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SPECIALTY SECTION

This article was submitted to
Models in Ecology and Evolution,
a section of the journal
Frontiers in Ecology and Evolution

RECEIVED 10 October 2022

ACCEPTED 20 February 2023

PUBLISHED 16 March 2023

CITATION

Varela D, Romeiras MM and Silva L (2023)
Present and future distribution of *Faidherbia*
albida in Cabo Verde as revealed by climatic
modelling and LULC analysis.
Front. Ecol. Evol. 11:1057852.
doi: 10.3389/fevo.2023.1057852

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Present and future distribution of *Faidherbia albida* in Cabo Verde as revealed by climatic modelling and LULC analysis

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Climate change poses one of the most significant challenges to conserve biodiversity, especially in tropical dry islands, as is the case of Cabo Verde (northeast Atlantic Ocean). This archipelago has a low percentage of forest cover and hosts only seven native tree species, among them, *Faidherbia albida* (Delile) A.Chev. (Fabaceae). Therefore, protective afforestation is extremely important in Cabo Verde, one of the most vulnerable West African countries to climate change. With this work, we aimed to estimate the current distribution and potential shifts in suitable areas for *F. albida* under climate change, using species distribution models (i.e., random forest, generalized linear and additive models), covering its distribution range in Cabo Verde and mainland Africa. The best model was then projected for the studied area, at two different slice times, using Representative Concentration Pathway (RCP) 4.5 and 8.5 scenarios. Based on current bioclimatic variables, we estimated that almost two thirds of Cabo Verde's territory is highly suitable for *F. albida*, which contrasts with its current occurrence. By overlaying the present habitat suitability with land use and land cover data, we concluded that habitat availability and suitability could be constrained by that factor. On average, the predicted suitable habitat for future distributions gradually decreases by 2080 under both scenarios compared with the current, with a smaller effect of RCP4.5 than of RCP8.5. Local authorities can benefit from this research and develop actions to promote sustainable reforestation in Cabo Verde, which should include native tree species that are best adapted to the local climate and could thus contribute to mitigate the effects of climate change.

KEYWORDS

tropical dry islands, native tree species, species distribution models, climate change, habitat suitability, Cabo Verde, West Africa

1. Introduction

There is a very high probability that global warming will reach or surpass 1.5°C in the near future (2030–2050), as a result of insufficient progress toward the Sustainable Development Goals (IPCC, 2022). The West African region is already experiencing a continuous warming trend, with a temperature increase of 0.2–0.5°C per decade in countries such as Ghana, Côte

d'Ivoire, Guinea, and Senegal (Sylla et al., 2016). The vulnerability of these low-income African countries to natural disasters (e.g., drought) is particularly worrying in the context of climate change, mainly due to the importance of rainfed agriculture in the regional economy, and the low level of water management (Varela et al., 2020). Also, climate and land-use changes have significant impacts on the distribution of plant species (e.g., Parmesan and Hanley, 2015; Feeley et al., 2020).

Among the many semi-arid climates of Africa, most countries – including Cabo Verde islands, the focus of this study – have been heavily affected by drought over the last decades, with highly variable and low rainfall (Monteiro et al., 2020). This situation worsens in the climate change projections and will directly and indirectly affect species distribution and abundance at a global scale, as well as the growth and productivity of forests through changes in temperature, precipitation, and other climate-related factors (Chang et al., 2015; Ferrarini et al., 2019; Santini et al., 2021).

Located in the northeast Atlantic Ocean, Cabo Verde is located c. 500 km west of Senegal (West Africa) and is the southernmost archipelago encompassed by Macaronesia (i.e., Azores, Madeira, Savage Islands, Canary Islands, and Cabo Verde; Florencio et al., 2021). The wide variety of habitats within the Macaronesian Islands is home to a rich biodiversity, but this huge diversity is not properly understood, and many lineages remain understudied, in Cabo Verde (Romeiras et al., 2019). Particularly, the dry tropical forests and tree species of these islands are poorly known in comparison to those of the other Macaronesian archipelagos (Norder et al., 2020; Castilla-Beltran et al., 2021). There are only seven tree taxa native to Cabo Verde (Romeiras et al., 2016): three are endemic – *Dracaena draco* subsp. *caboverdeana* Marrero Rodr. & R.S.Almeida; *Phoenix atlantica* A.Chev; *Sideroxylon marginatum* (Decne. ex Webb) Cout.; and four are non-endemic – *Faidherbia albida* (Delile) A.Chev. (Family: Fabaceae Lindl.), *Ficus sur* Forssk., *Ficus sycomorus* L., and *Tamarix senegalensis* DC.

Effective protection, management, and future expansion of the remnants of Cabo Verde's native forests with their full array of species depend on the basic knowledge of the factors constraining species distribution, as well as a comprehensive account of their current distribution and environmental suitability maps. Nowadays, Species Distribution Models (SDMs) emerged as useful tools in spatial ecology, conservation, and land management (e.g., Thuiller et al., 2011, 2016; Casajus et al., 2016) by establishing a connection between species occurrences (i.e., species presence or abundance) and a set of explanatory variables that are mostly environmental descriptors (e.g., bioclimatic variables, altitude, solar radiation, land cover/use, type of soil/substrate) that can then be projected under various climate scenarios (Silva et al., 2017, 2019, 2021). Moreover, SDMs generate geographical maps of environmental suitability that can be used to prioritize conservation areas, forecast the effects of climate change on species ranges, and assess the likelihood of biological invasions (e.g., Araújo et al., 2005; Casajus et al., 2016; Thuiller et al., 2020). There are numerous modelling techniques, including Random Forest (RF), Generalized Linear Models (GLM), and Generalized Additive Models (GAM; e.g., Breiman, 2001; Prasad et al., 2006; Wooldridge, 2008; Dobson and Barnett, 2018; Silva et al., 2021). These methods are reliable to model tree species distributions in Macaronesian islands, namely in Canary Islands (Arévalo et al., 2005) and Azores (Silva et al., 2017, 2019); recently, Varela et al. (2022) delimited the climatic

niche of Cabo Verde endemics (i.e., *Dracaena draco* subsp. *caboverdeana*; *Phoenix atlantica*; *Sideroxylon marginatum*) and the effects of predicted climate change on their future distribution range.

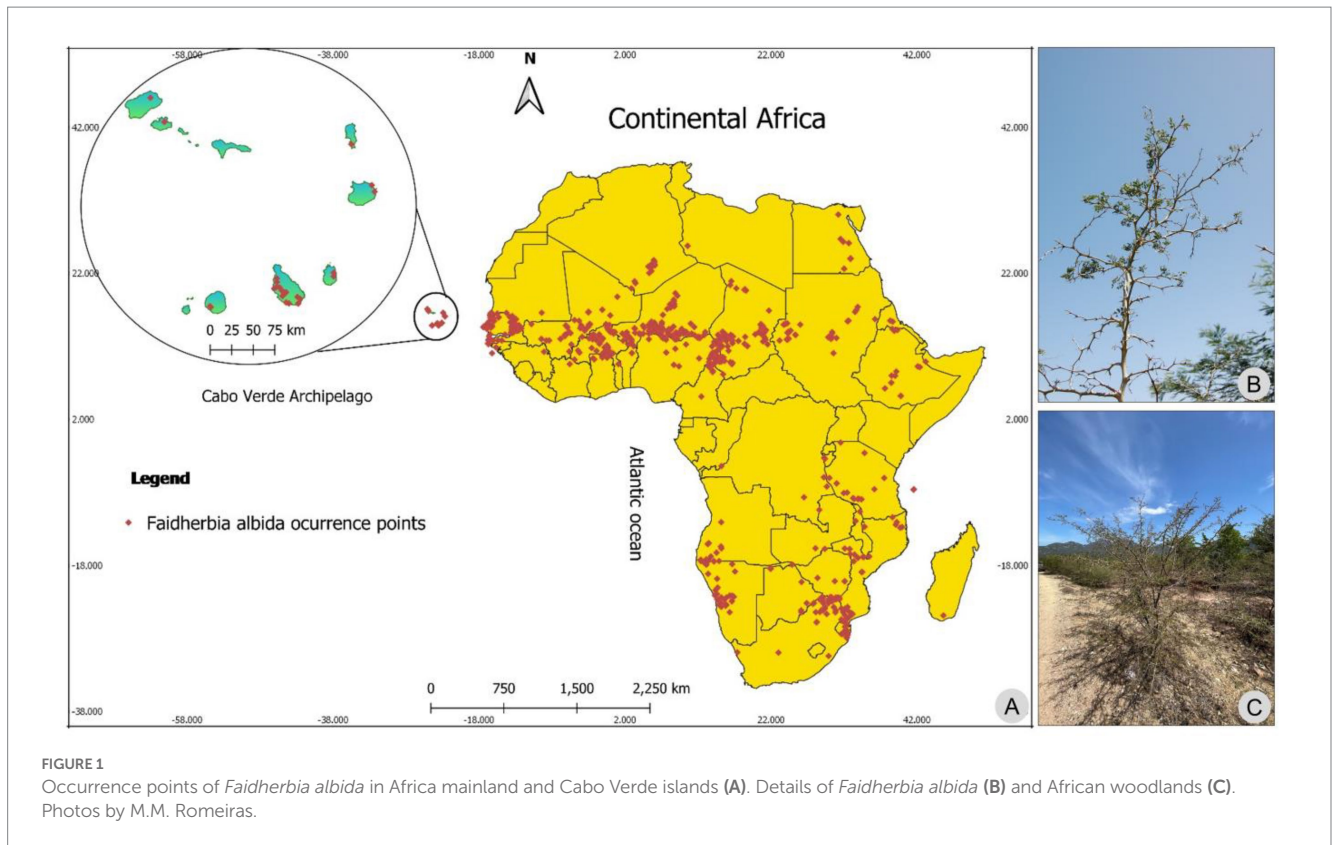
To our knowledge, the factors determining the present and future distribution of the native non-endemic Cabo Verde trees, all with their native range in Africa, were not yet documented. Among them, the leguminous *Faidherbia albida* that is an important component of Cabo Verde's native vegetation, and can be used in reforestation programs, as it is adapted to semiarid African habitats (Figure 1); it can thus contribute to biodiversity conservation in these islands, counteracting forest cover loss and promoting the cultivation of species resilient to climate change (Romeiras et al., 2016).

Based on current and future climate scenarios, some African legume trees were recently studied (e.g., Adjonou et al., 2020; Atanasso et al., 2021; Catarino et al., 2021). For instance, Nyairo and Machimura (2020) revealed the range projections for nine taxa in Kenya, highlighting that a mix of expansion or contraction occurs, according to the species' ability to establish in urban areas or inability to endure the predicted climate change. Furthermore, testing hypotheses on diversification processes, Gorel et al. (2019) suggested that ecological speciation through climate probably played a key role in the evolution of the tropical Leguminosae *Erythrophleum* genus, widespread across Africa, whose climatic niche indicates adaptive divergence along rainfall gradients, probably boosted by past climate fluctuations.

Faidherbia albida is a monospecific taxon (POWO - Plants of the World Online, 2021) and is a potentially good model to test possible environmental correlations among different geographic areas, using climatic modelling and analyses of land use and land cover (LULC) changes. The distribution of this tree has been successfully modeled using MAXENT, in central Senegal, although with a relatively low AUC (0.64), where a relationship between tree diversity and human activities was found (Ndao et al., 2022).

