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RECEIVED 26 September 2024

ACCEPTED 24 December 2024

PUBLISHED 10 January 2025

## CITATION

Wijenayake PR, Shitara T, Hirata A, Matsui T,  
Kubota Y and Masaki T (2025) Changes in  
forest ecosystem stability under climate  
change in a temperate landscape.  
*Front. For. Glob. Change* 7:1501987.  
doi: 10.3389/ffgc.2024.1501987

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# Changes in forest ecosystem stability under climate change in a temperate landscape

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Climate change poses significant threats to forests globally. Understanding the relationship between environmental variables and species distribution is crucial for evaluating the vulnerability of tree species assemblies to anticipated climate change. Here, we address whether projected future changes in climate suitability are related to the structural stability of the old-growth forest community in Japan. We hypothesize that even with the expected changes in climate, the structural stability of the species assembly will remain unchanged until the end of this century. We modeled the influence of climate change on the spatial distribution of major tree species in a temperate deciduous forest reserve using local and regional presence data. We used the Maxent model and QGIS software to project potential habitat changes. Focusing on the period 2081–2,100, we used the MRI-ESM2-O general circulation model under baseline (SSP5–8.5) and mitigation (SSP1–2.6) future climate scenarios. This revealed that winter temperature is the most crucial factor affecting the distribution of tree species in the temperate landscape. Canopy tree species such as *Acer pictum* and *Castanea crenata* are projected to remain stable under SSP5–8.5 in 2100. Our results also suggest that the distribution of *Quercus serrata*, the dominant species in the forest studied, will expand, particularly under extreme climate conditions in 2100. However, there may be potential reductions in the abundance of subcanopy species, indicating a change in the structure of the forest stand. In this sense, the stability of forest ecosystems and local species diversity may be vulnerable under future climate change scenarios. Exploring the future species distribution and stand structure can improve understanding of habitat changes in temperate landscapes and requires more focused research efforts.

## KEYWORDS

climate change, Ogawa Forest Reserve, tree species distribution, canopy, subcanopy species

## 1 Introduction

Climate change poses an imminent threat to global biodiversity, prompting urgent investigations into its profound impacts on ecosystems (IPBES, 2019). Moreover, a decline in biodiversity will deepen the climate crisis, with consequences including reduced species abundance, local extinctions, and the rapid degradation or loss of ecosystems such as mangroves, forests, peatlands, meadows, and seagrass (Macreadie et al., 2021). These changes not only affect the capacity of the planet to store carbon but also impede the ability of nature and humanity to adapt to or cope with evolving climate conditions (Pettorelli et al., 2021).

Forests, vital agents in climate change mitigation, face unprecedented challenges due to the anticipated pace of environmental shifts (Hansen et al., 2001; Hamann and Wang, 2006; Warren et al., 2018). Despite their critical role, our understanding of how climate influences the spatial distribution (Canham and Murphy, 2017), abundance, and assembly of tree species remains limited (Chu et al., 2019). Future changes in tree assembly will significantly affect forest ecosystems and ecosystem services.

The capability of conservation and management practices to address climate change depends on the ability to forecast climate trends in the coming decades. This includes predicting ecological and biogeographic responses to climate change and anticipating the impacts of future land management practices (Colavito, 2017). A comprehensive understanding of how forest ecosystems react to climate change is crucial for reforestation initiatives, such as selecting species adapted to future climate conditions (Harger, 1993). In many temperate forest ecosystems, a shift from conifer-dominated to broadleaf-dominated ecosystems is anticipated (Hanewinkel et al., 2014; Thom et al., 2017). To forecast such changes, species distribution models (SDMs), also known as Ecological Niche Models, bio-envelope models, or species envelope models, have been widely used to determine the responses of species under different climate change scenarios (e.g., Guisan and Zimmermann, 2000; Elith and Leathwick, 2009; Tanaka et al., 2012; Nakao et al., 2013; Tang Y. et al., 2018; Ohashi et al., 2019).

