



Impact of Climate Change on Priority Plant Species Conservation in The Forests Under ONAB (the National Wood Office of Benin) Management

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Abstract

The sustainable conservation of forest resources in a context of climate change and population growth would be compromised in their current form of exploitation by rural communities. The objective of this study is to assess the impact of climate change on the dynamics of habitats favorable to species of

conservation priority in the forests under ONAB management as these species are heavily used by the populations living along the shores of these forests. The study will consist of: *Khaya senegalensis*, *Azizelia africana*, *Khaya grandifoliola*, *Pterocarpus erinaceus*, *Anogeissus leiocarpa*, *Milicia excelsa*, *Albizia zygia*, *Vitex doniana*, *Antidesma laciniatum* and *Bombax costatum*. Techniques based on the principle of maximum entropy (Maxent) combined with GIS were used to project the favorable habitats of these ten species under current and future climatic conditions (Horizon 2050). Species occurrence data were collected and combined with bioclimatic data derived from the Worldclim database and the edaphic (soil) variable. Two climate models were used for future projections (CNRM-CM5, HadGEM-ES models) under the IPCC A2 scenario, and the partial ROC approach was used for the evaluation of the predictions of ecological niche models. Variables such as cec2 (cation exchange capacity, horizon 5-15cm), bio17 (precipitation of the driest quarter), bio12 (annual precipitation), bio3 (isothermality), bio6 (minimum temperature of the coldest month) and bio7 (annual thermal amplitude) were found to be the most relevant respectively for the distribution of *Khaya grandifoliola*, *Albizia zygia*, *Anogeissus leiocarpa*, *Antidesma laciniatum*, *Azizelia africana* and *Khaya senegalensis*. Under current conditions, only 7% of the Beninese territory would be very favorable to the conservation of *Khaya senegalensis* and the CNRM-CM5 model predicts an increase of 27.5% and 13.2% respectively of these very favorable and moderately favorable areas by 2050 through conversion of unfavorable areas (7.4%). On the other hand, this model predicts an opposite trend at the level of *Azizelia africana* where it predicts a decrease of 8.1% and 1.8% respectively of the very favorable and moderately favorable areas and an increase of 5.1% of the unfavorable areas. Ecological niche modeling has basically revealed the conversion of some currently unfavorable habitats into very favorable habitats for conservation (this is the case of *Khaya grandifoliola*, *Khaya senegalensis* and *Vitex doniana*) and the extension of some habitats unfavorable to conservation (*Anogeissus leiocarpa*, *Bombax costatum*, and *Pterocarpus erinaceus*) by 2050. This study provides scientific support for planning and is a decision support tool for the conservation of these species at the socio-economic level.

Keywords: Utilitarian species, ecological niche, classified forests, climate change, model

Introduction

Many species are becoming rare in their natural habitats because of the threats to biodiversity. Climate change is now recognized as one of the major threats to the survival of species and the integrity of ecosystems around the world. Knowledge of the specific properties of these changes, which may have

an impact on species or their habitats, is central to adaptation strategies (Heller and Zavaleta, 2009). Since climate fluctuations influence biodiversity and determine the spatial distribution of favorable geographical areas, climate change is a crucial environmental issue that needs to be integrated into conservation planning (Fandohan et al., 2013). In addition, many plant species are used by local populations for food, medicine, and also make a substantial contribution to household income (Assogbadjo et al., 2008; Vodouhê et al., 2009). In Africa, 25-42% of plant species could be threatened with extinction due to a loss of 81-97% of suitable habitats by 2085 (Boko et al., 2007). It is projected that 20-30% of plant and animal species will face a greater risk of extinction if global warming exceeds 1.5-2.5°C in Africa (Jovic et al., 2010; Parry et al., 2007).

Furthermore, much of the impressive biodiversity of sub-Saharan Africa, including Benin, remains one of the most vulnerable to the impacts of climate change (Parry et al., 2007). It is clear that fluctuations in climatic variables such as rainfall and temperature will affect biodiversity and the geographical distribution of species-friendly habitats (Parry et al., 2007). Several studies have revealed that due to the manifestations of climate change, many current crop varieties will need to be replaced in the future if cropping areas are to be maintained (Deryng et al., 2011; Duveiller et al., 2007; Jones et al., 2003). These effects are likely to be felt on species that are useful to the riparian populations of the forests under ONAB management. However, the use of ecological niche models can reveal the distribution of habitats favorable to species in the present as well as the future and is a powerful tool for understanding the current and future distribution of a species (Phillips et al., 2006). These models also allow better reasoning for the choice of areas where these species will be now and in the future in order to anticipate future changes in response to environmental disturbances (Idohou et al., 2016). However, climate change represents a threat that is fundamentally different from other threats because, in the past, ecological stress has never called into question the effectiveness of protected area networks in conserving representative samples of species (Baxter et al., 2005) or in covering habitats that are favorable to them. In short, there is considerable uncertainty (at the country and regional scales) about how these static areas will continue to conserve or provide favorable habitat for the species concerned as a result of climate change. Climate change is therefore a real challenge for the development of future policies and options in agriculture and conservation because of the uncertainties associated with it.

In Benin, 280 plant species are threatened with extinction and 90% have a high probability of extinction (Adomou, 2005). Of these, 19 are of regional conservation interest and 10 are on the IUCN Red List (Adomou, 2005). *Azefelia africana* and *Khaya senegalensis* are two of these threatened

species, already on the Red List of the International Union for the Conservation of Nature (IUCN), although they even appear to be critically endangered in Benin (Adomou et al., 2006). In addition to these two species, which are well known for their usefulness as a source of fodder for livestock and as timber highly prized in the international timber trade, other species such as *Daniellia oliveri* and *Anogeissus leiocarpa* are proving to be very vulnerable to anthropogenic pressures and are becoming increasingly rare (Houehanou et al., 2013). Exploitation is therefore being shifted to other plant species, among which *Detarium microcarpum*, *Prosopis africana* and *Burkea africana* can be cited (Houehanou et al., 2013). According to Djègo-Djossou (2003), the Central Nucleus of the Lama Classified Forest, located in southern Benin, constitutes the last great remnant of the natural dense forest. Despite the mono-species nature of the plantations of the Office National du Bois du Bénin (ONAB) most of the time, they also shelter species of interest to the local population. On the one hand, the local populations have an increased need to use the forest resources of these forests under ONAB management (classified forests of Bassila, Pénéssoulou, Atchéribé, Dogo-Kétou, Bonou, Lama, Agrimey, Djigbé). This is due to population growth, to the needs of plant organs in the pharmacopoeia, and to the accelerated and continuous degradation of the forests by the regular passage of vegetation fires and transhumant herds. On the other hand, we have the difficulties encountered in the protection policies of these classified forests and wildlife reserves, which constitute increasingly serious threats to the protection and survival of these important resources. The biodiversity of these forests is disappearing at an alarming rate. It is therefore important, in the face of this conservation problem, to ask to what extent the protected areas in Benin are effectively conserving biological diversity (Fandohan et al., 2013).

This study mainly aims at assessing the potential impact of climate change on the dynamics of habitats favorable to priority species used by the ONAB riparian populations. It thus contributes to the establishment of databases on the potential impact of climate change on certain species of interest according to the different scenarios available. Also, it could accelerate decision-making for adaptive measures. From a bioclimatic point of view, what is the extent of the areas favorable to the conservation of these primate species? With regard to climate projections, what is the potential effect of climate change on the extent of these areas and their geographical distribution by 2050?

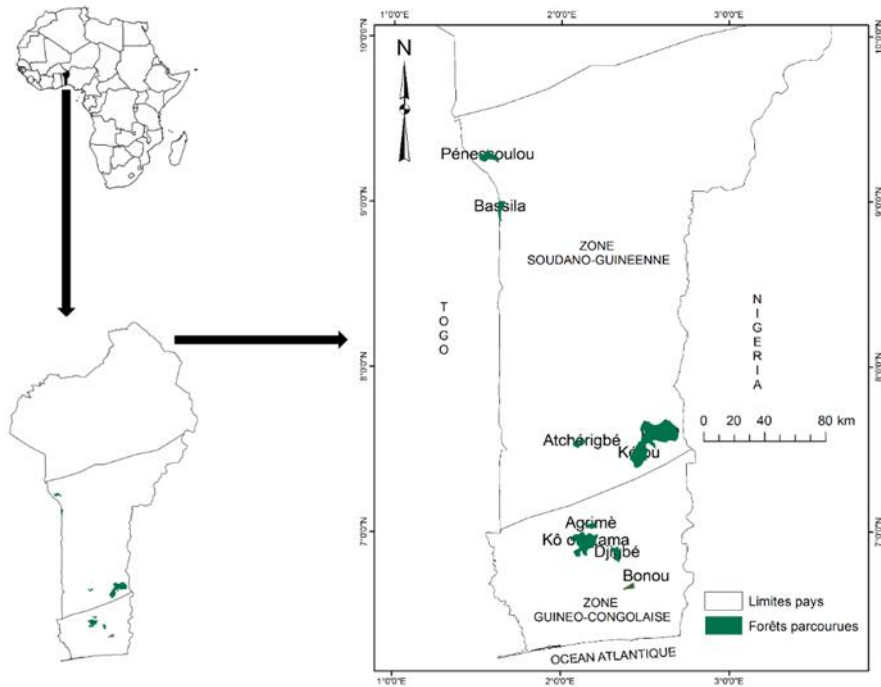
Materials and Methods

Study Environment

This study was conducted in the Republic of Benin (6°-12.25°N; 0.40°-3°E). The study covers two biogeographical zones of Benin: the

Sudano-Guinean zone ($7^{\circ} 30' N$ and $11^{\circ} 5' N$) and the Guineo-Congolese zone ($6^{\circ} 25' N$ and $7^{\circ} 30' N$) (Figure 1). The forests under ONAB management occupy from south to central Benin, two bio-geographical regions: the Guinean affinity zone and the Sudano-Guinean zone.

