy the potential distribution of the native ferns? The predicted potential distribution
333 of *D. angelica* correctly predicted 82.75% of the occurrence points, while 3.2% of the non-native
334 ferns occurrence points are inside this range. For *D. decipiens*, the predicted potential distribution
335 prediction included 90.1% of the conspecific native occurrences and 35.6% of the non-native oc-
336 currences are inside the species' predicted potential distribution. 90% of the occurrences identified
337 as *D. decora* are inside its corresponding predicted potential niche, which contains the highest per-
338 centage of non-native occurrences at 54.8%. Finally, *P. ternifolia*'s predicted potential distribution
339 contained 92.9% of the occurrences and 33.5% of the non-native occurrences.

340 Impact of land-use change on projected species ranges

341 We overlaid the potential distribution (under Model 1 and Model 2) of each species with the land-
342 cover layer to partition the climatically suitable areas into land-use types. The bar plots presented
343 in Figure 5 show the proportion of suitable area covered by land-use type, according to each of
344 the land-use classifications of interest (see also Table S2 in the Supplemental Material). Under
345 the land-use classification based on biomes and general land-use units, 17 out of the 18 categories
346 intersected with the suitable area. For *D. angelica*, 80.6% of the suitable area is covered by mesic for-
347 est. About 29% of the potential distribution of *D. decipiens* presents some degree of development
348 or is not vegetated. For both *D. decipiens* and *D. decora*, about 6% of their potential distribution is

349 currently used for agriculture. About a quarter of *P. ternifolia*' potential distribution is not vege-
350 tated (likely because of this species' affinity for high elevation volcanic habitat with limited plant
351 cover).

352 Under the second classification, *D. angelica*'s potential distribution is mainly covered by native
353 (58.5%) and alien (35.2%) habitats (see Table S2 in the Supplemental Material). Developed (14%)
354 and alien (69.3%) habitats cover more than 80% of *D. decipiens*'s potential distribution, while only
355 6.9% of the suitable area is represented by native habitats. We observed a similar pattern for
356 *D. decora* where only a small fraction of the suitable area is covered by native habitats (16.9%).
357 However, *P. ternifolia*'s potential range still presents a native habitat coverage of 37.8%, while the
358 remaining area is mainly covered by mixed (27.9%) and alien habitats (32.4%).