Most of native and endemic trees of Cabo Verde have seen their distribution ranges reduced (Varela et al., 2022), most likely due to human action, while they could play a decisive role in a more sustainable development, further considering the possible effects of future climate change. Therefore, to support the definition of sounder land management actions devoted to the valorization of the ecosystem services provided by those taxa, we need a better understanding of the factors affecting their distribution. In this context, this study aimed to estimate the current distribution and potential shifts in suitable areas for *F. albida* under climate change, using species distribution models. We used SDMs to determine the climatic niche of *F. albida*, both in mainland Africa and in Cabo Verde. We then projected its potential distribution in Cabo Verde based on the current climatic conditions and according to the future climate change scenarios. Finally, we overlapped the present potential distribution of *F. albida* in Cabo Verde with LULC data to identify possible constraints. We developed our distribution models also including data from mainland Africa, since we wanted to determine, as accurately as possible, the climatic niche of the study species. Working only with distribution and climatic data from Cabo Verde could potentially lead to a biased circumscription of the climatic niche, since *F. albida* distribution range in Cabo Verde is much more limited than in mainland Africa.

We think that this first step constitutes the baseline for the development of innovative research and management actions that will



allow the valorization of *F. albida* in terms of biodiversity preservation, soil conservation, carbon accumulation and human use in Cabo Verde.

2. Materials and methods

2.1. Study area

The archipelago of Cabo Verde (Figure 1) lies between latitudes 14–18°N and longitudes 22–26°W and is included in the African Sahelian arid and semi-arid climatic region, the largest semi-arid region of the world, which extends across Africa from the Atlantic Ocean to the Indian Ocean (Varela-Lopes and Molion, 2014; Neto et al., 2020). The hydrological resources are scarce, with an average annual precipitation of 352 mm/year from 1990 to 2016, well below the average of 653 mm/year recorded in the African continent for the same period (CCKP, 2019).

The islands of Cabo Verde were uninhabited until the 15th century, and their climate, tropical and drier than in the other Macaronesian archipelagos, did not encourage the discovery of its flora (Romeiras et al., 2020). The first field expeditions made by naturalists only occurred at the end of the 18th century (more details in Romeiras et al., 2014). The broad scale patterns of plant richness are mainly affected by climate and topography, as altitude and exposure to northeast trade winds lead to contrasting weather conditions (Duarte et al., 2008). The northern (i.e., São Nicolau, São Vicente, and Santo Antão) and southern islands (i.e., Brava, Fogo, and Santiago) are characterized by high mountains, offering a wide range of habitats over relatively short distances, whereas the eastern islands (i.e., Maio, Boavista, and Sal) are lower and drier (Neto et al., 2020).

2.2. Study species

The apple-ring acacia or *espinheiro-branco* (common name in Cabo Verde) *Faidherbia albida* (Figures 1B,C; synonyms: *Acacia caboverdeana* Rivas Mart., Lousã, J.C. Costa & Maria C. Duarte; *Acacia albida* Delile: see Rivas-Martínez et al., 2017) is considered native to Cabo Verde (Romeiras et al., 2016, 2020). Segregated from the genus *Acacia* by Chevalier in 1934 and confirmed by Luckow et al. (2003) based on molecular studies, this genus belongs to Ingeae and not to Acaciae. It is a leguminous tree species with an important role in West African agroforestry areas (Vandenbeldt, 1992; Lu et al., 2022). In addition, it contributes to symbiotic nitrogen fixation and provides a variety of ecosystem services, such as nutrient recycling and carbon sequestration, among others (Lu et al., 2022). *Faidherbia albida* has a unique phenology, as compared with other tree species since it is leafless during the wet season and grows its leaves and pods during the dry season, which reduces the competition for water, light, and nutrients with associated crops during the wet season (Vandenbeldt, 1992; Hadgu et al., 2009; Lu et al., 2022).

2.3. Occurrence data

The data on the occurrence of the study species (see Figure 1A) were collected from the following sources:

- Cabo Verde data were obtained from herbarium records (see Supplementary Table with the list of herbarium specimens); and field surveys undertaken by MR that also originated herbarium material;

TABLE 1 Description of representative concentration pathway (RCP) scenarios.

RCP	Description	Developed by
RCP 4.5	A stabilization scenario in which total radiative forcing is stabilized before 2,100 through the use of technologies and strategies to reduce greenhouse gas emissions. By 2,100, global temperatures will have risen by 1.1 to 2.6°C.	MiniCAM modelling team at the Pacific Northwest National Laboratory's Joint Global Change Research Institute
RCP 8.5	This scenario assumes a "business-as-usual" approach. CO ₂ concentrations in the atmosphere will be three to four times higher by 2,100 than they were before the industrial revolution. Thus, it is considered an unlikely scenario, with a predicted increase in global temperature of 2.6 to 4.8°C by 2,100.	MESSAGE modelling team and the IIASA Integrated Assessment Framework at the International Institute for Applied Systems Analysis (IIASA), Austria

Source: Chaturvedi et al. (2012) and Symon (2013).

- For mainland Africa, 1,354 records from 55 published datasets were available in Global Biodiversity Information Facility (GBIF.org, 2021; <https://doi.org/10.15468/dl.rxn1jv>), but several occurrence points in very specific areas (e.g., rivers, and seas) were eliminated to avoid possible outliers.

The precise geographic coordinates of each record were obtained, with an error below 10m, and all the considered occurrences of *F. albida* for Cabo Verde (34) and for mainland Africa (442) were organized into a database.

2.4. Environmental variables

The bioclimatic variables for continental Africa (Supplementary data 1, 2) were downloaded from the WorldClim v2.0 dataset¹ at a resolution of 30 arc-seconds (Fick and Hijmans, 2017). The bioclimatic variables available in WorldClim and CCAFS² have been extensively used in plant distribution modelling (Deblauwe et al., 2016; Booth, 2018; Perrin et al., 2020). To investigate the relationship between variables and to avoid excessive collinearity, we used the Variance Inflation Factor (VIF; Newman et al., 2010). The predictors with the highest VIF values were initially eliminated, and new VIF scores were calculated for the remaining predictors; this procedure was repeated until all VIF scores were less than 10. This procedure was conducted in R v3.6.0 (R Development Core Team, 2020). The data layers were prepared by using the QGIS software 2.18.20 (QGIS Development Team, 2021). All variables were projected into the same Geographic Coordinate System (WGS84), selecting the method of the nearest neighbor resampling. This approach followed Silva et al. (2019) and Varela et al. (2022).

2.5. Future climate

To examine the effects of climate change on the distribution of *Faidherbia albida*, potential distributions were projected using the 5th assessment report (AR5) of the Intergovernmental Panel on Climate Change (IPCC), available at the International Center for Tropical Agriculture website (see <http://www.ccafs-climate.org>). The AR5 was chosen as the most comprehensive assessment of climate change and

as a reliable basis for policymaking (Symon, 2013). The AR5's underlying scenarios of human influence, dubbed Representative Concentration Pathways (RCPs), are expressed in terms of greenhouse gas concentrations. Each RCP implies a different amount of human-driven climate change, resulting in a different amount of extra heat energy being stored in the Earth system due to greenhouse gas emissions (Table 1). These scenarios were developed using assumptions concerning economic growth, technology choices, and land use (Symon, 2013). In this study, we followed two scenarios (RCP 4.5 and RCP 8.5) described in Table 1.

2.6. Modelling

2.6.1. Modelling framework and algorithms

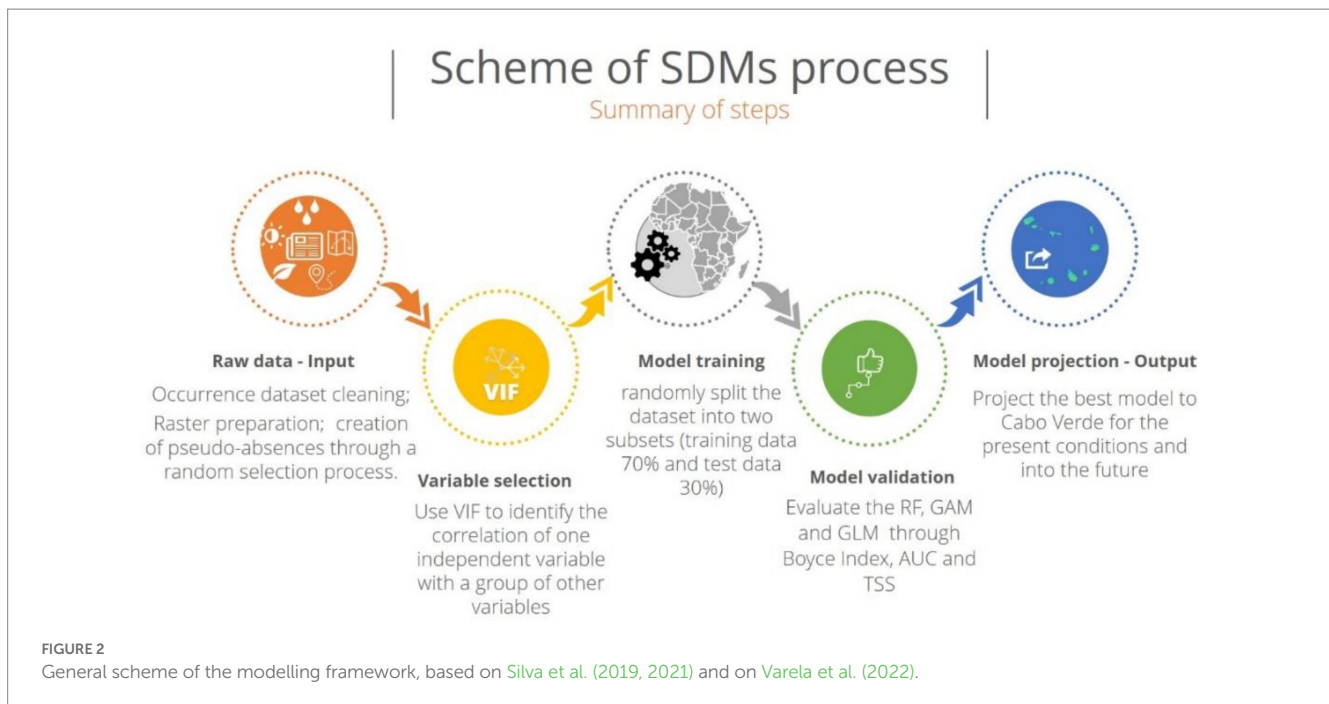
Implications of climate change on the distribution of *Faidherbia albida* were predicted through a modelling approach - BIOMOD (BIODiversity MODelling; "biomod2"; Thuiller et al., 2020). BIOMOD is a freeware, open-source package implemented in R version 3.6.1 that aims to maximize the predictive accuracy of current species distributions and the reliability of future potential deliveries, using different types of statistical modelling methods (Thuiller, 2003; R Development Core Team, 2020). We followed the approach implemented by Silva et al. (2019, 2021), and Varela et al. (2022) and the main steps are summarized in the Figure 2.

We used a total of three different SDM algorithms implemented in the BIOMOD2 package: Generalized Linear Models (GLMs; Guisan et al., 2002; Emptage and Dobson, 2008), Generalized Additive Models (GAM; Guisan et al., 2002; Rigby et al., 2005), and one machine learning method, Random Forest (RF; Breiman, 2001). RF is a regression tree created by using random feature selection in tree induction and bootstrap samples of the training data (Breiman, 2001). GLMs and GAM use a link function to establish a relationship between the mean of the response variable and the explanatory variables (Guisan et al., 2002; Rigby et al., 2005). For further methodological details about the three SDM algorithms implemented, see Silva et al. (2019, 2021) and Varela et al. (2022).

The use of RF in SDMs has been demonstrated to be robust and stable when handling multicategory or binary abundance information, e.g., presence and absence data (Hegel et al., 2010). Comparisons of different SDMs revealed that presence-absence models perform better than presence-only models (Elith et al., 2006). Because absence points were not included in our initial data set, we generated pseudo-absences through a random selection process. This technique involves randomly selecting pseudo-absence points from the background data across the entire study area. The number of pseudo-absences used for

¹ <http://www.worldclim.org/bioclim>

² <http://www.ccafs-climate.org>



all the algorithms was equal to the number of occurrences because the predictive accuracy increased when the number of pseudo-absences was approximately equal to the number of presences (Barbet-Massin et al., 2012). We used the “random Points” function in the R package “dismo” to reduce projection bias (Hijmans et al., 2020).