Figure 1: Location of plantations and forests under ONAB management by biogeographical zone in Benin



Source: Deguenon 2017

Rainfall distribution is bimodal in the Guinean zone (April to June and September to November), with an average annual rainfall of 1200 mm. This zone has deep, low fertility ferralitic soils (700,000 ha), alluvial soils and heavy clay soils (360,000 ha) located in the valleys of the Mono, Niger, Couffo and Ouémé rivers including the Lama depression (Table 1). These soils are rich in clay, humus and inorganic elements (Sala et al., 2000).

Beyond the Guinean zone, the Sudano-Guinean zone is characterized by a unimodal rainfall distribution (May-October) with a mean annual rainfall of more than 900 mm. In this zone, infertile mineral soils and ferruginous soils are encountered (Table 1).

Table 1: Different climatic zones in Benin associated with climatic data and soil properties

Zones	Rainfall (mm)	RainySeasonP eriods	Temperature (°c)	Relative Humidity (%)	Soil Types
Sudano-Guinean	900-1110	May- October	25-29	31-98	Infertile mineral soils, ferruginous soil
Guinean	1200 (annual average)	April-June and September- November	25-29	69-97	Soils ferralitic, alluvial and heavy clay soils

Data Collection

The geographical coordinates (longitude and latitude) of the presence of priority species were collected from fieldwork in protected areas and agroforestry systems in Benin. To maximize the accuracy of the modeling results, the presence coordinates of these species were completed by exploring online biodiversity databases such as the GBIF (Global Biodiversity Information Facility: www.gbif.org) and published articles on the species. Current and future climate data for the global area under consideration were downloaded from the Worldclim website (www.worldclim.org). Nineteen bioclimatic variables (Busby, 1991) were generated from the raw climate data (precipitation and temperature) prepared and put into formats compatible with the MaxEnt program. For this purpose, the QGIS software was used.

Modeling and model validation

The maximum entropy approach (Maxent, version 3.3.3 k (Phillips et al., 2006)) was used to model the ecological niche of the species. This approach is based on probability density estimation (where occurrence data are assumed to be drawn from a certain probability distribution over the study area). The Maxent models were developed using 10,000 background points. A maximum of 1000 iterations, a convergence threshold of 0.00001, and a random 75% of the data points are set aside for the intrinsic test (Fielding and Bell, 1997; Guisan and Zimmermann, 2000). The interest of this method for this study is that it combines the observed occurrence data of a given species with the current bioclimatic characteristics at the observation points to generate: a global map of the potential habitats of the species in the area under consideration; a global map of the future distribution of these favorable habitats in relation to the climatic projections made for the area under study. Due to collinearity and dependence of climate dimensions (Zuur et al., 2010), we examined the correlations between the variables to select those that are not closely correlated ($r < 0.85$). Next, a Jackknife test was performed on the environmental variables considered to determine which ones contribute most to the modeling. A total of 28 environmental variables were identified and

contribute to the different models according to the ecological preferences of the priority utility species, including 14 bioclimatic and 14 soil variables.

To evaluate the model, 25% of the observation points of each species were used to test the model and 75% of the points were used to calibrate the model. The area under the Receiver Operating Characteristic (ROC) curve, known as the Area Under Curve (AUC), is widely used to assess the accuracy of the prediction of distribution models. The database for each species was randomized five times in R software (3.4.0). A model is said to be of good quality if the AUC value is greater than 0.80 (Swets, 1988). However, some authors have begun to criticize the blind use of AUC as a measure of the level of precision in distribution models (Austin and Van Niel, 2011; Lobo et al., 2008). For this, new indices were proposed and proved to be effective in evaluating the performance of the model. Among these new approaches, we have the partial ROC approach (Wisz et al., 2008) which was used within the framework of this study. This new approach provides a firmer basis for evaluating the predictions of ecological niche models. ROC analysis is a method for assessing the specificity (absence of commission error) and sensitivity (absence of omission error) of a diagnostic test (Fielding and Bell, 1997). We calculated the partial AUCs using a program based on the trapezoidal method (Barve et al., 2011). Partial AUC values were presented as a ratio of AUC (with x-axis modification, for traditional applications) to AUC zero expectation (which, unlike traditional approaches ROC is not equal to 0.5) and are variable (Wisz et al., 2008). For each species, we use the sort with the greatest partial AUC value. Significant AUC statistics rating (against zero expectations) will be achieved by resampling 50% of the test points with the replacement 1000 times from the global pool of test data. A good model should identify regions of the high probability that cover most records of presence, and areas of low probability should contain few or no points of presence (Saupe et al., 2012). The available documentation for each priority species was used to determine whether the high probability area models corresponded to areas known to contain the species.

Results

Contribution of Variables and Model Validation

The below table 2 reports on the variables that have contributed to the production of the 10 priority species models for conservation in the classified forests under ONAB management. Overall, species respond differently to environmental conditions. The variables most frequently used in the models produced are bio2 (mean diurnal amplitude), bio3 (isothermality), bio12 (annual precipitation), bio14 (precipitation in the driest month), bio17 (precipitation in the driest quarter), and cec5 (cation exchange capacity, horizon 60-100cm) while the variables cec1 (cation exchange capacity,

horizon 0-5cm), clay1 (clay, horizon 0-5cm), oc1 (organic carbon, horizon 0-5cm), oc4 (organic carbon, horizon 30-60cm), bio13 (precipitation of the wettest month) and clay2 (clay, horizon 5-15cm) contribute the least to the current distribution of the 10 priority species. On the other hand, variables such as cec2 (cation exchange capacity, horizon 5-15cm), bio17 (precipitation of the driest quarter), bio12 (annual precipitation), bio3 (isothermality), bio6 (minimum temperature of the coldest), and bio7 (annual thermal amplitude) are shown to be the most relevant for the distribution of *Khaya grandifoliola*, *Albizia zygia*, *Anogeissus leiocarpa*, *Antidesma laciniatum*, *Azelia africana* and *Khaya senegalensis*, respectively. On the other hand, variables such as bio17 (precipitation of the driest quarter), bio2 (average diurnal amplitude), bio11 (average temperature of the coldest quarter), clay2 (clay, horizon 5-15cm), bio3 (isothermality) and bio15 (seasonal precipitation) are those which contribute the least to the distribution of areas respectively of *Antidesma laciniatum*, *Albizia zygia*, *Anogeissus leiocarpa*, *Khaya grandifoliola*, *Bombax costatum* and *Milicia excelsa*.

Table 2: Contribution of variables to the model (%)

Species	K.	A.	K.	P.	A.	M.	A.	V.	A.	B.
Variables	<i>senegalensis</i>	<i>africana</i>	<i>grandifoliola</i>	<i>erinaceus</i>	<i>leiocarpa</i>	<i>excelsa</i>	<i>zygia</i>	<i>doniana</i>	<i>laciniatum</i>	<i>costatum</i>
bio1	-	-	-	-	16.3	-	-	10.2	-	-
bio2	-	22.8	7.9	12.8	-	25.4	1.8	-	-	9.7
bio3	21.3	-	-	-	-	9.2	-	-	48.2	5.6
bio4	-	15.8	-	-	-	-	5.5	-	-	8.1
bio5	-	-	-	-	-	-	15	15.4	-	-
bio6	-	27.9	-	23.8	-	-	-	-	-	-
bio7	25.7	-	-	24.5	-	-	-	17	-	-
bio10	-	9.8	-	-	-	-	-	10.8	-	-
bio11	-	-	8.3	19.2	4	-	-	-	-	-
bio12	15.6	-	-	11	42	-	-	-	19.9	-
bio13	-	-	-	-	-	-	5.6	14.7	-	-
bio14	-	-	15.2	-	-	16.8	-	-	9.1	15.2
bio15	-	-	-	-	-	6.8	-	-	18.1	-
bio16	10.7	11.9	-	8.6	-	-	-	-	-	-
bio17	-	-	-	-	-	27.1	58.8	-	0.1	43.9
cec1	-	-	13.4	-	-	-	-	-	-	-
cec2	-	-	50.9	-	24.2	-	-	-	-	-
cec5	-	11.8	-	-	9	14.7	-	-	4.6	-
clay1	-	-	-	-	-	-	-	31.9	-	-
clay2	15	-	4.3	-	-	-	-	-	-	-
oc1	-	-	-	-	-	-	-	-	-	17.5
oc2	11.6	-	-	-	-	-	13.4	-	-	-
oc4	-	-	-	-	4.5	-	-	-	-	-

bio1 (mean annual temperature), bio2 (mean diurnal amplitude), bio3 (isothermality), bio4 (seasonality), bio5 (maximum temperature of the hottest month), bio6 (minimum temperature of the coldest month), bio7 (thermal amplitude annual), bio10 (mean temperature of the warmest quarter), bio11 (mean temperature of the coldest quarter), bio12 (annual precipitation), bio13 (precipitation of the wettest month), bio14 (precipitation of the driest month) bio15 (seasonal precipitation); bio16 (precipitation from the wettest quarter), bio17 (precipitation from the driest quarter), cec1 (cation exchange capacity, horizon 0-5cm), cec2 (cation exchange capacity, horizon 5-15cm), cec5 (cation exchange capacity, horizon 60-100cm), clay1 (clay, horizon 0-5cm), clay2 (clay, horizon 5-15cm), oc1 (organic carbon, horizon 0-5cm), oc2 (organic carbon, horizon 5 -15cm), oc4 (organic carbon, horizon 30-60cm).