Increasingly, research is examining long-term changes in vegetation in response to climate change (Anderson and Song, 2020) and the associated changes in the carbon balance (Mo et al., 2023). It seems unlikely that forests will be able to maintain their current stand structure and total carbon sequestration and storage potential with the changing climate (Thompson et al., 2009). Considering the likelihood that some species will decline under climate change, then which other tree species will be more resilient to the anticipated climate and may therefore be considered suitable alternatives? Instead of relying on models centered on individual species, it might be more beneficial to predict forthcoming changes in spatial distribution as a collective assembly (Zurell et al., 2020). It is important to predict potential habitat distributions at national, regional, and landscape scales under current and future climate conditions (Tanaka et al., 2005).

Temperate deciduous forests in the Northern Hemisphere, situated between evergreen and subarctic coniferous forests, experience distinct seasonal cycles and diverse compositions (Gilliam, 2016). These characteristics make them especially vulnerable to compositional changes driven by climate shifts. Despite their ecological significance, research on future species assemblages in East Asia, particularly in Japan, remains limited. Here the southern (lower) limit of distribution lies adjacent to the upper limit of evergreen broad-leaved forests, and if these forests are invaded, the light environment of the forest floor is likely to change significantly, leading to shifts in species composition and a potential decline in species diversity. The unique position of temperate deciduous forests makes them ideal systems for studying the impacts of climate change on forest ecosystems (Ohsawa, 1993). Moreover, temperate forests are home to numerous tertiary relict plants and endangered species, emphasizing their critical importance for biodiversity conservation and ecological research (Tang C. Q. et al., 2018).

Within this context, the Ogawa Forest Reserve (OFR) emerges as a unique example of an old-growth forest, situated in the southern

Abukuma Mountains of Japan. Despite detailed studies on its forest dynamics and structure (Masaki et al., 1992; Nakashizuka et al., 1992; Shibata et al., 2010; Masaki et al., 2021; Wijenayake et al., 2023) there is little in-depth information on the future distribution of species in the vicinity of the Forest Reserve.

This study seeks to bridge this knowledge gap using species distribution models to predict the current and future geographical distribution of major tree species found in vicinity of the OFR under different climate change scenarios. More specifically, we address whether the projected changes in climate suitability are related to the structural stability of a plant community in terms of vertical and horizontal spatial organization and verify the generality of the findings based on the previous related studies on temperate forests. We hypothesized that even with the expected changes in climate, the structural stability of the species assembly will remain unchanged until the end of this century. This hypothesis is based on the observation that populations are already responding to climate trends in the same direction as expected in the future (Jump et al., 2009).