The model training was performed using the whole African territory. Cross-validation was used because a validation set was not available, thus we randomly split the dataset into two subsets (training data 70% and test data 30%). The first subset was used to fit the model, and the second subset was used to evaluate the model’s performance, before making predictions and comparing them to the expected values. This procedure was repeated 10 times for each modelling algorithm, as suggested in the BIOMOD2 package and applied in previous research (Silva et al., 2019, 2021; Varela et al., 2022), allowing the calculation of a mean and a standard deviation for the three evaluation parameters and for each modelling approach.

The Boyce Index, the area under the curve (AUC) of the receiver-operating characteristic (ROC), and the true skill statistic (TSS) were used to evaluate the models (Boyce et al., 2002; Allouche et al., 2006; Pearce and Boyce, 2006; Singer et al., 2017). Both evaluation indicators were calculated using the proportion of “sensitivity” (correctly predicted presences) and “specificity” (correctly predicted absences). The models’ performance to predict the currently suitable area was good, with TSS values ranging from 0.64 (GLM) to 0.92 (RF), AUC values ranging from 0.63 (GAM) to 0.99 (RF), and Boyce Index values ranging from 0.43 (GLM) to 0.84 (RF).

The best models for the current distribution of *Faidherbia albida* were projected for the present conditions and for two different future time slices (2050 and 2080) using the respective functions of BIOMOD 2 and the predicted future climate conditions from the IPCC’s 5 assessment taken from the International Centre for Tropical Agriculture website (see <http://www.ccafs-climate.org/>).

2.6.2. Land use and land cover probability maps

Taking into account the effect of Land Use and Land Cover (LULC) on habitat availability and suitability, as reported by Fahrig (2003), Zhang et al. (2017), and Wang et al. (2019), we developed maps of LULC probabilities. For this purpose, we used the Cabo Verde and West Africa LULC’s maps (see Figure 3; Supplementary data 3) provided by the United States Geological Survey (USGS; CILSS, 2016). These maps are based on visual interpretation of Advanced Spaceborne Thermal Emission and Reflection Radiometer (ASTER), which characterizes the landscapes at a resolution of 500 m and 2 km, for Cabo Verde and West Africa, respectively, between 2000 and 2013.

Based on the LULC data for Cabo Verde, on a sample of 342 cells in Cabo Verde and West Africa covering the different types of LULC, and on the occurrences and absences of *Faidherbia albida* in Cabo Verde per type of LULC category, a LULC Index was calculated using two different approaches (Supplementary data 4):

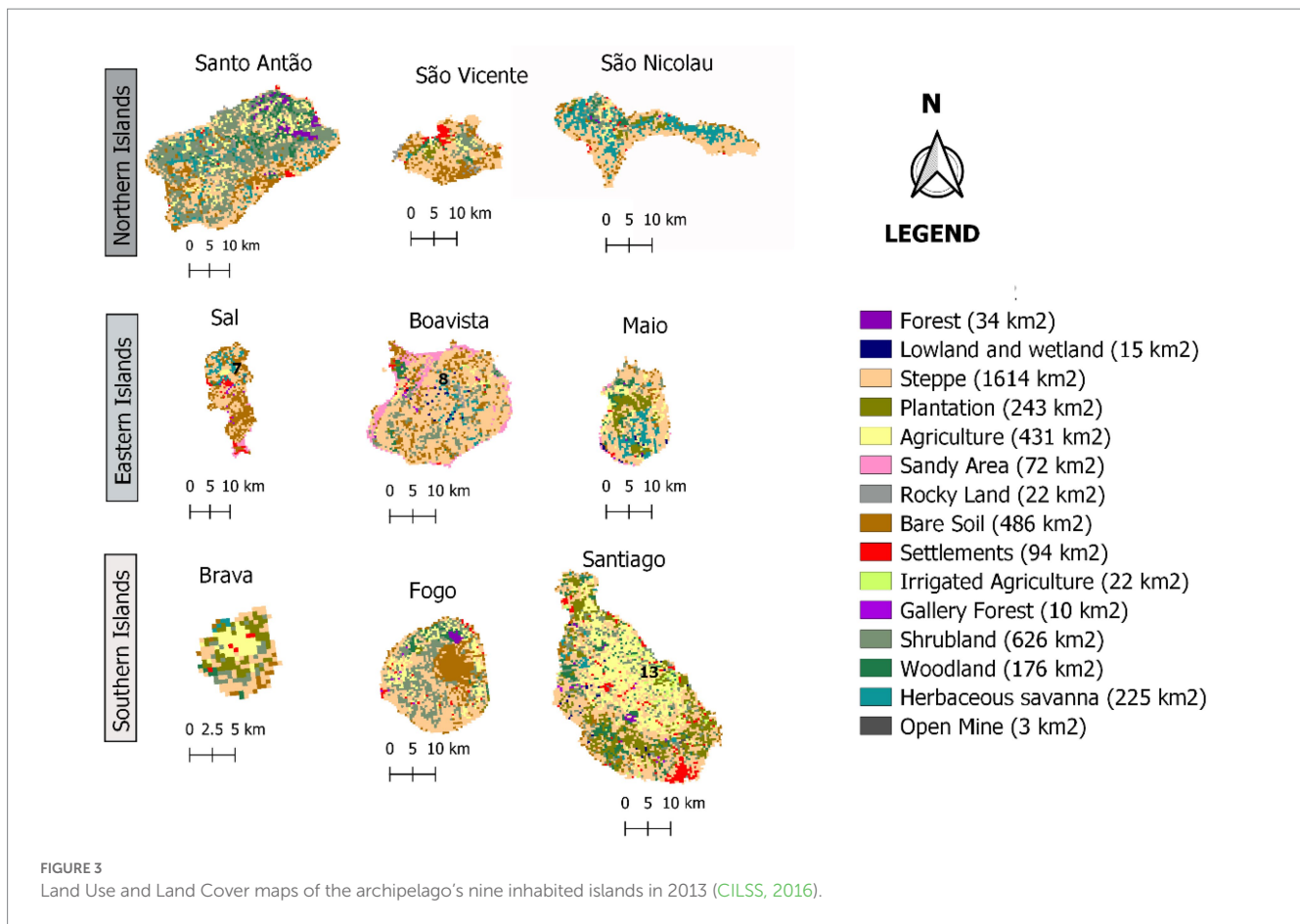
a) the ratio between the percentage of occurrences in a given LULC category and the percentage of Area of that LULC category:

Occurrences (%) for $LULC_i$ / Area (%) for $LULC_i$, for each LULC category, i .

b) the Bayes Theorem with the following calculations:

$$P(FA|LULC_i) = \frac{P(FA).P(LULC_i|FA)}{P(FA).P(LULC_i|FA) + P(!FA).P(LULC_i|!FA)}, \text{ for}$$

each LULC category, i , where $P(FA)$ is the probability of finding *Faidherbia albida* in Cabo Verde, based on its frequency (percentage of 1 km² cells with the species), $P(!FA)$ is the probability of not finding *Faidherbia albida* in Cabo Verde, based on its absence (percentage of 1 km² cells without the species), $P(LULC_i|FA)$ is the probability of finding the $LULC_i$ category in the cells occupied by this species in Cabo Verde, $P(LULC_i|!FA)$ is the probability of finding the $LULC_i$ category in the cells not occupied by this species



in Cabo Verde, and $P(FA|LULC_i)$ is the resulting probability of finding this species in a cell with the $LULC_i$ category.

Since both approaches fully agreed when we ranked the LULC categories, we then attributed the final values of the LULC Index to each category, as shown in Table 2. When producing the distribution map adjusted to the adequacy of each LULC category, for each cell in the map, we multiplied the respective LULC Index by the present habitat suitability value derived from SDMs.

According to the LULC index and human pressure, the original LULC was recategorized as Suitability Index, according to the criteria displayed in Table 2.

3. Results

3.1. Climatic niche of *Faidherbia albida*

The Table 3 summarizes the climatic conditions under which the studied tree species currently occurs, and Figure 4 depicts the descriptive statistic values of the bioclimatic variables used in modelling.

As expected, in continental Africa *Faidherbia albida* occurs under a much wider range of climatic conditions than in Cabo Verde. As a consequence, there is only a moderate difference in average values of bioclimatic variables between the two studied regions.

In terms of climate change scenarios, under RCP 4.5 in 2080 (Supplementary data 1), the average annual mean temperature in the areas currently occupied by *F. albida* will rise by 1.5°C. The mean values of some bioclimatic variables will increase, namely BIO6 (Min Temperature of Coldest Month), BIO9 (Mean Temperature of Driest Quarter), and BIO13 (Precipitation of Wettest Month). In contrast, BIO3 (Isothermality), BIO13 (Precipitation of Wettest Month), and BIO15 (Precipitation Seasonality) are expected to decrease.

3.2. Variable and model selection

Out of 19 bioclimatic variables available for modelling, nine were excluded after VIF analysis, and the remainder were used as model predictors (Table 4).

3.3. Model evaluation

Random Forest (RF) performed better than GLM and GAM (Table 5) to predict *Faidherbia albida* suitable distribution under present and future climate scenarios. TSS values ranged from 0.55 (GLM) to 0.92 (RF), and AUC values ranged from 0.64 (GLM) to 0.99 (RF).

The variables with more impact on the distribution of *F. albida* were BIO15, BIO13, BIO18, BIO6 (Figure 5).

TABLE 2 Recategorization of the LULC index into the Suitability Index.

Land Cover	Suitability index	Characteristics	Probability of occurrence
Highly appropriate (HA)	≥ 2	Low human pressure. Presence of trees, shrubs or vegetation cover. Presence of <i>F. albida</i> confirmed.	1
Semi-appropriate (SA)	0.5–2	Moderate human pressure. Areas that slightly constrain <i>F. albida</i> .	0.75
Less appropriate (LA)	$> 0-0.10$	Areas of poor edaphic conditions, such as bare soil or sandy areas like the Boavista desert. Presence of <i>F. albida</i> constrained.	0.5
Highly inappropriate (HI)	0	High human pressure. Almost no tree, shrub, or vegetation cover (e.g., settlements, open mines areas). Occurrence of <i>F. albida</i> hugely constrained.	0

Land cover refers to areas of Cabo Verde and West Africa. The attributed probability of occurrence was multiplied by the habitat suitability values resulting from the SDMs.

TABLE 3 Annual mean temperature and precipitation for the current habitat conditions of *Faidherbia albida* in mainland Africa and Cabo Verde.

	Annual Mean Temperature (°C)			Annual Precipitation (mm)		
	Median	Max	Min	Median	Max	Min
Mainland Africa	25.4	30.1	14.5	418	1,633	24
Cabo Verde	23.8	24.9	18.0	243	359	55

3.4. Model projections – Present climate

Our results show that under current climatic conditions (see Figure 4) about 89% of the African territory is unsuitable for *Faidherbia albida*. Most of the appropriate area for the occurrence of this species is the Sahel region, by the southern edge of the Sahara Desert, at the same latitude as Cabo Verde, according to our occurrences map (Figure 6).

When only bioclimatic variables were used to determine habitat suitability, almost the entire archipelago appeared suitable for the occurrence of the studied species, with 43% corresponding to values above 750, and 49% with values above 500 (Figure 6). However, most of the area of some islands (namely Santo Antão, Sal, and Boavista) showed low habitat suitability (Figure 7).

3.5. LULC effect on the climate-model projections

Despite its small size, the archipelago of Cabo Verde has an extremely diverse landscape. However, its arid steppe regions cover approximately 1,614 km², accounting for nearly 40% of the total land area. Additionally, the highest number of occurrences of *F. albida* is in steppe areas, but the lowlands and wetlands have the highest density (number of occurrences per km²).