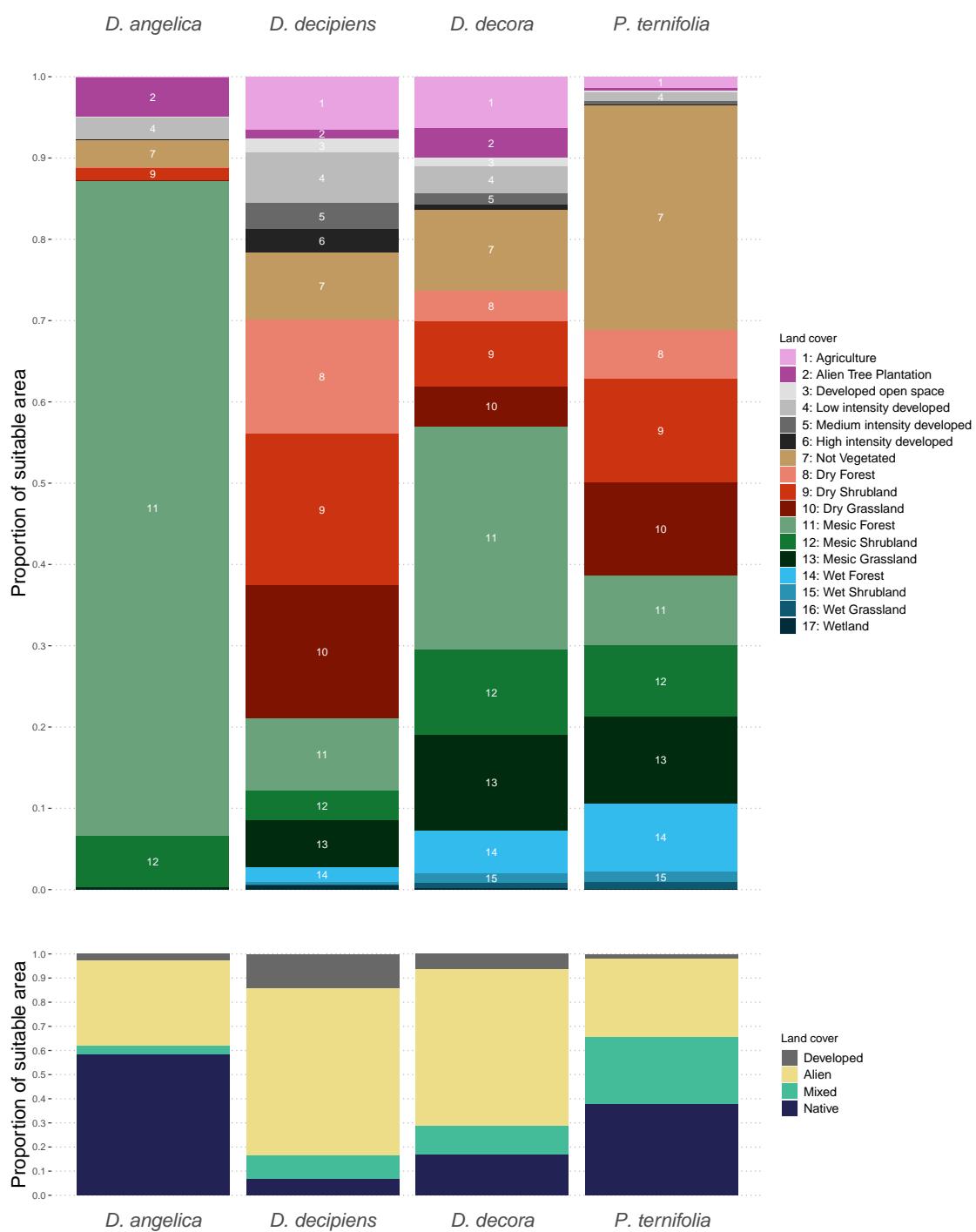


Figure 5: Land uses of inferred climatically suitable areas for native species species. The barplots show the proportion of the total suitable area (sites inside the 95% ellipsoids representing the fundamental niche) that is covered by each land-use type in the corresponding classification. The first classification combines biomes and general land-cover units and the second classification includes four habitat statuses from the most disturbed (Developed) to the most conserved (Native).

359 **DISCUSSION**

360 Hawai'i is a longstanding model for understanding ecological and evolutionary processes, and it
361 is exceedingly important to continue studying rapid changes in the islands in the face of the global
362 biodiversity crisis. Non-native plant species have a number of very different impacts on native
363 ecological processes in Hawai'i. They can deprive native species of water or nutrients and act as
364 primary or alternate hosts for pests and diseases (Smith, 1985). However, a common impact is
365 physical displacement when non-native species colonize disturbed sites and occupy them before
366 the native species can reestablish. Similarly, land-use changes such as agriculture and human de-
367 velopment may render climatically appropriate habitat unavailable to native species. Our pipeline
368 for modeling species niche while accounting for competition from non-native species and land-
369 use change demonstrates the importance to conservation efforts of modeling multiple dynamic
370 threats to biodiversity, especially in the highly disturbed habitat of Hawai'i dryland ecosystems.
371 We illustrate this pipeline using ferns—plants that tend to not be dispersal-limited given the ability
372 of wind-dispersed spores to travel long distances (Schaefer and Fontaneto, 2011) and thus whose
373 distributions are more likely to be shaped by abiotic and biotic factors.

374 **Importance of land-use data for predicting species' suitable areas**

375 Our results demonstrate an important lesson for plant conservation: an accurate picture of plants'
376 conservation statuses must account for present and historical land use. We demonstrate that while
377 native species *Doryopteris decipiens* and *D. decora* have broader fundamental niches—and thus
378 more suitable land area in Hawai'i—than the congeneric *D. angelica*, much of the land predicted as
379 suitable for *D. decipiens* and *D. decora* is dominated by human development or alien habitat (Fig. 5).
380 Of the four native species in our study, only one species (*Doryopteris angelica*) had more than half
381 of its predicted suitable area characterized as primarily native habitat, and this species also has by
382 far the smallest predicted suitable area of all natives. Agriculture accounts for a small percentage
383 of the species' suitable area, though *D. decora* is more affected than the other species. *D. angelica*
384 is considered endangered by the US Fish and Wildlife Service and is the target of several conser-
385 vation efforts including as Genetic Safety Net Species through the State of Hawai'i Department of
386 Land and Natural Resources and the Hawai'i State Wildlife Action Plan (DLNR, 2023). In contrast,