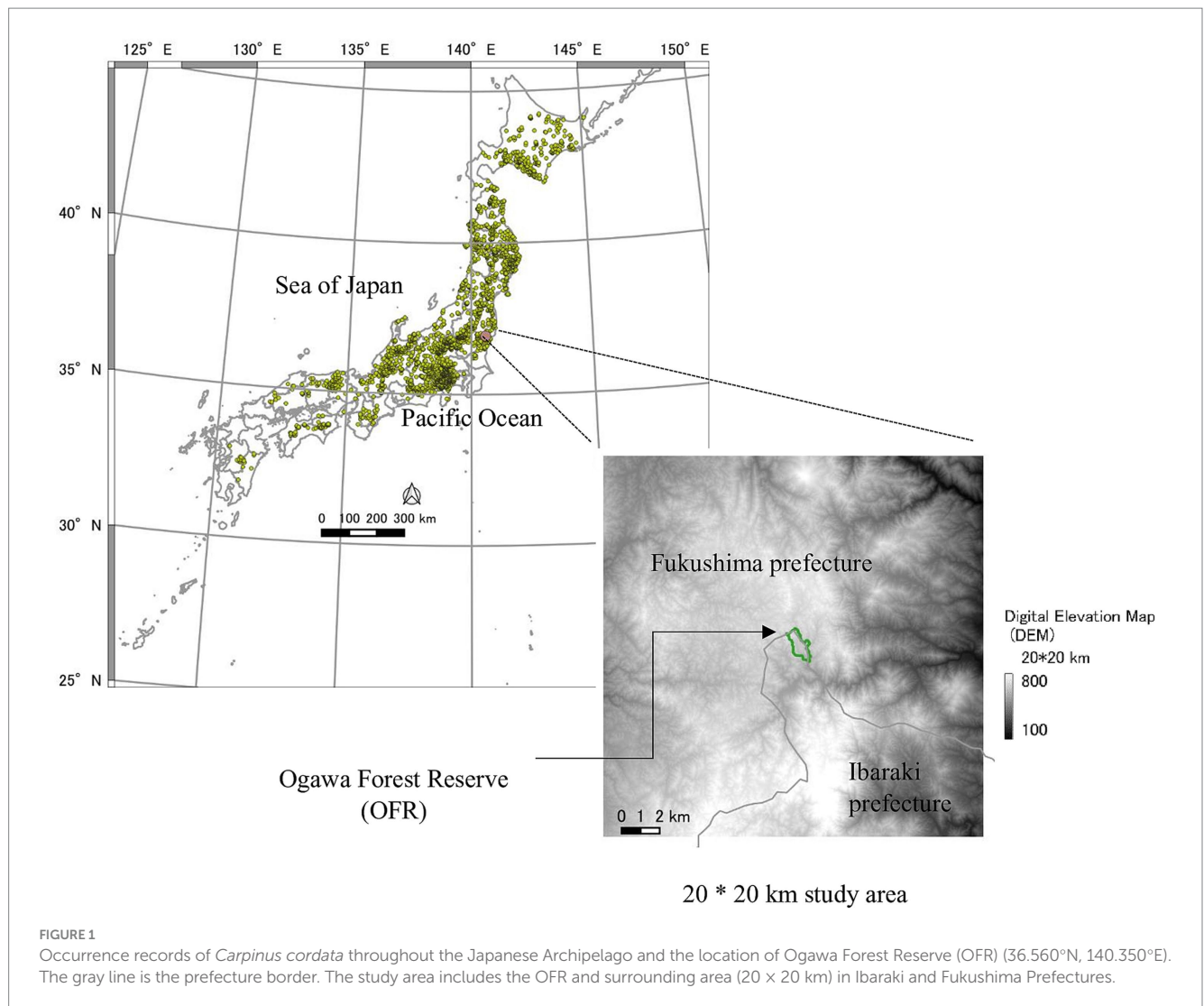
Our results provide detailed insights into the possible trends of the most abundant tree species and the impact of changes in the local tree species abundance and composition on the tree species assembly under future climate scenarios in the study area. They also provide information on potential forest ecosystem stability, which should be given special attention in the context of local species diversity. Overall, this work offers insights into the intricate relationships among species distribution, climate change, and the landscape surrounding OFR, offering valuable information for optimizing conservation and management strategies.

## 2 Materials and methods

### 2.1 Study area and target species

We studied the major tree species found in the OFR (100 ha) in northern Ibaraki Prefecture, Japan (36.56°N, 140.35°E) (Figure 1) at elevations of 610–660 m. The climate is variable, with average temperatures ranging from  $-0.9^{\circ}\text{C}$  in February to  $22.6^{\circ}\text{C}$  in August and a maximum snow depth of around 50 cm (Mizoguchi et al., 2002). The topography is gently undulating (Yoshinaga et al., 2002). This old-growth forest is characterized by the dominance of *Quercus* and *Fagus* (Masaki et al., 1992). The distributions of 17 major tree species were projected under both the present and future climate scenarios. The species included five subcanopy and 12 canopy species (life forms and life history traits of the species from Table 1 in Wijenayake et al., 2023). We defined a  $20 \times 20 \text{ km}^2$  region located centrally in the OFR to determine potential habitat changes surrounding the reserve.