In Western Africa, the savanna hosts the highest frequency of *F. albida* occurrences. However, during the mid-20th century, afforestation practices in Cabo Verde promoted plantations of fast-growing trees, such as *Prosopis juliflora*, which affected the vegetation structure and composition in these islands. Figure 8 shows the present

habitat suitability maps for Cabo Verde, after overlaying LULC constraints to the climatic model, which leads to a possible reduction in high habitat suitability for *F. albida*, from 49% to less than 20% of the territory.

3.6. Model projections – Future climate

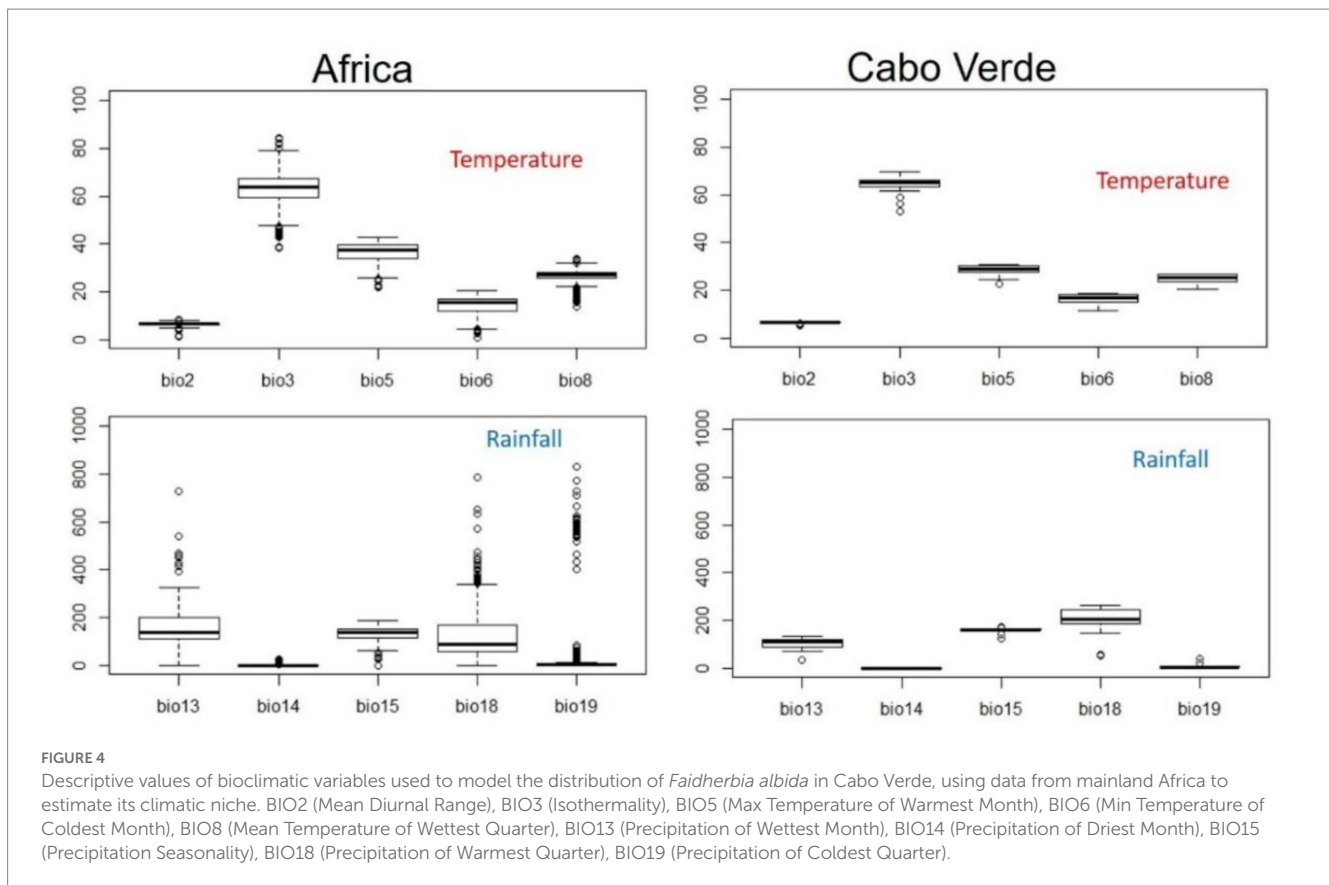
In both scenarios, according to model predictions, the extension of habitat suitable for *Faidherbia albida* in Cabo Verde is expected to slightly decrease in the future, and with almost identical patterns (Figure 9; Table 6).

By 2050, under RCP 4.5, the suitable area will decrease from 1934 to 1716 km² – about 11%. Under all scenarios, the suitable area (i.e., medium and high suitability) will remain high, corresponding to more than 30% of the territory. In the most pessimistic scenario (RCP 8.5), the expected decrease of suitable habitats is 13%. In Cabo Verde, under this scenario, the Annual Mean Temperature (BIO1) is expected to increase from 25.9°C (sd = 2.30) to a maximum of 29.6°C (sd = 2.8), a difference of 4°C. Annual Precipitation (BIO12) is expected to increase from 598 to 604 mm by 2080.

Gains occur when species colonize new areas, while species losses occur when habitat suitability is reduced in the future. Suitable habitat areas are expected to decrease mainly in Santo Antão and Santiago islands. The gain and loss maps are presented in Figure 9. For the year 2080, under scenario RCP 4.5, a slight habitat loss is expected. The high suitability category will greatly decrease in all scenarios, but the medium suitability area will increase under all climate change scenarios.

4. Discussion

Modelling the bioclimatic niche of *Faidherbia albida* – one of the few native tree species of Cabo Verde, provided new data to identify the environmental and anthropogenic constraints that might have affected their distribution in the past, and restrain future range. Our results enabling the establishment of priorities in the management and future afforestation activities, which are extremely important in Cabo Verde, one of the most vulnerable West African countries. SDMs are, therefore, an invaluable tool for decision-makers because they can improve the conservation strategies to be applied in areas deeply affected by climate change.



4.1. Climate niche, past, present, and future distributions

Concerning the global model evaluation, the RF algorithm produced the highest-quality models. Numerous studies have demonstrated that RF is excellent in predicting species distribution patterns (Ren-Yan et al., 2014; Silva et al., 2019, 2021; Varela et al., 2022). The highest quality of RF algorithm was confirmed not only through TSS and AUC (Pearce and Boyce, 2006) but also by analyzing the continuous Boyce index and the shape of the corresponding curves, which had a low standard deviation and ascending, nearly linear curves (Boyce et al., 2002; Marmion et al., 2009; Costa et al., 2013). Furthermore, Varela et al. (2022) concluded that RF was the best model to estimate the climatic niche of endemic tree species in Cabo Verde. According to the values of those parameters, we obtained better model fits than those previously obtained when modelling *F. albida* in Senegal (Ndao et al., 2022).

Faidherbia albida thrives under a broad variety of climatic conditions. For instance, in West Africa, it can occur in a desert environment where the mean annual rainfall is only 24 mm, and the mean annual temperature is 14.5°C (see Table 2). It also occurs in humid tropical areas, with 1,633 mm of annual precipitation and a mean annual temperature of 30°C. In Cabo Verde, it occurs under a wide range of climatic conditions as well. However, in locations with limited rainfall, *F. albida* requires episodic downpours of rain to grow (Barnes and Fagg, 2003).

According to our modelling results, the Cabo Verde climate seems appropriate for *F. albida*. However, its present distribution does not

reflect its potential climatic niche, which suggests that factors such as land-use changes (USGS, 2022) and the widespread expansion of non-indigenous trees (e.g., *Prosopis* spp., Cienciala et al., 2013) might have limited the present occurrence of this species in the archipelago.

A significant difference exists between endemic tree species and *Faidherbia albida* in terms of the impact of the archipelago's climate on their distribution. Varela et al. (2022) examined the effect of climate change on the distribution of the threatened endemics *Dracaena draco* subsp. *caboverdeana*, *Phoenix atlantica* and *Sideroxylon marginatum*, and found it will significantly impact their distribution. For instance, less than 5% of the territory is highly suitable for *Dracaena draco* subsp. *caboverdeana*. For *F. albida*, currently, roughly 800 km² correspond to highly suitable land, but by 2080 this area will be less than 400 km², an expected reduction of 50%, and corresponding to 9% of the territory. The medium suitable area, on the other hand, will increase, thus globally reducing the area by only 10–14%. Even though the direct consequences of climate change indicated by the models are not severe, it is vital to consider all the impacts on the biosphere on a global scale, particularly in Africa, where climate change is expected to be more severe than in other parts of the globe (Fick and Hijmans, 2017; Noulèkoun et al., 2017; IPCC, 2018a,b, 2022).

In Cabo Verde, the island of Santiago is the most suitable one for *F. albida*, since it is the largest and has the most favorable soil and climatic conditions, thus explaining the highest concentration of native tree species on this island. Indeed, in comparison to Sal, Boavista, and Maio, the mountainous Santo Antão, São Nicolau, Santiago, Fogo, and Brava, all with peaks over 1,000 meters, have the most suitable areas due to their more productive soil, cooler

TABLE 4 Environmental variables used to predict *Faidherbia albida* distribution in Cabo Verde; non-highlighted variables were removed after applying Variance Inflation Factor (VIF) analysis to reduce collinearity.

BIO1	Annual Mean Temperature
BIO2	Mean diurnal range [Mean of monthly (max temp–min temp)]
BIO3	Isothermality (BIO2/BIO7) (*100)
BIO4	Temperature seasonality (standard deviation*100)
BIO5	Max temperature of warmest month
BIO6	Min temperature of coldest month
BIO7	Temperature annual range (BIO5-BIO6)
BIO8	Mean temperature of wettest quarter
BIO9	Mean temperature of driest quarter
BIO10	Mean temperature of warmest quarter
BIO11	Mean temperature of coldest quarter
BIO12	Annual precipitation
BIO13	Precipitation of wettest month
BIO14	Precipitation of driest month
BIO15	Precipitation seasonality (Coefficient of variation)
BIO16	Precipitation of wettest quarter
BIO17	Precipitation of driest quarter
BIO18	Precipitation of warmest quarter
BIO19	Precipitation of coldest quarter

TABLE 5 Statistics used to evaluate the performance of climatic models explaining the presence/absence of *Faidherbia albida* in Cabo Verde and Africa Mainland.

Algorithms	TSS	AUC	Boyce
GLM	0.64 ± 0.04	0.77 ± 0.05	0.43 ± 0.12
GAM	0.55 ± 0.09	0.63 ± 0.07	0.51 ± 0.06
RF	0.92 ± 0.02	0.99 ± 0.03	0.84 ± 0.06

Algorithms: GLM, generalised linear models; GAM, generalised additive models; RF, random forest. Model statistics: TSS, true skill statistic; AUC, area under the curve; Boyce, Boyce index.

temperatures, and higher precipitation. According to Neto et al. (2020), climate and topography play a significant role in shaping the diversity of plant communities in Cabo Verde islands, particularly in northeast slopes of mountain areas.