Validation of Model Results

To evaluate the performance of the models produced, the values of the mean ratio of the AUC statistic (Area Under the ROC Curve) generated by the models were used (Figure 2). Table 4 and Figure 3 report the values of the mean ratio of the AUC statistic relating to the model of each priority species for conservation in classified forests under ONAB management.

Figure 2: Maps showing the variation of AUC ratios for 1000 replicates using the Partial ROC procedure for (a) *Khaya senegalensis*, (b) *Azelia africana*, (c) *Khaya grandifoliola* (d) *Pterocarpus erinaceus* (e) *Anogeissus leiocarpa*, (f) *Milicia excelsa*, (g) *Albizia zygia*, (h) *Vitexdoniana*, (i) *Antidesma laciniatum*, (j) *Bombax costatum*.

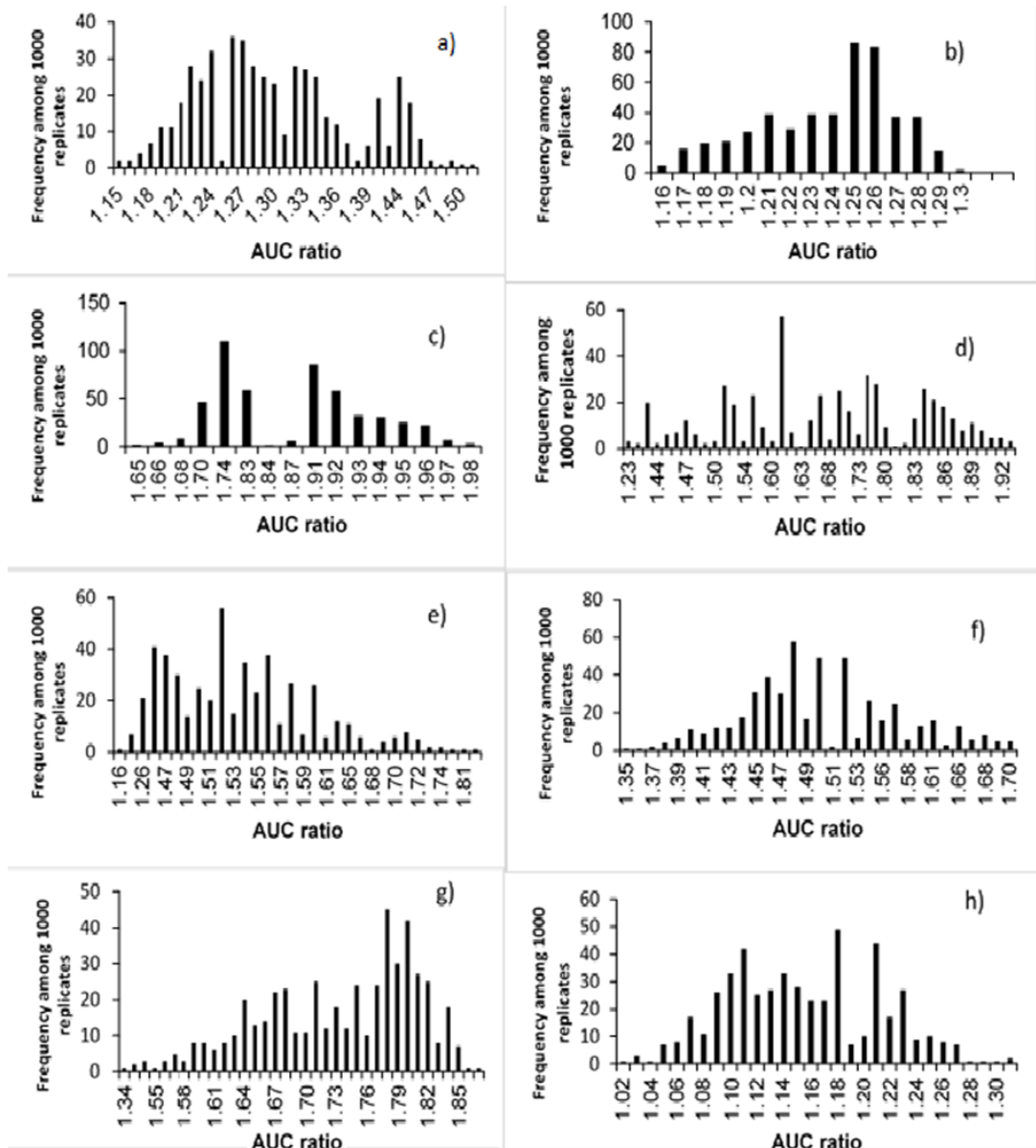


Table 3: Average AUC ratio values relating to the model of each priority species

Species	average ratio AUC	Variation
<i>Khaya senegalensis</i>	1,30	[1,15; 1,51]
<i>Azizelia africana</i>	1,24	[1,16; 1,30]
<i>Khaya grandifoliola</i>	1,85	[1,65; 1,98]
<i>Pterocarpus erinaceus</i>	1,68	[1,23; 1,93]
<i>Anogeissus leiocarpa</i>	1,52	[1,16; 1,81]
<i>Milicia excelsa</i>	1,51	[1,35; 1,70]
<i>Albizia zygia</i>	1,73	[1,34; 1,86]
<i>Vitex doniana</i>	1,16	[1,02; 1,30]
<i>Antidesma laciniatum</i>	1,52	[1,19; 1,99]
<i>Bombax costatum</i>	1,15	[1,05; 1,47]

From the analysis of Table 3, it emerges that the value of the mean ratio of the AUC statistic for each priority species is greater than 1. Therefore, the model is very efficient for each of these species. Thus, the current distribution areas predicted by the model for each species constitute areas of high probability where the species would be present or absent depending on the related ecological conditions.

Impact of Climate Change on the Extent of Habitats Favorable for the Conservation of the 10 Priority Species in Forests Classified Under ONAB Management

- **Khaya senegalensis**

According to the modeling results, around 7% of Benin's national territory (not including the islands on the Niger River) is currently very favorable for the conservation of *Khaya senegalensis* (Figure 3a1, Table 4). These habitats, which are very favorable to the conservation of the species, are essentially between 9.5 ° and 8.5 ° then 7.5 ° and 6.5 ° North latitude, which corresponds respectively to the dry subhumid and humid subhumid zone. Habitats unfavorable to the species are found throughout the country, mainly between 12 ° and 10 ° (arid zone, dry subhumid) and below 7.3 ° (humid subhumid zone).

According to the bioclimatic projections of the CNRM-CM5 model, *Khaya senegalensis* will experience an expansion of nearly 27% of habitats which are currently very favorable for conservation by 2050 (Table 4). This model also predicts shrinkage of nearly 7% of habitats that are unfavorable to the species, mainly by conversion of currently moderately favorable habitats.

The HadGEM-ES model the species will experience an expansion of nearly 8% of habitats that are currently very favourable for conservation by 2050 (Table 4). However, compared to the previous one, this model predicts the conversion of unfavorable habitats into moderately favorable habitats for the conservation of *Khaya senegalensis*, especially between 9 ° and 7.5° North latitude (Table 4, figure 3c1).

- **Afzelia Africana**

The modeling results reveal that around 16% of Benin's territory (not including the islands on the Niger River) are currently very favorable for the conservation of *Afzelia africana* (Figure 3a2, Table 4). These habitats, which are very favorable for the conservation of the species, are essentially between 11° and 7° North latitude, which corresponds to the dry subhumid and humid subhumid zone. Habitats unfavorable to the species are more below 8° (humid subhumid zone) and above 11° (arid zone).

According to the bioclimatic projections of the CNRM-CM5 model, a reduction of nearly 8% of highly favorable conservation areas is forecast for *Afzelia africana* by 2050 (Table 4). We will also observe an extension of the less favorable areas mainly by conversion of the currently moderately favorable areas (Figure 3b2).

The HadGEM-ES model gives opposite trends. Indeed, compared to the previous one, this model predicts an extension of areas very favorable to the conservation of the species, particularly between 9.2° and 8.5° North latitude by conversion of less favorable areas but also of relatively favorable area (Figure 3c2).

- **Khaya grandifoliola**

Modeling reveals that a small portion (about 2.2%) of Benin's territory (not including the islands on the Niger River) is currently very favorable for the conservation of *Khaya grandifoliola* (Figure 3a3, Table 4). These habitats, which are very favorable for the conservation of the species, are essentially between 9.3° and 9° then below 7.2° North latitude, which corresponds respectively to the dry subhumid and humid subhumid zone. Habitats unfavorable to the species are found throughout the country; mainly between 12.2° and 6.5° (arid zone, dry subhumid and humid subhumid zone).

According to the bioclimatic projections of the CNRM-CM5 model, we will observe a reduction in very favorable and moderately favorable areas, respectively, of nearly (18%) and 24% for *Khaya grandifoliola* by 2050 (Table 4).

The HadGEM-ES model predicts an opposite trend. We will observe an extension of very favorable areas (Table 4). It also predicts an extension of moderately favorable areas (Figure 3c3) essentially between 9.2° and 8° North latitude by conversion of areas not very favorable to the species.

- **Pterocarpus erinaceus**

The results of the modeling reveal that around 3% of Benin's territory (not including the islands on the Niger River) is currently very favorable to the conservation of *Pterocarpus erinaceus* (Figure 3a4, Table 4). These habitats, which are very favorable to the conservation of the species, are essentially between 9.2° and 8.8° North latitude, which corresponds to the dry

subhumid zone. Habitats unfavorable to the species are the most common across the country in all three climates.

According to the bioclimatic projections of the CNRM-CM5 model, a reduction in the very favorable and moderately favorable areas of *Pterocarpus erinaceus* will be observed (Figure 3b4, Table 4).

The HadGEM-ES model gives opposite trends. Indeed, compared to the previous one, this model predicts rather the extension of very favorable and moderately favorable areas (Figure 3c4) by a conversion of unfavorable habitats, particularly between 12° and 7° North latitude.