387 conspecifics *D. decora* and *D. decipiens* have much larger predicted suitable areas which are much
388 more dominated by non-native habitat (native = 6.95% for *D. decora* and native = 16.92% for *D.*
389 *decipiens*, see Table 1). Thus, while these species have much larger potential ranges, the quality
390 of this habitat is potentially lower, presenting additional conservation concerns. Our results thus
391 suggest that—by accounting for land use—additional conservation concern may be warranted for
392 these species. Our pipeline thus provides an avenue for conservationists to quantitatively assess
393 the impact of land use on species of interests, as suggested in Gillson et al. (2020).

394 We contextualize the temporally static land use change data from (Jacobi et al., 2017) with a
395 deeper understanding of the history of land use change in Hawai'i. The Jacobi et al. (2017) data
396 represents a snapshot of Hawai'i in 2014 when former agricultural lands had been mostly aban-
397 doned and had converted—largely through neglect—to non-native grass-dominated landscapes.
398 The low proportion of agricultural land projected as suitable for the species of our study is likely
399 an artifact of this history. This history also illuminates possible conservation futures, and we echo
400 calls for fallowed agricultural land to be targeted for Hawaiian dry forest conservation, including
401 of the native fern flora; restoration of fallow agricultural land has also been suggested as a fire
402 recovery and prevention mechanism (Bond-Smith et al., 2023; Trauernicht et al., 2018).

403 Impact of non-native ferns

404 *Doryopteris decora* is the native fern whose predicted range includes the highest proportion of non-
405 native occurrences (54.8%, Table 1), suggested that this species is the most potentially impacted
406 by competition from related non-native ferns. We emphasize that our study does not demon-
407 strate that non-native ferns are directly competing with natives, nor that these non-natives have
408 negative impacts on native species. Rather, we show that non-native ferns indeed share very sim-
409 ilar climatic preferences to native ferns and occur in habitats predicted to be suitable for natives,
410 especially of *D. decora*. While it is possible that *D. decora* persists in these environments despite
411 co-occurrence with non-natives, our difficulty in obtaining recent and accurate occurrence data
412 for this species (see METHODS: Species occurrence data) suggests that *D. decora* may be failing to
413 thrive in an invaded, altered landscape. We suggest that further conservation attention be given
414 to this species. While only 3.2% of non-native occurrences occur in the predicted suitable area
415 for federally listed *D. angelica*, the predicted area itself is significantly smaller than for all other

416 native species in this study. Visual inspect of the occurrences (Fig. 4A) shows many non-native
417 individuals (yellow) in and around the small suitable area (burgundy) and *D. angelica* occurrences
418 (green). Competition from non-native ferns may thus still be a concern for this species.

419 **Accuracy of modeling approach**

420 For most native species, our modeling approach results in predicted suitable areas that include
421 > 90% of observed occurrences. *D. angelica* is the one exception, with only 83.7% of occurrences
422 falling into the predicted suitable area. This may be because *D. angelica* is predicted to have far
423 less suitable area than the other native species, implying that our model is inferring a very specific
424 set of climatic conditions necessary for *D. angelica* to survive. *D. angelica* is known from only a
425 few populations that are clustered in similar regions of Kaua'i island, and it is possible our model
426 is inferring more strict climatic conditions because the existing individuals and populations are
427 highly clustered in climatic and geographic space.

428 On the other hand, our predicted suitable area estimate for *Pellaea ternifolia* is quite broad,
429 especially under Model (Fig. 3). Model 1 includes three climatic variables: relative humidity, tem-
430 perature seasonality, and diffuse radiation, and our results suggest that together these variables
431 hold little to no information about suitability for *P. ternifolia*. In order to compare across species,
432 we used the same set of climatic variables for all estimates, but future studies on *P. ternifolia* should
433 consider other variables that could be more appropriate for modeling suitability. Alternatively, we
434 may be accurately predicting the fundamental niche of *Pellaea ternifolia*, and the discrepancy be-
435 tween observed range and predicted suitable area may be due to the role of biotic interactions in
436 limited where *Pellaea ternifolia* can occur.