First, we examined the occurrence of the selected tree species throughout Japan. These species, recognized as pivotal components of old-growth forests, particularly within the OFR, formed the basis for a comprehensive understanding of their distribution nationwide. Subsequently, climate projections were developed to cover Japan, shedding light on the prevailing climate. Building on this broader perspective, we refined our analysis by narrowing the focus to the specific geographic area. This involved selectively cropping the designated region, centering it around the OFR. This facilitated a more detailed examination of climate scenarios and potential habitat changes within the immediate vicinity of the OFR. The transition from



nationwide considerations to localized assessments ensured a thorough reliable understanding of the distributions of the targeted tree species in response to varying climate conditions.

## 2.2 Distribution data compilation

We used the Japan Biodiversity Mapping Project (J-BMP) database, which contains species occurrence information for vascular plants accumulated through research activities and environmental assessments conducted independently by researchers, local governments, environmental assessment companies, and citizen scientists (Shiono et al., 2021). The J-BMP database includes comprehensive information on species occurrence, functional traits, and phylogeny, which are described in detail on the J-BMP website.<sup>1</sup> The data used in the current study included only presence records with defined geographic coordinates following the WGS84 geographic coordinate system. Geographic locations were verified using latitude

and longitude information, and any records with discrepancies were discarded. After this data selection process, the final dataset included 8,510 species. Species distribution models (SDMs) were then constructed for each of these species using MaxEnt, with presence data serving as the response variable. Inaccurate data were filtered using QGIS 3.22.7,<sup>2</sup> resulting in 66,424 occurrence points. These points were used to develop an initial model predicting the current species distributions. The nomenclature followed that of Yonekura and Kajita (2003).

## 2.3 Environmental data

We considered two future climate scenarios: the baseline and mitigation scenarios. In the baseline scenario (SSP5–8.5), in which no reductions in greenhouse gas emissions are assumed, the global mean surface temperature is projected to increase by 2.6–4.8°C. By contrast, the mitigation scenario (SSP1–2.6) assumes ambitious climate change

<sup>1</sup> <https://biodiversity-map.thinknature-japan.com/index.html>

<sup>2</sup> <https://qgis.org/ja/site/index.html>

mitigation efforts aligned with the goal of limiting global warming to less than 2°C by 2,100.

For species distribution models, the selection of environmental variables is crucial. In this study, we selected variables from WorldClim, which includes 19 global climatic layers from 1970 to 2000 at 1 km<sup>2</sup> spatial resolution. In addition, WorldClim CMIP Phase 6 (CIMP6) provided 95 global climate layers for future suitability models, focusing on the period 2081–2,100, using the MRI-ESM2–0 General Circulation Model (GCM) and two shared socioeconomic pathways (SSPs): SSP1–2.6 and SSP5–8.5 (see also Hijmans et al., 2005).<sup>3</sup>

## 2.4 Variable selection

Regions with low topographical complexity may be at particular risk of critical transitions under climate change, as evidenced in intermediate and uniform topography scenarios (Scheffer et al., 2012). Given the gentle topography of the study area, we used the first 19 bioclimatic variables for our SDMs, excluding nonclimate variables such as topography, soil, surface geology, slope, and aspect. Furthermore, the relative importance of these variables was minimal in previous studies (Nakao et al., 2013; Matsui et al., 2018; Shitara et al., 2021).