- **Anogeissus leiocarpa**

According to the results of the modeling, around 5% of the national territory (not including the islands on the Niger River) are currently very favorable for the conservation of *Anogeissus leiocarpa* (Figure 3a5, Table 4). These habitats, which are very favorable for the conservation of the species, are essentially between 12.2° and 10.8° then below 8° North latitude, which corresponds respectively to the arid zone and the humid subhumid zone. The most extensive habitats unfavorable to the species are throughout the country, essentially between 10.5° and 7.5° North latitude.

According to the bioclimatic projections of the CNRM-CM5 model, for *Anogeissus leiocarpa* a reduction of 42.33% of the very favorable areas will be observed (Table 4). Also, we will observe an extension of moderately favorable areas for the species, especially below 7° North latitude (humid subhumid zone; Figure 3b5).

The HadGEM-ES model for *Anogeissus leiocarpa* a reduction of 46.66% of the very favorable areas will be observed (Table 4). It also foresees the conversion of unfavorable habitats into moderately favorable habitats for the conservation of this species, particularly between 8.8° and 7° north latitude (Figure 3c5).

- **Milicia excels**

According to the modeling, around 8% of Benin's territory is currently very favorable for the conservation of *Milicia excelsa* (Figure 3a6, Table 4). These habitats, which are very favorable for the conservation of the species, are mainly included below 7.3° North latitude, which corresponds to the humid subhumid zone. Habitats unfavorable to the species are found further north of the country, mainly between 12° and 10° (arid zone, dry subhumid).

According to the bioclimatic projections of the CNRM-CM5 model, there will not be a great variation in *Milicia excelsa*. Nevertheless, 0.1% of very favorable areas will convert into moderately favorable areas by 2050 (Table 4).

The HadGEM-ES model gives opposite trends. In fact, compared to the previous one, this model predicts a higher extension of very favorable and

unfavorable areas by conversion of moderately favorable areas (table 4, figure 3c6).

- **Albizia zygia**

About 5% of the Beninese territory (not including the islands on the Niger River) is currently very favorable to the conservation of *Albizia zygia* (Figure 3a7, Table 4). These habitats, which are very favorable for the conservation of the species, are essentially below 8° North latitude, which corresponds to the humid subhumid zone. Habitats not very favorable to the species are the most extensive, mainly between 12° and 7.8° (arid, dry subhumid zone).

According to the bioclimatic projections of the CNRM-CM5 model, we will observe a conversion of nearly 40% of very favorable areas into moderately favorable areas by 2050 (Table 4, Figure 3b7).

The HadGEM-ES model gives similar trends. It predicts a conversion of 13.06% of very favorable areas to moderately favorable areas by 2050 (Table 4, Figure 3b7).

- **Vitex doniana**

According to the results of the modeling, around 13% of Benin's national territory (not including the islands on the Niger river) are currently very favorable to the conservation of *Vitex doniana* (Figure 3a8, Table 4). These habitats, which are very favorable for the conservation of the species, are mainly between 11.8° and 10° then below 8° North latitude, which corresponds respectively to the dry subhumid and humid subhumid zone. Habitats unfavorable to the species are mainly located in the arid and dry subhumid zone.

According to the bioclimatic projections of the CNRM-CM5 model, about 4% of areas very favorable to the species will be converted into less favorable areas by 2050 (Table 4, Figure 3b8).

The HadGEM-ES model gives opposite trends. In fact, compared to the previous one, this model predicts a large number (almost ten times more) the conversion of unfavorable habitats into very favorable and moderately favorable habitats for the conservation of *Vitex doniana*, in particular between 11.8° and 10° then in below ° North latitude (Table 4, figure 3c8).

- **Atidesma laciniatum**

According to the results, around 28% of Benin's territory (not including the islands on the Niger River) is currently very favorable for the conservation of *Antidesma laciniatum* (Figure 3a9, Table 4). These habitats, which are very favorable for the conservation of the species, are essentially between 10° and 7.5° North latitude, which corresponds to the dry subhumid and humid subhumid zone. Habitats unfavorable to the species are found further north of the country; mainly above 11 ° North latitude (arid zone).

According to the bioclimatic projections of the CNRM-CM5 model, we will observe a reduction of nearly (19.5%) of currently very favorable areas by 2050 (Table 4). These habitats will essentially be converted into habitats that are not very favorable to its conservation (Figure 3b9).

The HadGEM-ES model, compared to the previous one, predicts more the conversion of unfavorable habitats (around 5.22%) and moderately favorable (around 5.6%) to the conservation of *Antidesma laciniatum* into very favorable areas, particularly between 10° and 6, 8° North latitude (Table 4, figure 3c9).

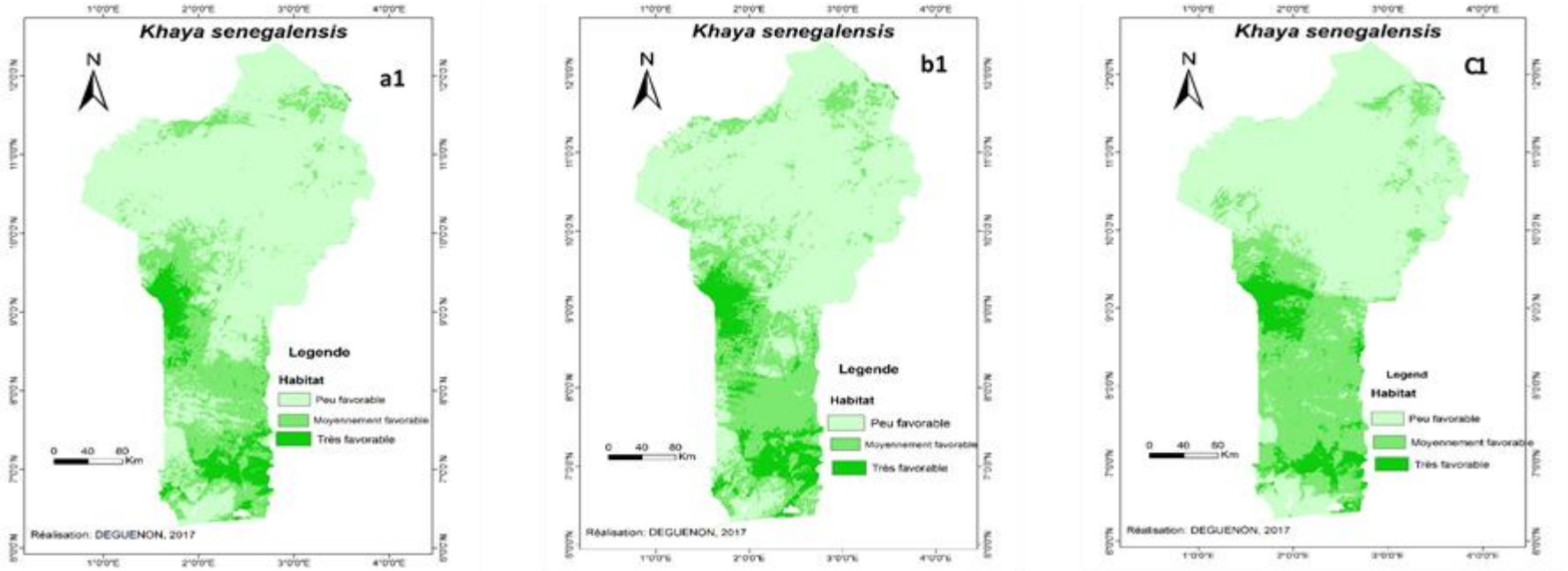
- **Bombax costatum**

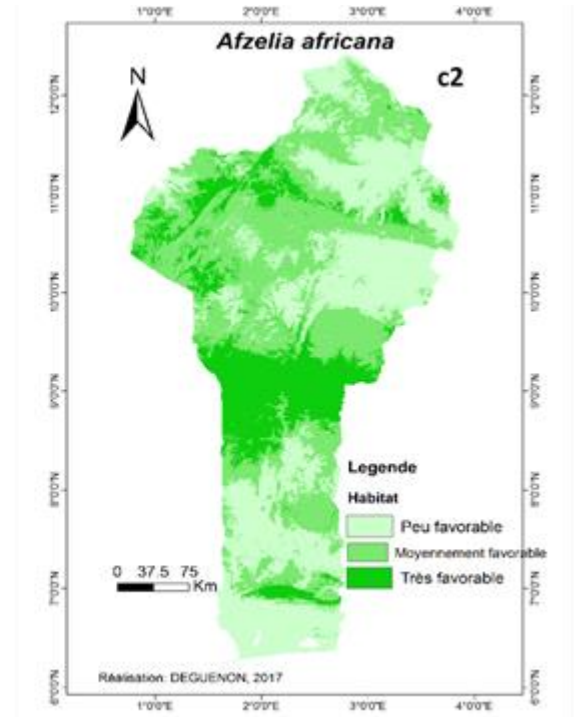
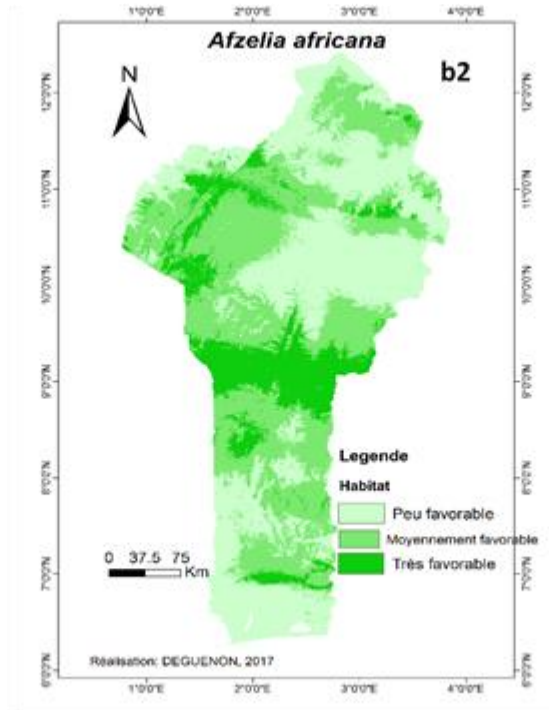
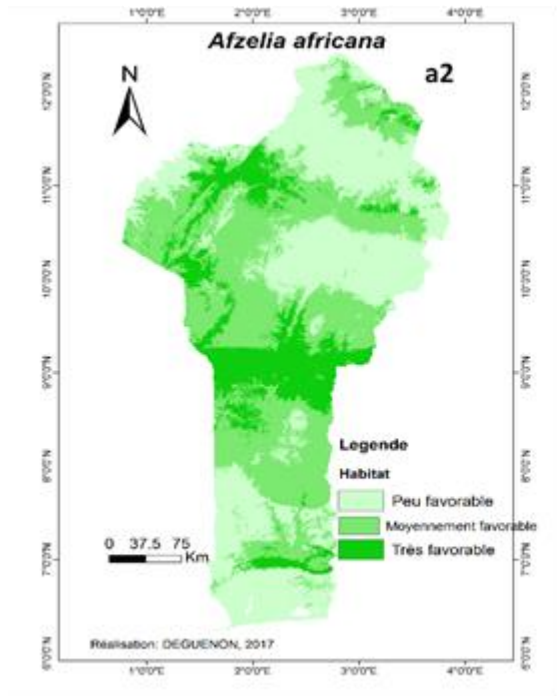
About 27% of the national territory (not including the islands on the Niger River) are currently very favorable to the conservation of *Bombax costatum* (Figure 3a10, table 4). These habitats, which are very favorable to the conservation of the species, are essentially between 12° and 8° North latitude, which corresponds to the dry and arid subhumid zone. Habitats unfavorable to the species are mainly below 8° (humid subhumid zone).