4.2. Effect of LULC

The climatic conditions of Cabo Verde were shown to be adequate for *Faidherbia albida*, but, the present distribution of this tree is restricted to fairly small patches, which does not reflect its potential climatic niche. The land use of Cabo Verde underwent significant changes during human settlement (Duarte et al., 2022). Mainly intensive agricultural practices, and many introduced species have cleared a large portion of native tree populations (Monteiro et al., 2020; Varela et al., 2022). In many areas, this led to soil erosion, and

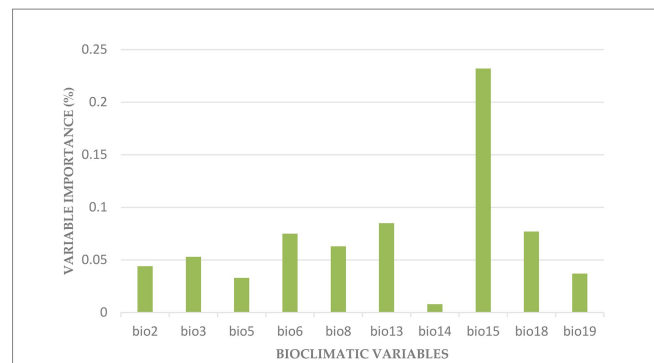


FIGURE 5

Importance of the bioclimatic variables in the final model: BIO2 (Mean Diurnal Range), BIO3 (Isothermality), BIO5 (Max Temperature of Warmest Month), BIO6 (Min Temperature of Coldest Month), BIO8 (Mean Temperature of Wettest Quarter), BIO13 (Precipitation of Wettest Month), BIO14 (Precipitation of Driest Month), BIO15 (Precipitation Seasonality), BIO18 (Precipitation of Warmest Quarter), and BIO19 (Precipitation of Coldest Quarter).

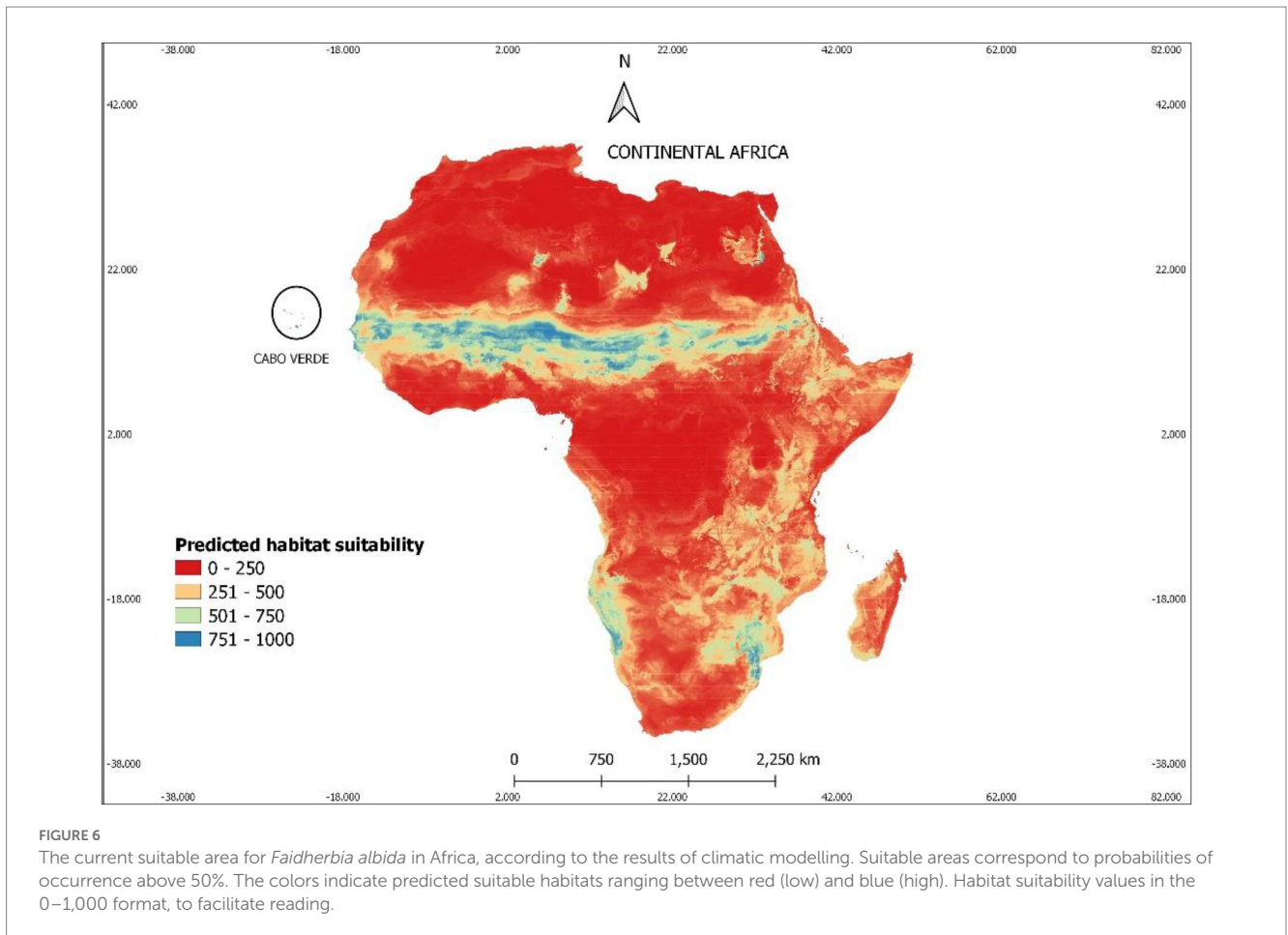
by the early twentieth century the islands were severely degraded (USGS, 2022).

The anthropogenic pressures associated with land-use and land-cover changes (e.g., fire, wood extraction, introduction of herbivores, and agricultural practices) have resulted in a shift toward sparser woody plant vegetation, after more than 500 years of human settlement in Cabo Verde (Romeiras et al., 2016; Norder et al., 2020). In the 1970s, exotic trees were widely introduced, but, until then, the only source of firewood were native trees, which may have contributed to reduce their numbers across the archipelago (Varela et al., 2022). A similar situation to that of *F. albida* was found for the endemic trees of Cabo Verde, which are presently very rare, most likely due to land use changes and native vegetation clearing in the past (see Varela et al., 2022).

According to our analysis, the most suitable land-use types for *F. albida* are, in descending order: lowlands and wetlands, rocky land, agriculture, irrigated agriculture, gallery forest, and forest. The least suitable are open mines and settlements, followed by shrubland, woodland, sandy area, savanna, grassland, bare soil, plantation, and steppe. The less suitable areas are characterized by the absence or very low occurrence of *F. albida* and low or no vegetation or tree cover. The most appropriate areas are characterized by the presence of *F. albida* and tree cover.

In the Central Rift Valley of Ethiopia, it was found that the limited seed source, caused by excessive pruning, was the main constraint for recruitment, and that an appropriate land management policy to ensure adequate seed production would avert current trends in decline of *F. albida* population (Sida et al., 2018a). A clear indication of how land use and land management practices might have been responsible for a similar decline in Cabo Verde.

Historically, the land-use change in Cabo Verde has been characterized as a transition from a dry but “well-wooded” savanna with “great quantity of grass,” and “streamlets of water,” at the time of human settlement, to a near desert landscape, especially at the lower altitudes, as a direct consequence of overexploitation of the vegetation-cover by humans and goats (see Lindskog and Delaite, 1996). Specific research devoted to that archipelago has shown that several areas are highly vulnerable to



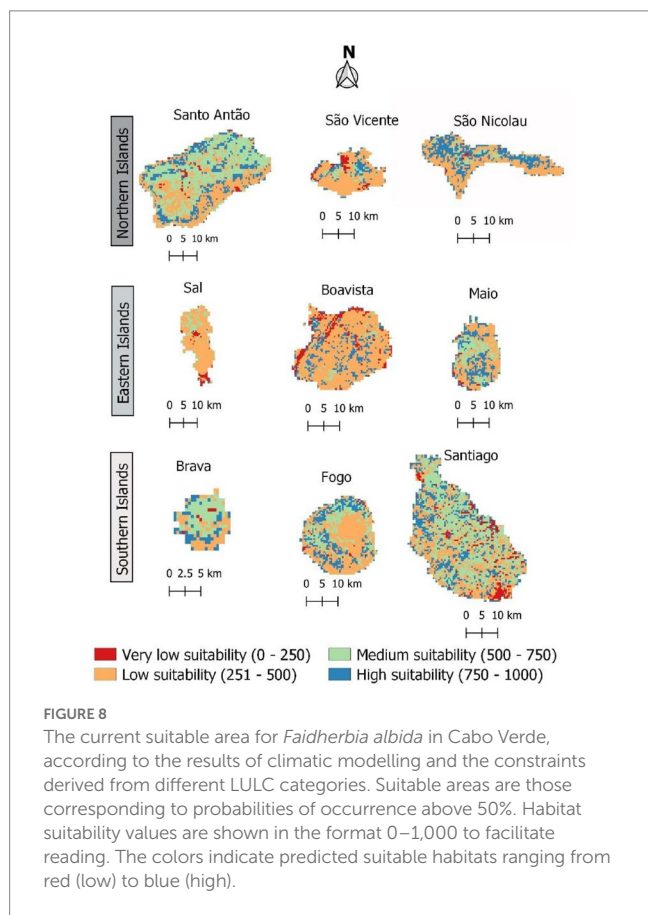
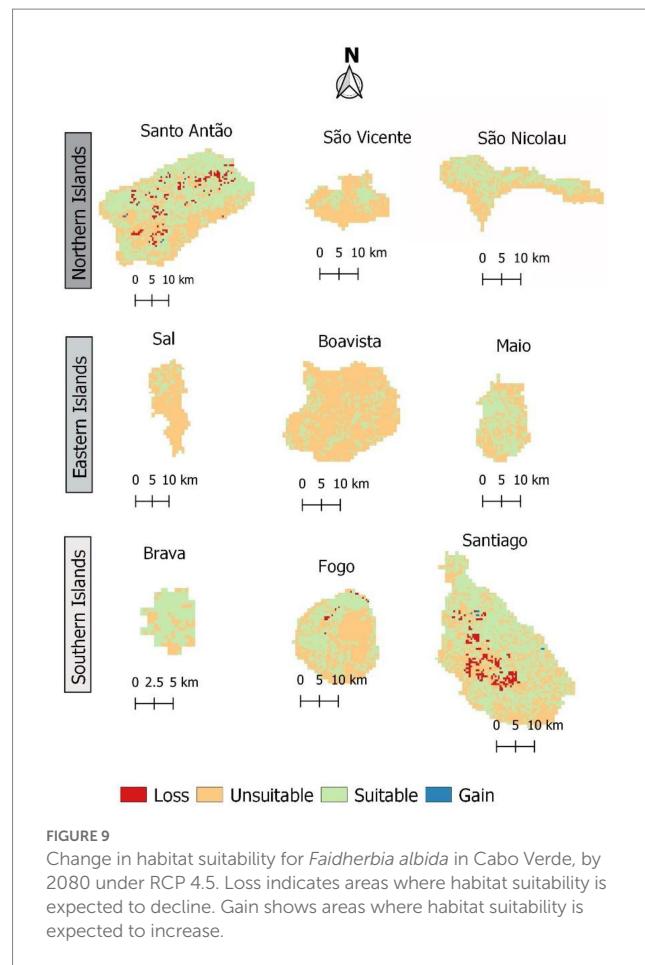
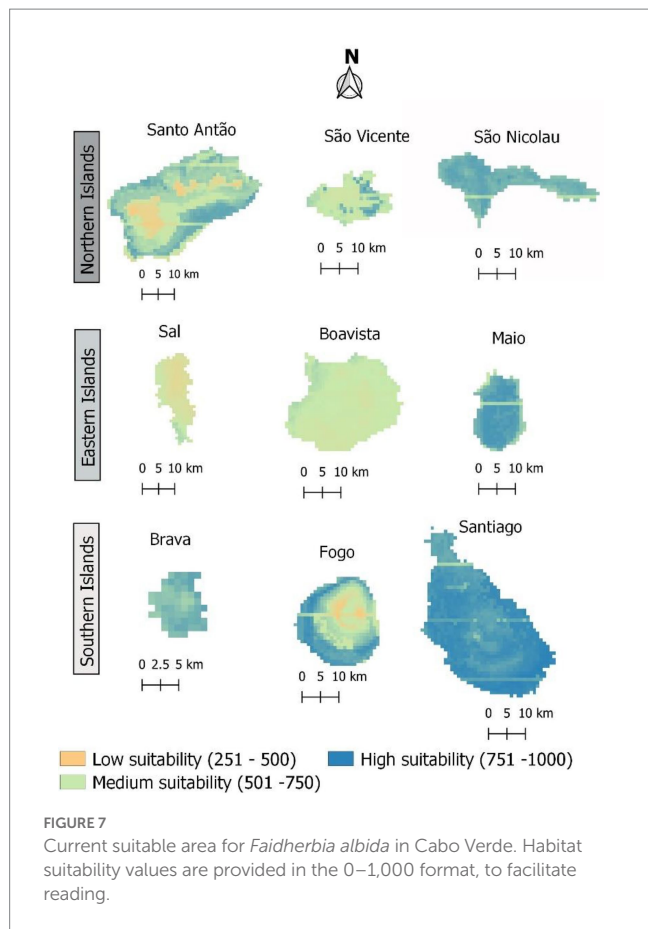
population growth, intensification of agricultural land use and increasing tourism (Olehowski et al., 2008), due to high slopes, friable geological substrates, and precipitation (higher at steeper locations). Moreover, Palaeoecological studies (Castilla-Beltrán et al., 2019) recently showed the anthropization of the highlands of Santo Antão Island, Cabo Verde. That study revealed that from ca. 350 to 250 cal yr. BP, human-induced environmental changes, such as the introduction of new pant taxa and of alien herbivores, and vegetation clearing for agricultural activities, has led to overgrazing, vegetation burning, and soil erosion. This strongly suggests that the distribution of *F. albida* in Cabo Verde was probably more widespread in the past, what would agree with the results of our modelling analyses, based on its climatic niche both for Africa and Cabo Verde, indicating a much wider potential distribution in the archipelago when compared to its presently observed occurrence.