437 **Applicability of pipeline to future studies**

438 The pipeline we outline in this study (illustrated in Fig. 2) may guide future studies that use
439 ENMs for conservation applications. Here, we outline three main takeaways for researchers to
440 use moving forward:

441 1. Our approach allows for and promotes the careful curation of both climatic and occurrence
442 data (steps (1) and (2) in Fig. 2). Rather than use a broad set of generic climatic variables, we

443 obtained regionally specific and high-resolution climatic layers that describe features likely
444 to be relevant to the species of interest based on their ecology. We encourage researchers
445 to seek out similar datasets whenever possible, though we also acknowledge that many
446 regions of the world do not have access to such data (e.g., see discussion in Soria-Auza
447 et al. 2010). Furthermore, we acknowledge that some of our selected climatic layers appear
448 to be insufficient for *Pellaea ternifolia*, and we recommend that researchers evaluate model
449 predictions given prior information on species ranges and ecology. This may be a challenge
450 for researchers hoping to parallelize analyses of many species (e.g., for regional assessments
451 of the vulnerability of a particular flora). Our approach is thus more appropriate for targeted
452 study of a few species of interest.

453 2. Our integration of land use into evaluations of predicted suitable area for natives (step (5)
454 in Fig. 2) requires a place-based understanding of the history behind current land use. We
455 position our study within the context of land-use change in Hawai‘i; land privatization during
456 Great Māhele encouraged the development of plantations in former Hawaiian drylands.
457 Despite the current abandonment of most plantations, much of that land is still unavailable
458 for native species because non-native species now occupy former native habitat. Our land
459 use change results demonstrate the impact of this history on native dryland plants; only
460 one native species is predicted to have more than 50% climatically suitable area covered by
461 native ecosystems.

462 3. We emphasize the importance of using a mathematical model that aligns with our theoretical
463 definitions of the niche. Our approach uses the Jiménez and Soberón (2022) model to
464 estimate the fundamental niche—the set of climatic conditions that an organism could in-
465 habit in the absence of biotic interactions. Thus we must also evaluate the potential role of
466 other species. In our case we set out to evaluate how non-natives may impose additional re-
467 strictions on where natives occur. We find evidence suggesting that non-native species may
468 be competing with natives (particularly with *D. decora*) for limited available dryland habitat.
469 We encourage researchers to ensure that their model of choice aligns with their theoretical
470 understanding of the ecological niche.

471 Conclusions

472 Our novel pipeline unites new developments in ecological niche modeling with a place-based un-
473 derstanding of land-use change and potential competition from non-native species. We demon-
474 strate that native Hawaiian dryland ferns may be competing for limited suitable habitat with non-
475 native relatives, and our results point to the drastic reduction in available area due to land-use
476 changes such as human development and non-native ecosystems, especially fallowed agricultural
477 lands. We advocate for continued conservation efforts for the endangered *Doryopteris angelica*, ad-
478 ditional conservation assessment for *D. decora*, and restoration of fallowed agricultural lands to
479 build up more native habitat. Our study also demonstrates the value of our integrative approach
480 and serves as a model for multidimensional conservation assessments based on ecological niche
481 modeling.

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491 Author contributions

492 K.E.C, C.M.T., and R.Z.F conceived the study. K.E.C, L.J, and C.M.T performed analyses and wrote
493 the paper. R.Z.F., M.K.T., and K.H. provided feedback on analyses and edited the paper.

494 **Data availability statement**

495 All the R code needed to reproduce all our analyses and figures is available at this GitHub repos-
496 itory: <https://github.com/LauraJim/Hawaiian-Ferns>. The occurrence data for all the species
497 is provided in the same GitHub repository, except for *Doryopteris angelica* given conservation con-
498 cerns regarding this native Hawaiian species. The rasters with the climate variables and land-use
499 layers can be found at their original sources using the links provided in the Methods section.

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