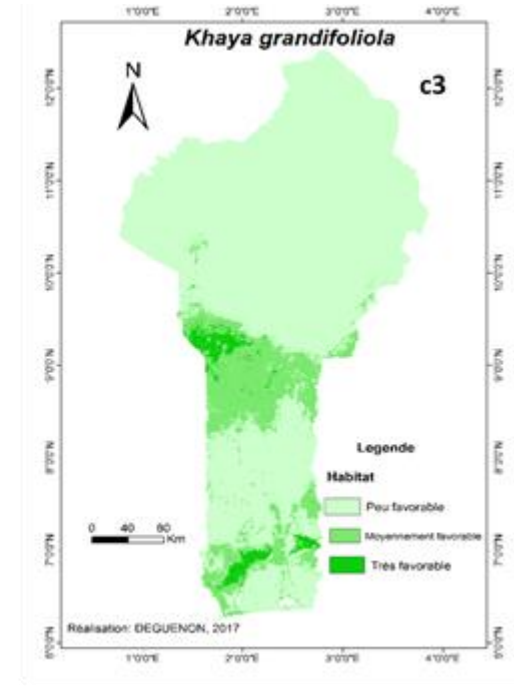
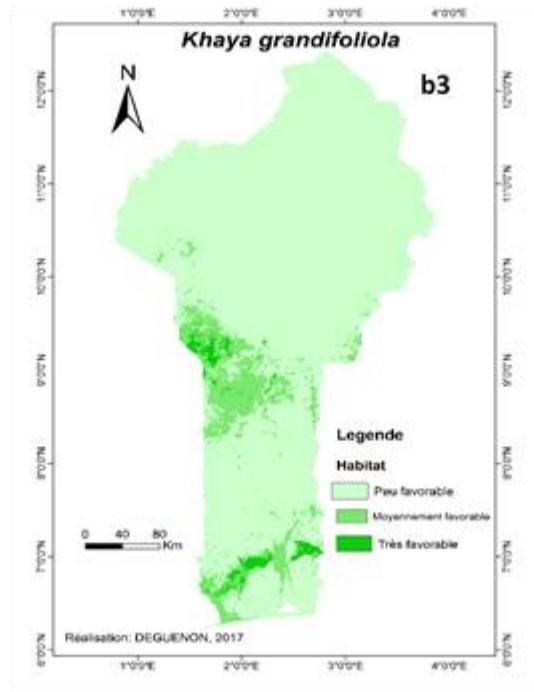
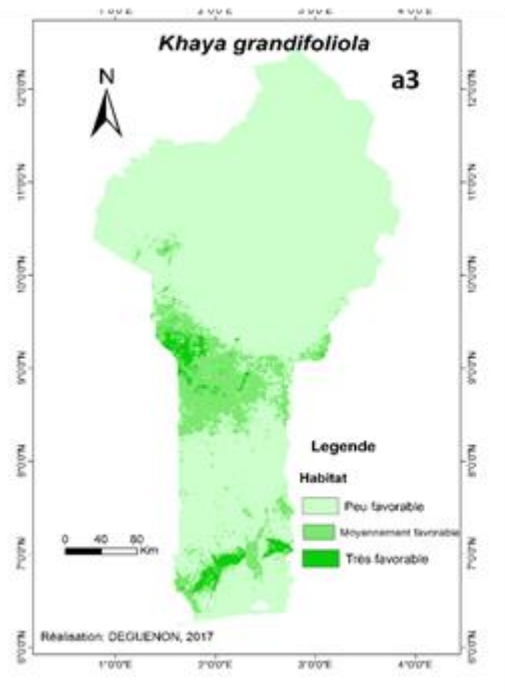
According to the bioclimatic projections of the CNRM-CM5 model, we will note an extension of the less favorable areas of *Bombax costatum*, in particular by converting the moderately favorable areas by 2050 (Table 4).

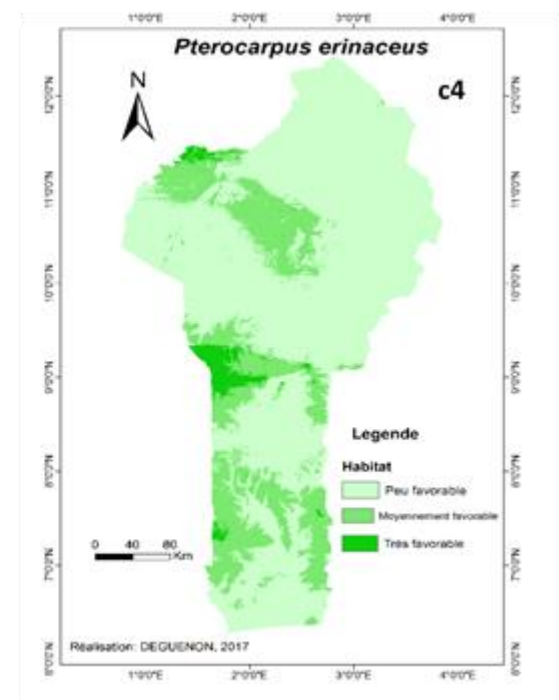
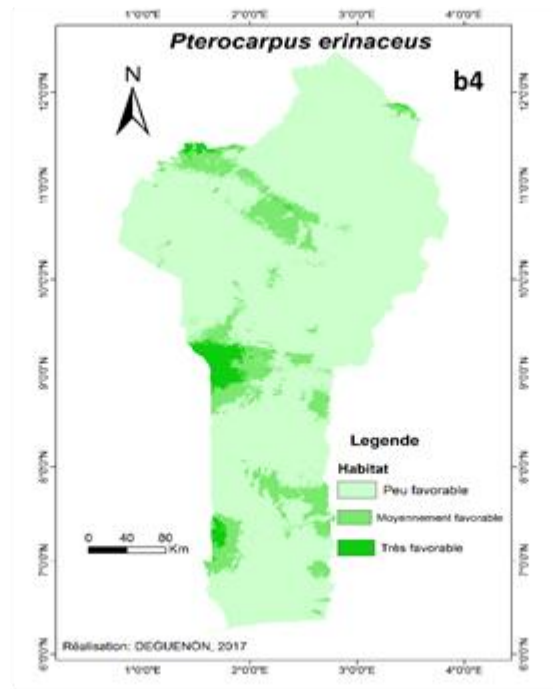
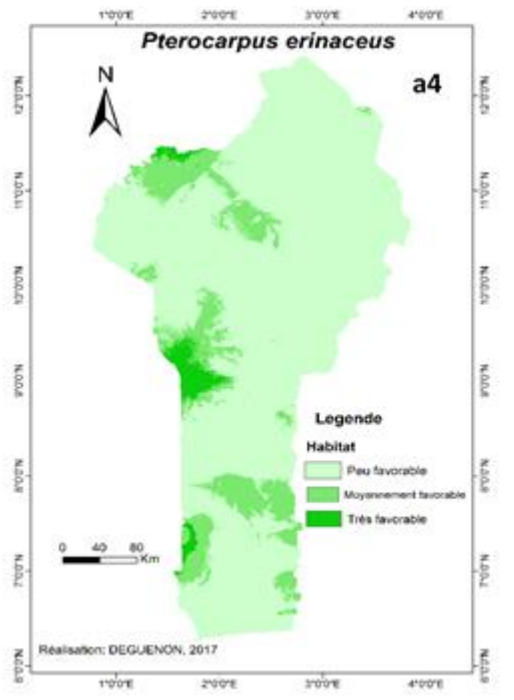
The HadGEM-ES model, compared to the previous one, predicts more the conversion of very favorable habitats (about 40%) into habitats not very favorable for the conservation of *Bombax costatum*, in particular between 12° and 8° North latitude (Table 4, figure 3c10).