The combined effects of land use and climate change have been considered the greatest threat to global biodiversity (García-Valdés et al., 2015; Wang et al., 2019; Chen and Leites, 2020). Thus, climate and land-use changes are expected to substantially alter future plant species distributions leading to higher extinction rates. Landscape composition and structure can impose constraints on habitat availability and suitability (Fahrig, 2003; Zhang et al., 2017; Wang et al., 2019). Therefore, the LULC probability maps that we developed seem to be a useful framework in future modelling to

better understand the role of LULC in constraining the potential distributions mainly defined by the climatic niche.

4.3. Valorization of *Faidherbia albida* ecosystem services

Given our results, areas that are climatically suitable for *Faidherbia albida* and where LULC do not profoundly constrain its occurrence should be identified for reforestation or agroforestry parklands using the studied species. *Faidherbia albida* is an important multipurpose and nitrogen-fixing tree, well known for its ecosystem services, and a keystone species for Evergreen Agriculture in Africa, with crop yields under its canopy often observed to double or triple (Hadgu et al., 2009). Agroforestry with *F. albida* significantly increases maize yields, demonstrating a direct link between this tree-based energy input and increased food security, especially given *F. albida*'s adaptation to drylands such as in the Sahel (Haskett et al., 2019). This was also found for maize production in Zambia (Yengwe et al., 2018a), where *Faidherbia* tree canopies showed to contribute to food security and mitigate the risk of crop failure on resource-poor smallholder farmers' fields, especially in drought years. Similar results have been found using *Faidherbia* plant material as maize fertilizer in Nigeria (Chinke et al., 2022). *Faidherbia* trees increased soil mineral N, wheat water



use efficiency and reduced heat stress, significantly increasing yield in the Central Rift Valley of Ethiopia (Desta et al., 2018; Sida et al., 2018b), not only in wheat but also in sorghum (Abdella et al., 2020). Furthermore, it has been shown that maize production and profitability could be maintained or improved through only partial pruning of *F. albida* in Ethiopia (Dilla et al., 2019a). In the latter, its use has been suggested as a mean to increase crop productivity, contain weed infestations, to produce livestock fodder, and as a source of nectar for honey production (Ereso, 2019), also having a considerable potential in carbon sequestration (Dilla et al., 2019b). Yengwe et al. (2018b) found that the improvement of soil fertility status by *F. albida* could be attributed to a combination of both long-term modifications of the soil biological and chemical properties, under the canopy, as well as short term litter fall addition. The global positive impact of *F. albida* on crop productivity and on soil enrichment has been recently revised by Sileshi et al. (2020).

Given the aridity of Cabo Verde islands and the possibility of contributing to the preservation and enrichment of the soil, which greatly aids in mitigating climate change, it is critical to reforest areas with a species well adapted to the present and future local climate. Furthermore, given the results obtained by Varela et al. (2022) regarding the climatic adaptation of one of the endemic trees to extreme aridity (*Phoenix atlantica*), the possibility of using consociations of both species should be considered in the future. Moreover, the possibility of using both species to improve the

TABLE 6 Suitable habitat in km² and variation in percentage, for the tree species *Faidherbia albida* in Cabo Verde, under two climate change scenarios (RCP 4.5 and RCP 8.5), based on climatic modelling.

Very low suitability	Low suitability	Medium suitability	High suitability	Suitability change
Present				
201	2,138	1,092	842	
2050 RCP 4.5				
275	2,106	1,238	486	−11%
2050 RCP 8.5				
331	2039	1,191	545	−10%
2080 RCP 4.5				
335	2060	1,233	476	−12%
2080 RCP 8.5				
310	2,111	1,290	377	−14%

conditions for agricultural practices and for the recovery of other endemic and native species should also be investigated.

Finally, it is emphasized that knowledge of the geographical range of the native trees in Cabo Verde provides essential baseline data toward achieving the goals of the Paris Agreement, which are in the country's agenda as a priority to conserve the fragile biodiversity of these islands. In this sense, we do hope that local authorities will benefit from this research and develop actions to promote sustainable reforestation, which should include native trees that are highly adapted to the local climate and could contribute to mitigate the effects of climate change.

5. Conclusion

Faidherbia albida occurs in Africa in forest galleries along rivers and is a frequent floristic element in open tropical deciduous forests and, mainly, savannas. In Cabo Verde, until the late 20th century, natural populations were significantly reduced due to its utility for livestock feeding and as firewood. Nowadays, this species is confined to temporary river zones, valleys or rocky areas, and only sporadically, as in Ribeira da Barca (Santiago Island), is it possible to find small patches of what the pristine savannas of *F. albida* in Cabo Verde might have been (Neto et al., 2020). Thus, the continental African savannas where most of the *F. albida* are found, currently have no correspondence in Cabo Verde, having been occupied by large-scale plantations of *Prosopis juliflora* or other non-native trees. Modelling the bioclimatic niche of *F. albida* in Cabo Verde, and constraining the potential distribution with LULC data, allowed to identify the environmental and anthropogenic constraints that might have affected *F. albida* distribution in the past, and restrain its future range, thereby enabling the establishment of priorities in its management. The knowledge of the geographical range of Cabo Verde native trees provides essential baseline data toward achieving the goals of the Paris Agreement, which are on the country's agenda as a priority to conserve the fragile biodiversity of these islands. Thus, our methodological approach could be further extended to other native Cabo Verde trees in a straightforward manner. In this sense, we do hope that local authorities will benefit from this research, in the development of actions to promote reforestation, which should include native

trees that are highly adapted to the local climate and could contribute to further mitigate the effects of climate change.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found in the article/[Supplementary material](#).

Author contributions

All authors listed have made a substantial, direct, and intellectual contribution to the work and approved it for publication.

Funding

This research was funded by the Fundação para a Ciência e Tecnologia (FCT) and Aga Khan Development Network (AKDN) through the project CVAgrobiodiversity/333111699. DV was supported by FCT grant (SFRH/BD/135354/2017).

Acknowledgments

Special thanks to Silvia Catarino and Lara Dutra Silva for all their support in modelling. The authors thank the herbarium curator (Maria Cristina Duarte, FCUL) of the LISC Herbarium (IICT/ULisboa). We thank to Fundação para a Ciência e Tecnologia (FCT) and Aga Khan Development Network (AKDN) under the project CVAgrobiodiversity. Also to the research unit UID/AGR/04129/2020 (LEAF) and CIBIO-Açores funded by Portuguese National Funds through FCT, Portugal.

Conflict of interest

The authors declare no conflict of interest. The funders had no role in the design of the study; in the collection, analyses, or

interpretation of data; in the writing of the manuscript, or in the decision to publish the results.

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Supplementary material

The Supplementary material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fevo.2023.1057852/full#supplementary-material>