## 2.5 Species distribution model

As the modeling statistical method, MaxEnt (ver. 3.4.3)<sup>4</sup> was chosen due to its higher performance and accuracy compared to other SDM tools. MaxEnt is successful because its regularization process avoids overfitting, particularly when using small sample sizes (Elith et al., 2011). From among 19 variables, we initially selected 12 climate variables that could be interpreted physiologically or ecologically as candidate variables for modeling. Then the bioclimatic variables that best explained the species distribution were determined using the ‘varSel’ function in the ‘SDMtune’ R package. The correlation threshold for removing highly correlated variables was set at a Spearman’s correlation coefficient of 0.7. Next, the models were optimized using the regularization multiplier and feature combination. The regularization multiplier values were set from 0.5 to 5 at 0.5 increments. For feature combinations, we tested the pattern, which combines linear (l), quadratic (q), product (p), threshold (t), and hinge (h). The best models were selected using the Akaike Information Criterion adjusted for small samples (AICc) and assessed using the area under the curve (AUC) of the receiver operating characteristic (ROC) curve. The mean of the MaxEnt predictions was mapped using QGIS. Model performance was evaluated using the continuous Boyce index (CBI), a metric designed to assess the quality of predictions based solely on presence data. The index ranges from –1 to 1, where negative values indicate a poor model, values near 0 represent random predictions, and positive values reflect predictions aligned with the presence distribution in

the test data. In this study, models with a CBI > 0 within the 95% confidence interval were selected for subsequent analyses.

When examining future-scenario models, it can be challenging to determine the areas that are either more or less suitable for each species. In this study, the distribution maps for each species were assessed using QGIS 3.22.7. Habitat suitability was calculated for each current scenario and 2,100 climate-model scenarios across all species. Then we combined each future scenario model projection with the present-time data to generate a categorical map showing stable, shrinking, and expanding areas.

## 3 Results

### 3.1 Model performance and variable importance

The predictive accuracies of all SDM algorithms were generally good across most species. The mean AUC varied between 0.613 and 0.829 and the mean CBI varied between 0.775 and 1.000 (Table 1). This indicates that the habitat suitability suggested by the models resembles the real probability of species occurrence. These metrics also indicate the strength of climate controls on the distribution of each species. In terms of AUC, the predictive accuracies were smaller for *Cornus controversa* and *Kalopanax septemlobus* (AUC ≈ 0.62).

Variable selection using Maxent reduced the environmental suitability components to 11 variables (Table 2). The importance of each variable differed across species; however, temperature in winter, represented by BIO06 and BIO11, was the most important predictor in more than half of the species. BIO11 (the mean temperature of the coldest quarter) suggests that *Carpinus tschonoskii*, *Fagus crenata*, *C. crenata*, *Cerasus leveilleana*, and *Cornus controversa* thrive in locations where the temperature remains at about 0° during winter. BIO06 is related to the minimum temperature of the coldest month and suggests that *Carpinus cordata*, *Carpinus laxiflora*, *Styrax obassia*, and *Acer amoenum* can survive at temperatures in the –10°C to –5°C range while lower temperatures are much less suitable. The variable BIO05 is related to the maximum temperature of the warmest month and suggests that *Q. serrata* prefers a range of 20–25°C while *Acer rufinerve* has a slightly wider range. The majority of species are associated with the mean temperature of the wettest quarter (BIO08). The distributions of *F. crenata* and *Carpinus japonica* are mostly influenced by the precipitation in the warmest (BIO18) and coldest (BIO19) quarters. Figure 2 shows the response curve of *Cerasus leveilleana* while the other response curves are in Appendix 1.

### 3.2 Current and future geographical distribution of major tree species

Figure 3 compares the environmental suitability at present and in 2100. This depicts a future where the species seem to have limited spread. This extreme CO<sub>2</sub> emissions scenario (SSP5–8.5) negatively affected all of the major tree species except for *Castanea crenata*, *A. pictum*, and *Q. serrata*. The distributions of the subcanopy species *Carpinus cordata*, *Carpinus japonica*, *C. tschonoskii*, *A. rufinerve*, and *S. obassia* all shrink (Table 1 of Wijenayake et al., 2023).

<sup>3</sup> <https://www.worldclim.org/>

<sup>4</sup> [https://biodiversityinformatics.amnh.org/open\\_source/maxent/](https://biodiversityinformatics.amnh.org/open_source/maxent/)

TABLE 1 Evaluation of algorithm performance based on the mean area under the curve (AUC) for the species distribution models (SDMs) of each tree species.