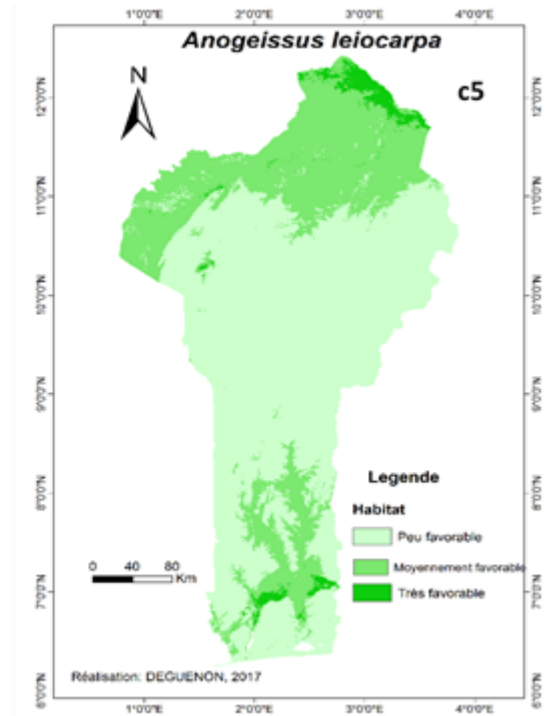
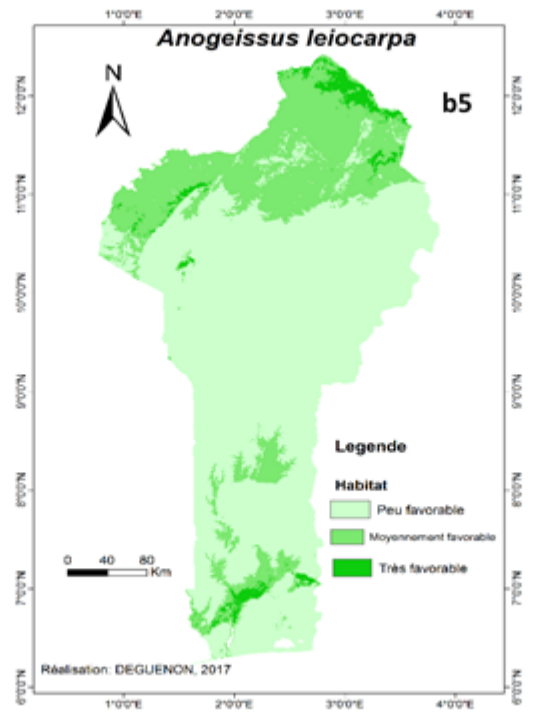
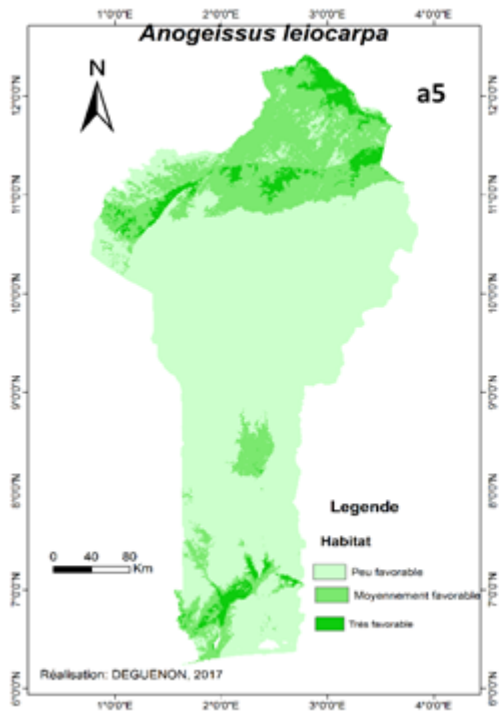
Figure 3: Habitats favorable to the 10 priority species for conservation (current (a) and future distribution (CNRM-CM5: b; HadGEM-ES: c): 2050 as predicted by climate models)

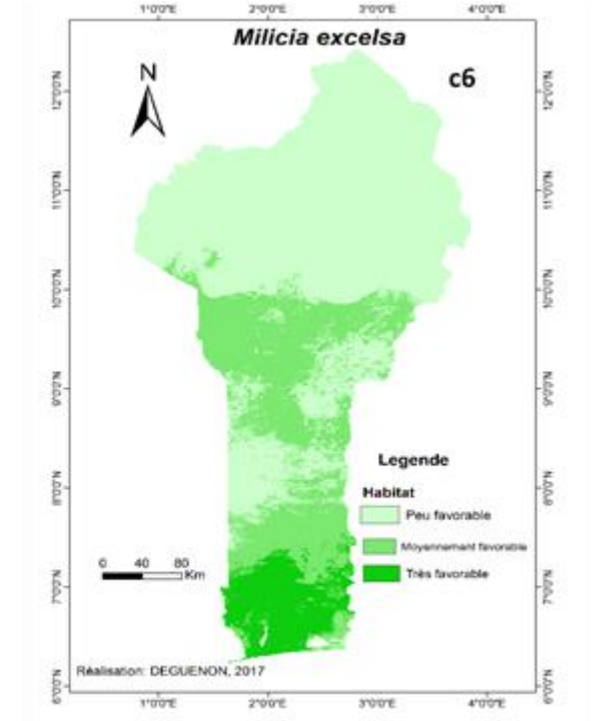
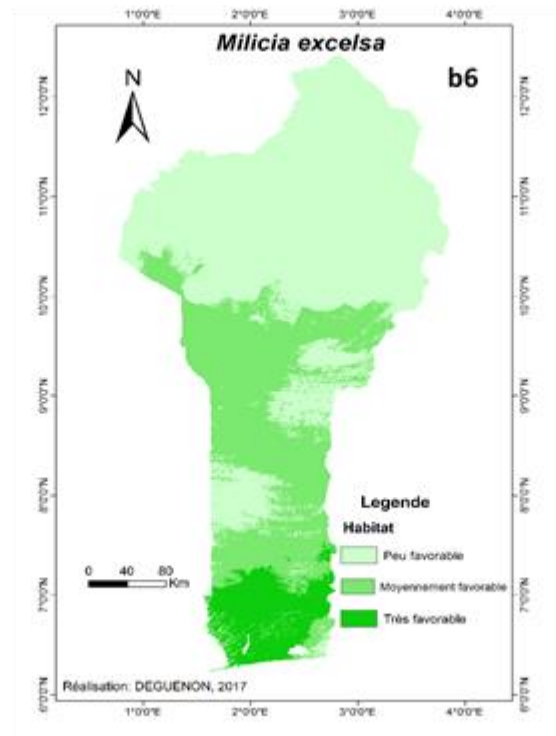
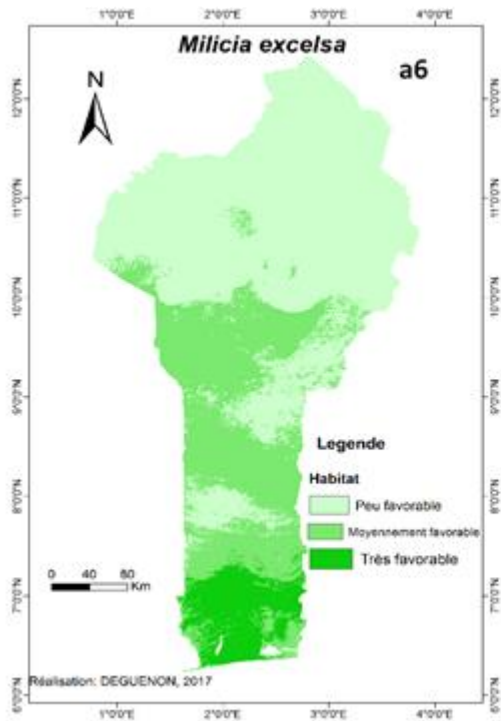


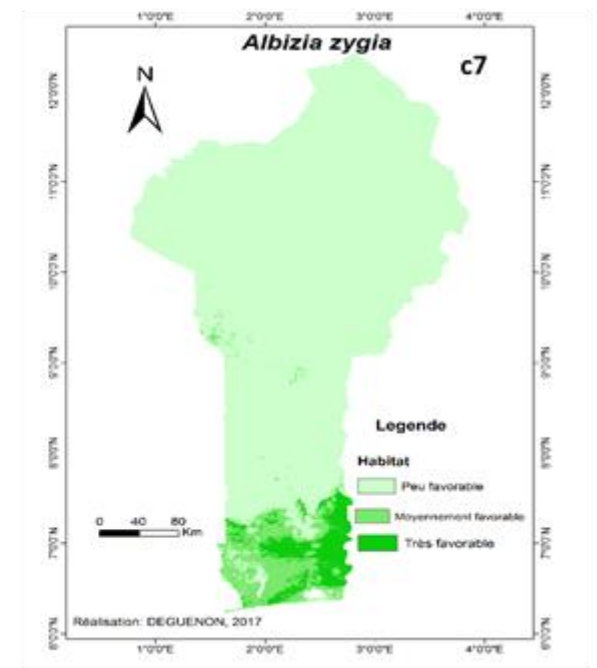
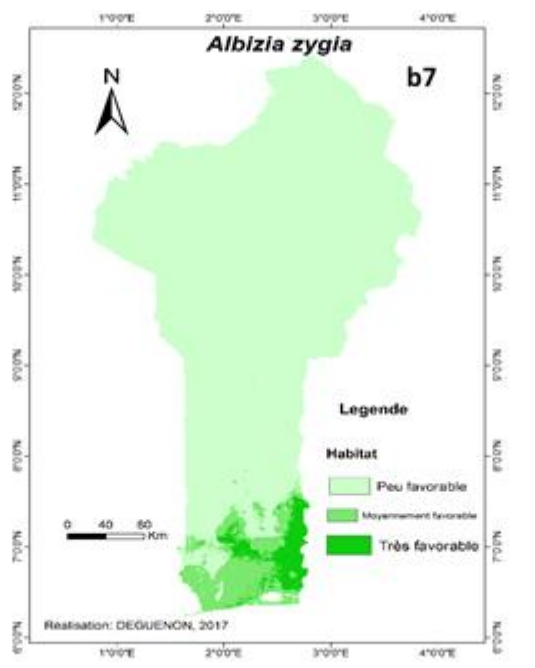
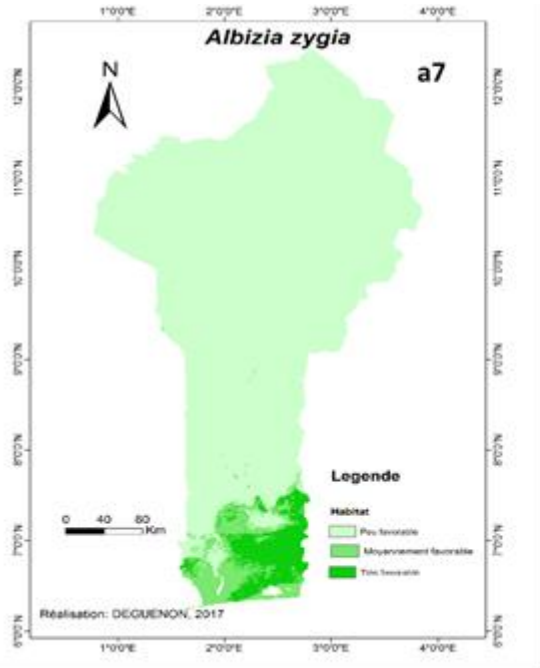