References

- Abdella, M., Nigatu, L., and Akuma, A. (2020). Impact of parkland trees (*Faidherbia albida* Delile and *Cordia africana* Lam) on selected soil properties and sorghum yield in Eastern Oromia, Ethiopia. *Agric. Forest. Fisheries* 9, 54–73. doi: 10.11648/j.aff.20200903.13
- Adjonou, K., Abotsi, K. E., Segla, K. N., Rabiou, H., Houetcheqnon, T., Sourou, K. N. B., et al. (2020). Vulnerability of African Rosewood (*Pterocarpus erinaceus*, Fabaceae) natural stands to climate change and implications for silviculture in West Africa. *Heliyon* 6:e04031. doi: 10.1016/j.heliyon.2020.e04031
- Allouche, O., Tsoar, A., and Kadmon, R. (2006). Assessing the accuracy of species distribution models: prevalence, kappa and the true skill statistic (TSS). *J. Appl. Ecol.* 43, 1223–1232. doi: 10.1111/j.1365-2664.2006.01214.x
- Araújo, M., Pearson, R., Thuiller, W., and Erhard, M. (2005). Validation of species-climate impact models under climate change. *Glob. Chang. Biol.* 11, 1504–1513. doi: 10.1111/j.1365-2486.2005.001000.x
- Arévalo, J. R., Delgado, J. D., Otto, R., Naranjo, A., Salas, M., and Fernández-Palacios, J. M. (2005). Distribution of alien vs. native plant species in roadside communities along an altitudinal gradient in Tenerife and Gran Canaria (Canary Islands). *Perspect. Plant Ecol. Evol. Syst.* 7, 185–202. doi: 10.1016/j.ppees.2005.09.003
- Atanasso, J. A., Salako, K. V., Mensah, S., Tohou, R. J., Djossa, B. A., Kakai, R. G., et al. (2021). Spatial distribution patterns of *Azalia africana* (Fabaceae – Detarioideae) in a tropical savanna of Benin: implications for management. *Plant Ecol. Evol.* 154, 362–375. doi: 10.5091/PLECEVO.2021.1713
- Barbet-Massin, M., Jiguet, F., Albert, C. H., and Thuiller, W. (2012). Selecting pseudo-absences for species distribution models: How, where and how many? *Methods Ecol. Evol.* 3, 327–338. doi: 10.1111/j.2041-210X.2011.00172.x
- Barnes, R., and Fagg, C. (2003). *Faidherbia albida monograph and annotated bibliography*. Oxford: Oxford Tropical Forestry Papers, 289.
- Booth, T. H. (2018). Species distribution modelling tools and databases to assist managing forests under climate change. *For. Ecol. Manag.* 430, 196–203. doi: 10.1016/j.foreco.2018.08.019
- Boyce, M. S., Vernier, P. R., Nielsen, S. E., and Schmiegelow, F. K. A. (2002). Evaluating resource selection functions. *Ecol. Model.* 157, 281–300. doi: 10.1016/S0304-3800(02)00200-4
- Breiman, L. (2001). *Random forests*. Berkeley: University of California.
- Casajus, N., Perie, C., Casajus, N., Périé, C., Logan, T., Lambert, M., et al. (2016). An objective approach to select climate scenarios when projecting species distribution under climate change: an objective approach to select climate scenarios when projecting species distribution under climate change. *PLoS One* 11:e0152495. doi: 10.5061/dryad.1sf74
- Castilla-Beltrán, A., de Nascimento, L., Fernández-Palacios, J. M., Fonville, T., Whittaker, R. J., Edwards, M., et al. (2019). Late Holocene environmental change and the anthropization of the highlands of Santo Antão Island, Cabo Verde. *Palaeogeogr. Palaeoclimatol. Palaeoecol.* 524, 101–117. doi: 10.1016/j.palaeo.2019.03.033
- Castilla-Beltrán, A., de Nascimento, L., Fernández-Palacios, J. M., Whittaker, R. J., Romeiras, M. M., Cundy, A. B., et al. (2021). Effects of Holocene climate change, volcanism and mass migration on the ecosystem of a small, dry island (Brava, Cabo Verde). *J. Biogeogr.* 48, 1392–1405. doi: 10.1111/jbi.14084
- Catarino, S., Romeiras, M. M., Pereira, J. M., and Figueira, R. (2021). Assessing the conservation of Miombo timber species through an integrated index of anthropogenic and climatic threats. *Ecol. Evol.* 11, 9332–9348. doi: 10.1002/ece3.7717
- CCKP (2019). World Bank climate change knowledge portal | for global climate data and information! Available at: <https://climateknowledgeportal.worldbank.org/country/cabo-verde> (Accessed August 1, 2020).
- Chang, X.-Y., Chen, B.-M., Liu, G., Zhou, T., Jia, X.-R., and Peng, S.-L. (2015). Effects of climate change on plant population growth rate and community composition change. *PLoS One* 10:e0126228. doi: 10.1371/journal.pone.0126228
- Chaturvedi, R. K., Joshi, J., Jayaraman, M., Bala, G., and Ravindranath, N. H. (2012). Multi-model climate change projections for India under representative concentration pathways. *Curr. Sci.* 103, 791–802. doi: 10.2307/24088836
- Chen, X., and Leites, L. (2020). The importance of land-use legacies for modeling present-day species distributions. *Landsc. Ecol.* 35, 2759–2775. doi: 10.1007/S10980-020-01119-0
- Chin, N. M., Hamzat, S., Oyinlola, E. Y., Idris, S., and Yusuf, S. A. (2022). Assessment of nitrogen mineralization of *Faidherbia albida* and its effect on maize (*Zea mays* L.) growth on an alfisol in Samaru, Northern Guinea Savanna of Nigeria. *J. Agric. Environ.* 18, 167–181.
- Cienciala, E., Centeio, A., and Blazek, P., Cruz Gomes Soares, M. da, & Russ, R. (2013). Estimation of stem and tree level biomass models for *Prosopis juliflora pallida* applicable to multi-stemmed tree species. *Trees - Struct. Funct.*, 27, 1061–1070. doi: 10.1007/s00468-013-0857-1
- CILSS (2016). Landscapes of West Africa—A window on a changing world. Comité Permanent Inter-états de Lutte contre la Sécheresse dans le Sahel.
- Costa, H., Medeiros, V., Azevedo, E. B., and Silva, L. (2013). Evaluating ecological-niche factor analysis as a modelling tool for environmental weed management in island systems. *Weed Res.* 53, 221–230. doi: 10.1111/wre.12017
- Deblauwe, V., Droissart, V., Bose, R., Sonké, B., Blach-Overgaard, A., Svenning, J. C., et al. (2016). Remotely sensed temperature and precipitation data improve species distribution modelling in the tropics. *Glob. Ecol. Biogeogr.* 25, 443–454. doi: 10.1111/geb.12426
- Dest, K. N., Lisanenwork, N., and Muktar, M. (2018). Physico-chemical properties of soil under the canopies of *Faidherbia albida* (Delile) A. Chev and *Acacia tortilis* (Forssk.) Hayen in park land agroforestry system in Central Rift Valley, Ethiopia. *J. Hortic. Forest.* 10, 1–8. doi: 10.5897/JHF2016.0491
- Dilla, A. M., Smethurst, P. J., Barry, K., and Parsons, D. (2019b). Preliminary estimate of carbon sequestration potential of *Faidherbia albida* (Delile) A. Chev in an agroforestry parkland in the Central Rift Valley of Ethiopia. *Forest. Trees Livelihoods* 28, 79–89. doi: 10.1080/14728028.2018.1564146
- Dilla, A. M., Smethurst, P. J., Barry, K., Parsons, D., and Denboba, M. A. (2019a). Tree pruning, zone and fertiliser interactions determine maize productivity in the *Faidherbia albida* (Delile) A. Chev parkland agroforestry system of Ethiopia. *Agrofor. Syst.* 93, 1897–1907. doi: 10.1007/s10457-018-0304-9
- Dobson, A. J., and Barnett, A. G. (2018). *An introduction to generalized linear models*. 4th Edn. New York: Chapman and Hall/CRC.
- Duarte, M. C., Gomes, I., Catarino, S., Brilhante, M., Gomes, S., Rendall, A., et al. (2022). Diversity of useful plants in Cabo Verde Islands: a biogeographic and conservation perspective. *Plan. Theory* 11:1313. doi: 10.3390/plants11101313
- Duarte, M. C., Rego, F., Romeiras, M. M., and Moreira, I. (2008). Plant species richness in the Cape Verde Islands eco-geographical determinants. *Biodivers. Conserv.* 17, 453–466. doi: 10.1007/s10531-007-9226-y
- Elith, J., H. Graham, C., P. Anderson, R., Dudik, M., Ferrier, S., Guisan, A., et al. (2006). Novel methods improve prediction of species' distributions from occurrence data. *Ecography* 29, 129–151. doi: 10.1111/j.2006.0906-7590.04596.x
- Emptage, M., and Dobson, A. J. (2008). *An Introduction to generalized linear models*. Florida: Chapman & Hall/CRC.
- Ereso, T. (2019). The role of *Faidherbia albida* tree species in parkland agroforestry and its management in Ethiopia. *J. Hortic. Forest.* 11, 42–47. doi: 10.5897/JHF2018.0570
- Fahrig, L. (2003). Effects of habitat fragmentation on biodiversity. *Annu. Rev. Ecol. Syst.* 34, 487–515. doi: 10.1146/annurev.ecolsys.34.011802.132419
- Feeley, K. J., Bravo-Avila, C., Fadrique, B., Perez, T. M., and Zuleta, D. (2020). Climate-driven changes in the composition of New World plant communities. *Nat. Clim. Chang.* 10, 965–970. doi: 10.1038/s41558-020-0873-2
- Ferrari, A., Dai, J., Bai, Y., and Alatalo, J. M. (2019). Redefining the climate niche of plant species: A novel approach for realistic predictions of species distribution under climate change. *Sci. Total Environ.* 671, 1086–1093. doi: 10.1016/j.scitotenv.2019.03.353
- Fick, S., and Hijmans, R. (2017). Worldclim 2: new 1-km spatial resolution climate surfaces for global land areas. *Int. J. Climatol.* 37, 4302–4315. doi: 10.1002/joc.5086