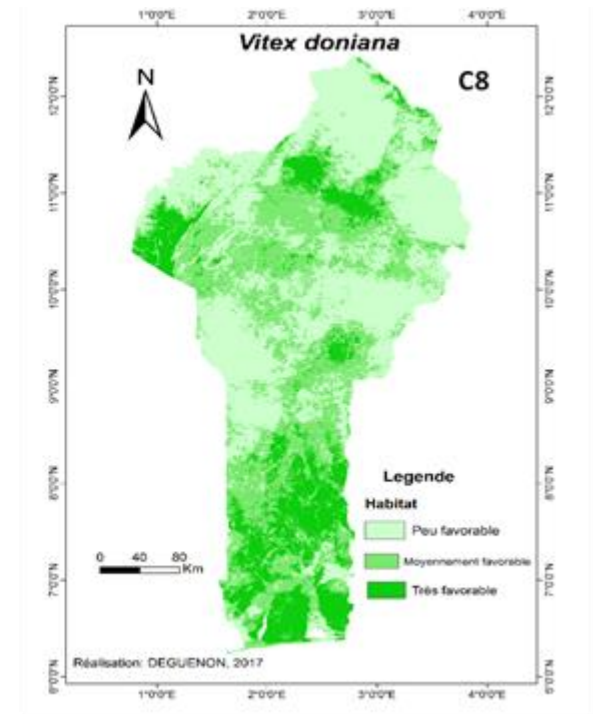
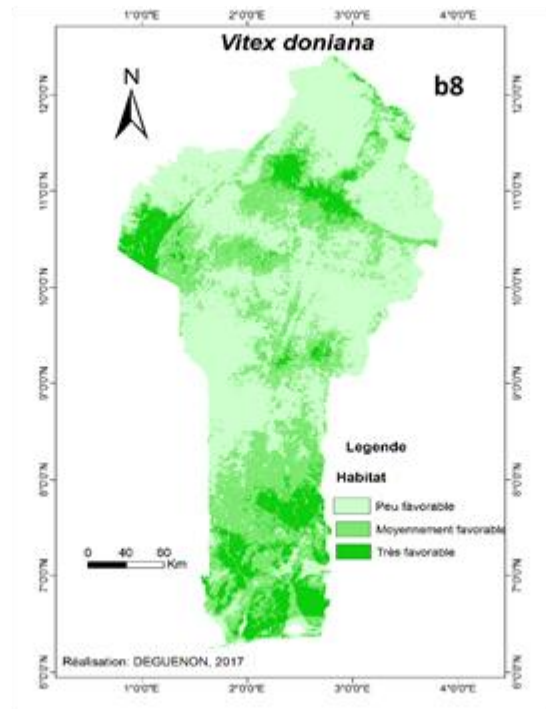
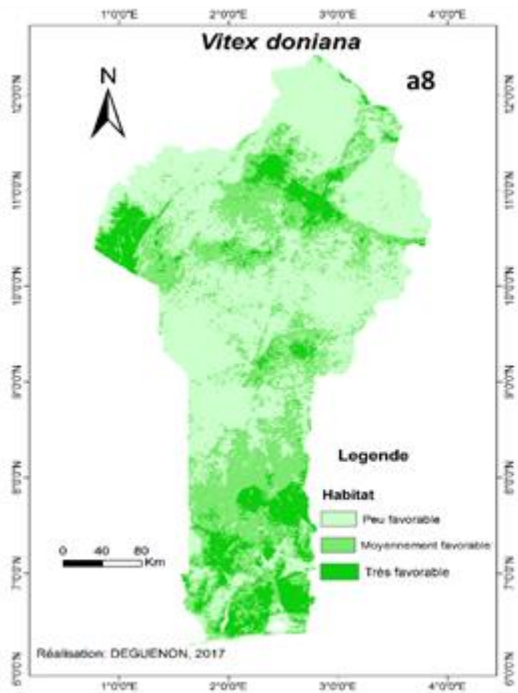


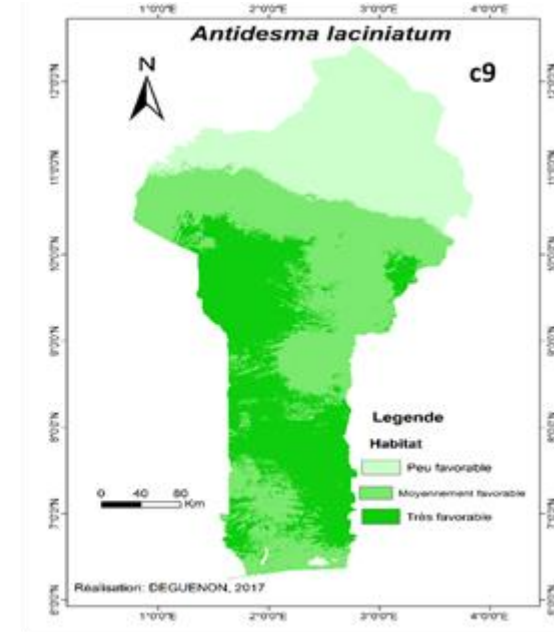
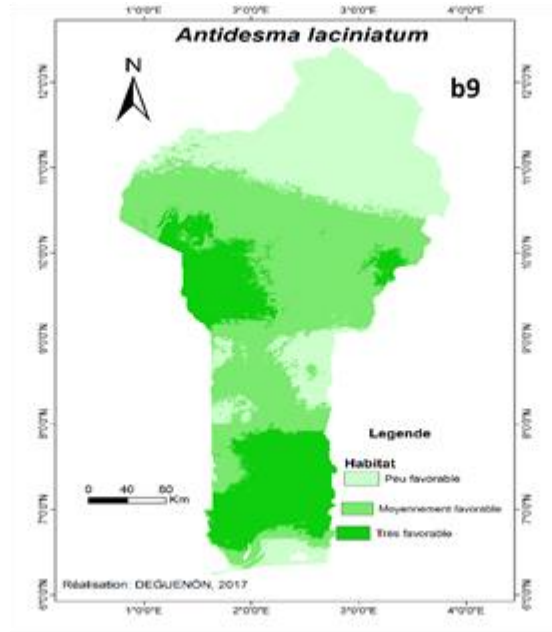
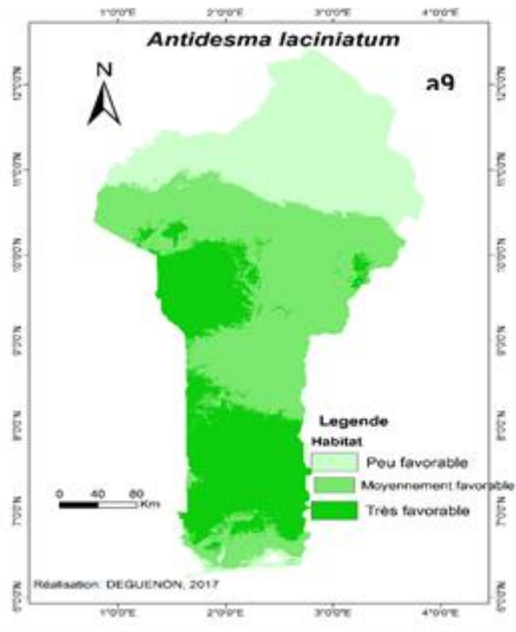












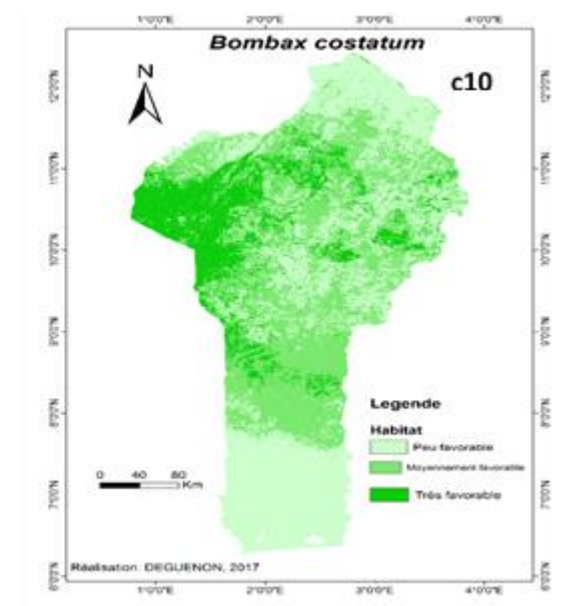
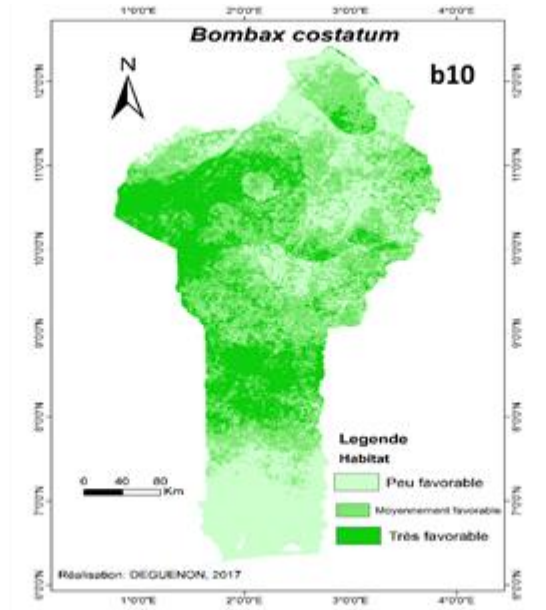
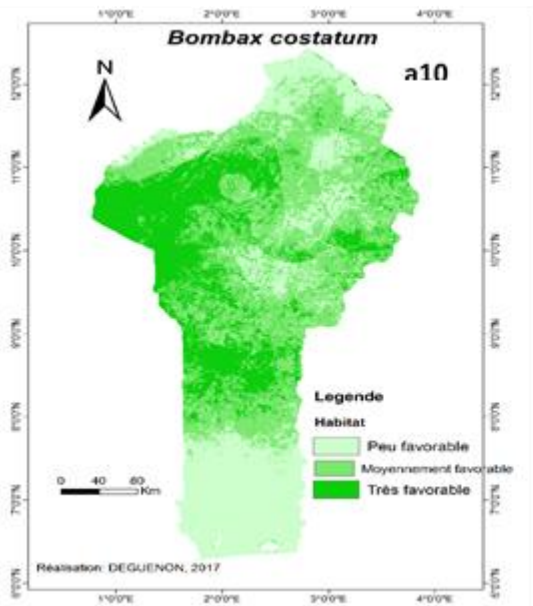


Table 4: Dynamics of areas favorable to the cultivation of ten priority species in classified forests under ONAB management

Species	Culture	Very favorable area		Relatively favorable area		Unfavorable area	
		Area (Km ²)	Trend (%)	Area (Km ²)	Trend (%)	Area (Km ²)	Trend (%)
<i>K. senegalensis</i>	Present	7790,164	-	27314,599	-	77519,238	-
	CNRM-CM5	9929,806	-27,464	30909,969	-13,163	71784,224	7,398
	HadGEM-ES	8409,673	-7,952	34605,718	-26,693	69608,608	10,205
<i>A. africana</i>	Present	17909,042	-	48910,364	-	45804,594	-
	CNRM-CM5	16454,677	8,121	48025,613	1,809	48143,709	-5,107
	HadGEM-ES	20204,166	-12,815	48230,988	1,389	44188,846	3,527
<i>K. grandifoliola</i>	Present	2389,997	-	13675,684	-	96558,3194	-
	CNRM-CM5	1948,035	18,492	10446,984	23,609	100228,981	-3,801
	HadGEM-ES	2672,083	-11,803	16187,653	-18,368	93764,2639	2,894
<i>P. erinaceus</i>	Present	2891,621	-	15189,574	-	94542,805	-
	CNRM-CM5	2597,163	10,183	12693,420	16,433	97333,417	-2,952
	HadGEM-ES	2958,358	-2,308	24578,451	-61,811	85087,191	10,001
<i>A. leiocarpa</i>	Present	5244,454	-	25873,081	-	81506,464	-
	CNRM-CM5	3024,471	42,330	29898,506	-15,558	79701,028	2,215
	HadGEM-ES	2797,468	46,658	34433,035	-33,084	75393,496	7,499

<i>M. excelsa</i>	Present	9008,450	-	35217,968	-	68397,582	-
	CNRM-CM5	8998,887	0,106	35288,738	-0,201	68336,375	0,089
	HadGEM-ES	9498,825	-5,443	31134,080	11,596	71991,095	-5,254
<i>A. zygia</i>	Present	5591,592	-	5970,558	-	101061,85	-
	CNRM-CM5	3368,451	39,759	7410,592	-24,119	101844,956	-0,775
	HadGEM-ES	4861,468	13,057	7361,158	-23,291	100401,374	0,653
<i>V. doniana</i>	Present	14205,137	-	37432,126	-	60986,737	-
	CNRM-CM5	13641,975	3,964	37316,845	0,308	61665,180	-1,112
	HadGEM-ES	19806,766	-39,434	41392,169	-10,580	51425,065	15,679
Species	Culture	Very favorable area		Relatively favorable area		Unfavorable area	
		Area (Km ²)	Trend (%)	Area (Km ²)	Trend (%)	Area (Km ²)	Trend (%)
<i>A. laciniatum</i>	Present	31001,917	-	46185,033	-	35437,050	-
	CNRM-CM5	24969,994	19,457	47885,059	-3,681	39768,947	-12,224
	HadGEM-ES	35434,598	-14,298	43602,515	5,592	33586,887	5,221
<i>B. costatum</i>	Present	30042,872	-	48393,144	-	34187,984	-
	CNRM-CM5	30029,613	0,044	46054,616	4,832	36539,771	-6,879
	HadGEM-ES	18152,927	39,576	49183,571	-1,633	45287,502	-32,466

Discussion

Modeling, Model Reliability, and Contribution of Variables

Modeling of species distribution is widely used nowadays to determine favorable habitats at large scales and to produce comprehensive maps that are particularly used for identifying areas where conservation efforts are needed (Hounkpèvi et al., 2016). During this study, the CNRM-CM5 and HadGEM-ES models were used to predict the future distribution of 10 priority species. The results of this study revealed that the model presents good discrimination of the ecological niche for each of the priority species (AUC mean ratio >1). Thus, the current distribution areas predicted by the model for each species constitute areas of high probability where the species would be present or absent depending on the related ecological conditions. Overall, the distribution of these species depends on climatic and edaphic factors. Variables such as precipitation in the driest quarter, annual precipitation, annual thermal amplitude, and isothermality were found to be the most relevant for most species. This result confirms the fact that direct parameters such as temperature and precipitation are more effective when the modeling of the distribution of species concerns a large area, unlike indirect parameters such as altitude, topography, plant cover which are effective for small areas (Guisan and Zimmermann, 2000). Any significant change in the average states of the climate could therefore generate profound upheavals in the distribution of these species. Climate change is now recognized as one of the main threats to the survival of species and the integrity of ecosystems around the world (Fandohan et al., 2013). Similar results were obtained by Idohou et al. (2016) who showed that the distribution of wild palm depends on these same factors. Other studies have shown that the distributions of species such as *Azelia africana*, *Anogeissus leiocarpa*, *Khaya senegalensis* (Adjahossou et al., 2016) and *Vitex doniana* (Hounkpèvi et al., 2016) are sensitive to environmental variations at several spatial scales.

The used prediction models show that the current distribution areas of the species will undergo profound changes in the future. For *Khaya senegalensis*, the two models used predicted an extension of moderately favorable and very favorable areas by conversion of unfavorable areas. For species such as *Anogeissus leiocarpa* and *Albizia zygia*, very favorable areas will experience a retraction in favor of moderately favorable areas. While *Vitex doniana* will experience a retraction of its very favorable range according to the CNRM-CM5 model, the HadGEM-ES model foresees rather a large extension of the very favorable ranges. Climate change could generate a spatial dynamic in the geographic distribution of habitats favorable to these species and thus make certain regions favorable today, very unfavorable in the future and vice versa (Hannah et al., 2002). For a good conservation action, it would be preferable to study for each of these species the tolerance limits of

these two parameters (water and temperature) which significantly affect the distribution through the phytodistricts. Most species with a wide distribution (*Vitex doniana*, *Azelia africana*, *Antidesma laciniatum*, *Bombax costatum*) are expected to remain fairly stable under the effect of future climate changes, even under the most drastic scenarios. According to Adjahossou et al. (2016), species such as *Khaya senegalensis*, *Anogeissus leiocarpa* and *Azelia africana* being potential agroforestry species, used in agroforestry systems in the Guinean zone, where there are very few protected areas, could be considered in reforestation policies.

Implications of the Study

Overall, the distribution of these species depends on climatic and edaphic factors. The direct (driest quarter precipitation, annual precipitation) and indirect (soil) parameters were found to be more effective in predicting favorable priority areas in the conservation of the species studied. The Sudano-Guinean (7 ° 30'-11 ° 5'N) and Guinean (6 ° 25'-7 ° 30'N) zones appear to be more concentrated in priority habitats, while the Sudanese zone (10 ° 00'- 12 ° 27'N) is found to be concentrated in non-priority habitats. These areas seem likely to provide the ten species with suitable climatic conditions for their cultivation and conservation. Restoration policies combining in situ conservation strategies for the ten species should favor these priority unprotected areas because they are likely to provide the species with adequate climatic conditions. Also, the constitution of databases on the potential impact of climate change on forest species according to the different scenarios available could accelerate adaptive decision-making for their regeneration.

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