- Florencio, M., Patiño, J., Nogué, S., Traveset, A., Borges, P. A. V., Schaefer, H., et al. (2021). Macaronesia as a fruitful arena for ecology, evolution, and conservation biology. *Front. Ecol. Evol.* 9:718169. doi: 10.3389/fevo.2021.718169
- García-Valdés, R., Svenning, J. C., Zavala, M. A., Purves, D. W., and Araújo, M. B. (2015). Evaluating the combined effects of climate and land-use change on tree species distributions. *J. Appl. Ecol.* 52, 902–912. doi: 10.1111/1365-2664.12453
- GBIF.org (2021). GBIF occurrence download: *Faidherbia albida* (Delile) A.Chev.
- Gorel, A. P., Dumnil, J., Doucet, J. L., and Fayolle, A. (2019). Ecological niche divergence associated with species and populations differentiation in *Erythrophleum* (Fabaceae, Caesalpinioideae). *Plant Ecol. Evol.* 152, 41–52. doi: 10.5091/plevevo.2019.1543
- Guisan, A., Edwards, T. C., and Hastie, T. (2002). Generalized linear and generalized additive models in studies of species distributions: setting the scene. *Ecol. Model.* 157, 89–100. doi: 10.1016/S0304-3800(02)00204-1
- Hadgu, K. M., Kooistra, L., Rossing, W. A. H., and van Bruggen, A. H. C. (2009). Assessing the effect of *Faidherbia albida* based land use systems on barley yield at field and regional scale in the highlands of Tigray, Northern Ethiopia. *Food Secur.* 1, 337–350. doi: 10.1007/s12571-009-0030-2
- Haskett, J. D., Simane, B., and Smith, C. (2019). Energy and climate change mitigation benefits of *Faidherbia albida* agroforestry in Ethiopia. *Front. Environ. Sci.* 7, 1–15. doi: 10.3389/fevs.2019.00146
- Hegel, T. M., Cushman, S. A., Evans, J., and Huettmann, F. (2010). “Current state of the art for statistical modelling of species distributions” in *Spatial complexity, informatics, and wildlife conservation* (Japan: Springer), 273–311.
- Hijmans, R. J., Eten, J. van, Sumner, M., Cheng, J., and Baston, D. (2020). CRAN - Package raster (3.4-5). Available at: <https://cran.r-project.org/web/packages/raster/index.html> (Accessed December 24, 2020).
- IPCC (2018a). Global warming of 1.5°C an IPCC special report on the impacts of global warming of 1.5°C above pre-industrial levels and related global greenhouse gas emission pathways, in the context of strengthening the global response to the threat of climate change, sustainable development, and efforts to eradicate poverty headline statements from the summary for policymakers* understanding global warming of 1.5°C. Available at: https://www.ipcc.ch/site/assets/uploads/sites/2/2018/07/sr15_headline_statements.pdf
- IPCC (2018b). Headline statements from the summary for policymakers. Global warming of 1.5°C. an IPCC special report on the impacts of global warming of 1.5°C above pre-industrial levels and related global greenhouse gas emission pathways, in the context of strengthening the global response to the threat of climate change, 1–2.
- IPCC (2022). *Climate change 2022: Impacts, adaptation, and vulnerability*. Switzerland: Cambridge University Press.
- Lindskog, P., and Delaite, B. (1996). Degrading land: an environmental history perspective of the Cape Verde Islands. *Environ. Hist.* 2, 271–290. doi: 10.3197/096734096779522266
- Lu, T., Brandt, M., Tong, X., Hiernaux, P., Leroux, L., Ndao, B., et al. (2022). Mapping the abundance of *Faidherbia albida* trees in Senegal with Sentinel-2 time series data. Remote Sensing, Submitted.
- Luckow, M., Miller, J. T., Murphy, D. J., and Livshultz, T. (2003). “A phylogenetic analysis of the Mimosoideae (Leguminosae) based on chloroplast DNA sequence data” in *Advances in legume systematics, part 10, higher level systematics*. eds. B. B. Klitgaard and A. Bruneau (Kew, UK: Royal Botanic Gardens), 197–220.
- Marmion, M., Parviainen, M., Luoto, M., Heikkinen, R. K., and Thuiller, W. (2009). Evaluation of consensus methods in predictive species distribution modelling. *Divers. Distrib.* 15, 59–69. doi: 10.1111/j.1472-4642.2008.00491.x
- Monteiro, F., Fortes, A., Ferreira, V., Essoh, A. P., Gomes, I., and Correia, M. A., & Romeiras, M. M., (2020). Status and trends in Cabo Verde agriculture. *Agronomy*, 10:74. doi: 10.3390/agronomy10010074
- Ndao, B., Leroux, L., Hema, A., Diouf, A. A., Bégué, A., and Sambou, B. (2022). Tree species diversity analysis using species distribution models: A *Faidherbia albida* parkland case study in Senegal. *Ecol. Indic.* 144:109443. doi: 10.1016/j.ecolind.2022.109443
- Neto, C., Costa, J. C., Figueiredo, A., Capelo, J., Gomes, I., Vitoria, S., et al. (2020). The role of climate and topography in shaping the diversity of plant communities in Cabo Verde Islands. *Diversity* 12:80. doi: 10.3390/d12020080
- Newman, D., Nimon, K., Walker, D. A., Lieberman, M. G., and Morris Florida, J. D. (2010). Comparing OLS and HLM models and the questions they answer: potential concerns for type VI errors. *Multiple Linear Reg. Viewpoints* 36:33.
- Norder, S. J., Lima, R. F., De Nascimento, L., Lim, J. Y., Fernandez-Palacios, J. M., Romeiras, M. M., et al. (2020). Global change in microcosms: environmental and societal predictors of land cover change on the Atlantic Ocean Islands. *Anthropocene* 30:100242. doi: 10.1016/j.anecene.2020.100242
- Noulékoun, F., Chude, S., Zenebe, A., and Birhane, E. (2017). Climate change impacts on *Faidherbia albida* (Delile) A. Chev. Distribution in dry lands of Ethiopia. *Afr. J. Ecol.* 55, 233–243. doi: 10.1111/aje.12345
- Nyairo, R., and Machimura, T. (2020). Potential effects of climate and human influence changes on range and diversity of nine Fabaceae species and implications for nature’s contribution to people in Kenya. *Climate* 8, 1–19. doi: 10.3390/cli8100109
- Olehowski, C., Naumann, S., Fischer, D., and Siegmund, A. (2008). Geo-ecological spatial pattern analysis of the island of Fogo (Cape Verde). *Glob. Planet. Chang.* 64, 188–197. doi: 10.1016/j.gloplacha.2008.09.006
- Parmesan, C., and Hanley, M. E. (2015). Plants and climate change: complexities and surprises. *Ann. Bot.* 116, 849–864. doi: 10.1093/aob/mcv169
- Pearce, J. L., and Boyce, M. S. (2006). Modelling distribution and abundance with presence-only data. *J. Appl. Ecol.* 43, 405–412. doi: 10.1111/j.1365-2664.2005.01112.x
- Perrin, G., Rapinel, S., Hubert-Moy, L., and Bioret, F. (2020). Bioclimatic dataset of metropolitan France under current conditions derived from the WorldClim model. *Data Brief* 31:105815. doi: 10.1016/j.dib.2020.105815
- POWO - Plants of the World Online (2021). Available at: <http://powo.science.kew.org> [Accessed September 15, 2021].
- Prasad, A. M., Iverson, L. R., and Liaw, A. (2006). Newer classification and regression tree techniques: bagging and random forests for ecological prediction. *Ecosystems* 9, 181–199. doi: 10.1007/s10021-005-0054-1
- QGIS Development Team (2021). QGIS geographic information system. Open Source Geospatial Foundation Project. Available at: <http://qgis.osgeo.org> [Accessed April 17, 2021].
- R Development Core Team (2020). R: A language and environment for statistical computing. R Foundation for Statistical Computing. Available at: <http://www.r-project.org/index.html> (Accessed December 26, 2020).
- Ren-Yan, D., Xiao-Qun, K., Min-Yi, H., Wei-Yi, F., and Zhi-Gao, W. (2014). The predictive performance and stability of six species distribution models. *PLoS One* 9:e112764. doi: 10.1371/journal.pone.0112764
- Rigby, R. A., Stasinopoulos, D. M., and Lane, P. W. (2005). Generalized additive models for location, scale and shape. *Journal of the Royal Statistical Society. Series C. Appl. Stat.* 54, 507–554. doi: 10.1111/j.1467-9876.2005.00510.x
- Rivas-Martínez, S., Lousã, M., Costa, J. C., and Duarte, M. C. (2017). Geobotanical survey of Cabo Verde Islands (West Africa). *Int. J. Geobot. Res.* 7, 1–103.
- Romeiras, M. M., Carine, M., Duarte, M. C., Catarino, S., Dias, F. S., and Borda-de-Água, L. (2020). Bayesian methods to analyze historical collections in time and space: a case study using Cabo Verde Endemic Flora. *Front. Plant Sci.* 11:278. doi: 10.3389/fpls.2020.00278
- Romeiras, M. M., Catarino, S., Gomes, I., Fernandes, C., Costa, J. C., Caujapé-Castells, J., et al. (2016). IUCN red list assessment of the Cape Verde endemic flora: towards a global strategy for plant conservation in Macaronesia. *Bot. J. Linn. Soc.* 180, 413–425. doi: 10.1111/boj.12370
- Romeiras, M. M., Duarte, M. C., Santos-Guerra, A., Carine, M. A., and Francisco-Ortega, J. (2014). Botanical exploration of the Cape Verde islands: from the pre-Linnaean records and collections to the late 18th century accounts and expeditions. *Taxon* 63, 625–640. doi: 10.12705/633.37
- Romeiras, M. M., Pena, A. R., Menezes, T., Vasconcelos, R., Monteiro, F., Paulo, O. S., et al. (2019). Shortcomings of phylogenetic studies on recent radiated insular groups: A meta-analysis using Cabo Verde biodiversity. *Int. J. Mol. Sci.* 20:2782. doi: 10.3390/ijms20112782
- Santini, L., Benítez-López, A., Maiorano, L., Čengić, M., and Huijbregts, M. A. J. (2021). Assessing the reliability of species distribution projections in climate change research. *Divers. Distrib.* 27, 1035–1050. doi: 10.1111/ddi.13252
- Sida, T. S., Baudron, F., Deme, D. A., Tolera, M., and Giller, K. E. (2018a). Excessive pruning and limited regeneration: are *Faidherbia albida* parklands heading for extinction in the central Rift Valley of Ethiopia? *Land Degrad. Dev.* 29, 1623–1633. doi: 10.1002/ldr.2959
- Sida, T. S., Baudron, F., Kim, H., and Giller, K. E. (2018b). Climate-smart agroforestry: *Faidherbia albida* trees buffer wheat against climatic extremes in the Central Rift Valley of Ethiopia. *Agric. For. Meteorol.* 248, 339–347. doi: 10.1016/j.agrformet.2017.10.013
- Sileshi, G. W., Teketay, D., Gebrekirstos, A., and Hadgu, K. (2020). “Sustainability of *Faidherbia albida*-based agroforestry in crop production and maintaining soil health” in *Agroforestry for degraded landscapes: Recent advances and emerging challenges*. eds. J. C. Dagar, S. R. Gupta and D. Teketay, vol. 2 (Singapore: Springer), 349–369.
- Silva, L., Bento Elias, R., and Silva, L. (2021). Modelling invasive alien plant distribution: A literature review of concepts and bibliometric analysis. *Environ. Model. Softw.* 145:105203. doi: 10.1016/j.envsoft.2021.105203
- Silva, L., Costa, H., de Azevedo, E. B., Medeiros, V., Alves, M., Elias, R. B., et al. (2017). Modelling native and invasive woody species: A comparison of ENFA and MaxEnt applied to the Azorean forest. *Springer Proc. Math. Stat.* 195, 415–444. doi: 10.1007/978-3-319-55236-1_20

- Silva, L., de Azevedo, E. B., Reis, F. V., Elias, R. B., and Silva, L. (2019). Limitations of species distribution models based on available climate change data: A case study in the Azorean forest. *Forests* 10, 1–29. doi: 10.3390/f10070575
- Singer, A., Millat, G., Staneva, J., and Kröncke, I. (2017). Modelling benthic macrofauna and seagrass distribution patterns in a North Sea tidal basin in response to 2050 climatic and environmental scenarios. *Estuar. Coast. Shelf Sci.* 188, 99–108. doi: 10.1016/j.ecss.2017.02.003
- Sylla, M. B., Nikiema, P. M., Gibba, P., Kebe, I., and Klutse, N. A. B. (2016). “Climate change over West Africa: recent trends and future projections” in *Adaptation to climate change and variability in rural West Africa*. eds. J. A. Yaro and J. Hesselberg (Springer International Publishing), 25–40.
- Symon, C. (2013). Climate change: action, trends and implications for business: the IPCC’s fifth assessment report, working group I. Available at: <http://www.europeanclimate.org/documents/IPCCWebGuide.pdf>
- Thuiller, W. (2003). BIOMOD - optimizing predictions of species distributions and projecting potential future shifts under global change. *Glob. Chang. Biol.* 9, 1353–1362. doi: 10.1046/j.1365-2486.2003.00666.x
- Thuiller, W., Georges, D., Engler, R., and Breiner, F. (2016). Ensemble platform for species distribution modelling (3.3-7.1). Available at: https://r-forge.r-project.org/R/?group_id=302 (Accessed June 18, 2020).
- Thuiller, W., Georges, D., Engler, R., and Breiner, F. (2020). Package “biomod2” (3.4.6).
- Thuiller, W., Lavergne, S., Roquet, C., Boulangeat, I., Lafourcade, B., and Araujo, M. B. (2011). Consequences of climate change on the tree of life in Europe. *Nature* 470, 531–534. doi: 10.1038/nature09705
- USGS (2022). Land use and land cover trends in West Africa | West Africa. Landscapes of West Africa: A Window on a Changing World. Available at: <https://eros.usgs.gov/westafrica/land-cover/land-use-and-land-cover-trends-west-africa> (Accessed February 28, 2022).
- Vandenbeldt, R. J. (1992). “Faidherbia albida in the west African semi-arid tropics.” in *proceedings of a workshop, 22-26 Apr 1991, Niamey, Niger. (in En. Summaries in En, Fr, Es.) Patancheru, A.P. 502 324, India: International Crops Research Institute for the Semi-Arid Tropics; and Nairobi, Kenya: International Centre for Research in Agrofor. International Centre for Research in Agroforestry.*
- Varela, D., Monteiro, F., Vidigal, P., Silva, L., and Romeiras, M. M. (2020). Mechanisms implemented for the sustainable development of agriculture: an overview of Cabo Verde performance. *Sustainability* 12, 3–4. doi: 10.3390/su12145855
- Varela, D., Romeiras, M. M., and Silva, L. (2022). Implications of climate change on the distribution and conservation of Cabo Verde endemic trees. *Glob. Ecol. Conser.* 34:e02025. doi: 10.1016/j.gecco.2022.e02025
- Varela-Lopes, G. E., and Molion, L. C. B. (2014). Precipitation patterns in Cape Verde Islands: Santiago Island Case study. *Atmos. Climate Sci.* 4, 854–865. doi: 10.4236/acs.2014.45075
- Wang, J., Wang, Y., Feng, J., Chen, C., Chen, J., Long, T., et al. (2019). Differential responses to climate and land-use changes in threatened Chinese *Taxus* species. *Forests* 10:16. doi: 10.3390/f10090766
- Wooldridge, J. M. (2008). *Introductory econometrics: A modern approach. 4th Edn.* Mason: Thomson Southwestern Ohio.
- Yengwe, J., Amalia, O., Lungu, O. I., and De Neve, S. (2018a). Quantifying nutrient deposition and yield levels of maize (*Zea mays*) under *Faidherbia albida* agroforestry system in Zambia. *Eur. J. Agron.* 99, 148–155. doi: 10.1016/j.eja.2018.07.004
- Yengwe, J., Gebremikael, M. T., Buchan, D., Lungu, O., and De Neve, S. (2018b). Effects of *Faidherbia albida* canopy and leaf litter on soil microbial communities and nitrogen mineralization in selected Zambian soils. *Agrofor. Syst.* 92, 349–363. doi: 10.1007/s10457-016-0063-4
- Zhang, J., Nielsen, S. E., Chen, Y., Georges, D., Qin, Y., Wang, S. S., et al. (2017). Extinction risk of North American seed plants elevated by climate and land-use change. *J. Appl. Ecol.* 54, 303–312. doi: 10.1111/1365-2664.12701