Species	Family	Abbreviation	Life form	AUC	Threshold level	CBI
<i>Betula grossa</i>	Betulaceae	B.g	Canopy	0.789	0.355	1.000
<i>Carpinus cordata</i>	Betulaceae	Ca.c	Subcanopy	0.771	0.424	0.820
<i>Carpinus japonica</i>	Betulaceae	Ca.j	Subcanopy	0.750	0.429	0.999
<i>Carpinus tschonoskii</i>	Betulaceae	Ca.t	Subcanopy	0.730	0.432	0.999
<i>Carpinus laxiflora</i>	Betulaceae	Ca.l	Canopy	0.691	0.445	0.992
<i>Fagus crenata</i>	Fagaceae	F.c	Canopy	0.829	0.358	0.997
<i>Fagus japonica</i>	Fagaceae	F.j	Canopy	0.788	0.337	0.999
<i>Quercus crispula</i> var. <i>crispula</i>	Fagaceae	Qu.c	Canopy	0.813	0.359	0.537
<i>Quercus serrata</i>	Fagaceae	Q.s	Canopy	0.768	0.403	0.873
<i>Castanea crenata</i>	Fagaceae	C.cr	Canopy	0.679	0.447	0.979
<i>Cerasus leveilleana</i>	Rosaceae	Ce.l	Canopy	0.724	0.453	0.989
<i>Acer amoenum</i>	Sapindaceae	A.a	Canopy	0.701	0.448	0.968
<i>Acer rufinerve</i>	Sapindaceae	A.r	Subcanopy	0.744	0.445	0.982
<i>Acer pictum</i>	Sapindaceae	A.p	Canopy	0.726	0.504	0.826
<i>Cornus controversa</i>	Cornaceae	Co.c	Canopy	0.613	0.481	0.639
<i>Kalopanax septemlobus</i>	Araliaceae	K.s	Canopy	0.618	0.518	0.136
<i>Styrax obassia</i>	Styracaceae	S.o	Subcanopy	0.750	0.413	0.989

CBI, continuous Boyce index.

TABLE 2 The final set of environmental variables for BIO1 to BIO19 used to build the models.

Species	BIO05	BIO06	BIO08	BIO010	BIO11	BIO13	BIO14	BIO16	BIO17	BIO18	BIO19
B.g			4.20	34.4				58.80			2.84
Ca.c		61.96	11.18			13.20	13.62				
Ca.j					41.74					43.36	14.88
Ca.t			10.80		59.3	22.08	7.78				
Ca.l		61.50	5.44				7.64			25.46	
F.c			22.86		36.50					15.26	25.34
F.j			12.60		46.72	20.28	20.38				
Qu.c			0.94	90.66			6.10	2.30			
Q.s	89.30					0.50					10.22
C.cr			30.10		31.5				21.92	16.54	
Ce.l			6.46		44.46	42.04			7.02		
A.a		40.32	8.10			13.74	37.82				
A.r	57.38		5.14			22.82					14.66
A.p						14.14	85.86				
Co.c			13.11		40.70	31.74			14.42		
K.s			6.40	90.28		3.18					0.16
S.o		61.26	8.46							22.1	8.10

Numbers indicate the contributions of major environmental factors. Bright and light pink colored cells are the two most important variables for each species. Values highlighted in bright pink produce the greatest response in the species. BIO5, maximum temperature of warmest month; BIO06, minimum temperature of coldest month; BIO08, mean temperature of wettest quarter; BIO10, mean temperature of warmest quarter; BIO11, mean temperature of coldest quarter; BIO13, precipitation of wettest month; BIO14, precipitation of driest month; BIO16, precipitation of wettest quarter; BIO17, precipitation of driest quarter; BIO18, precipitation of warmest quarter; and BIO19, precipitation of coldest quarter.



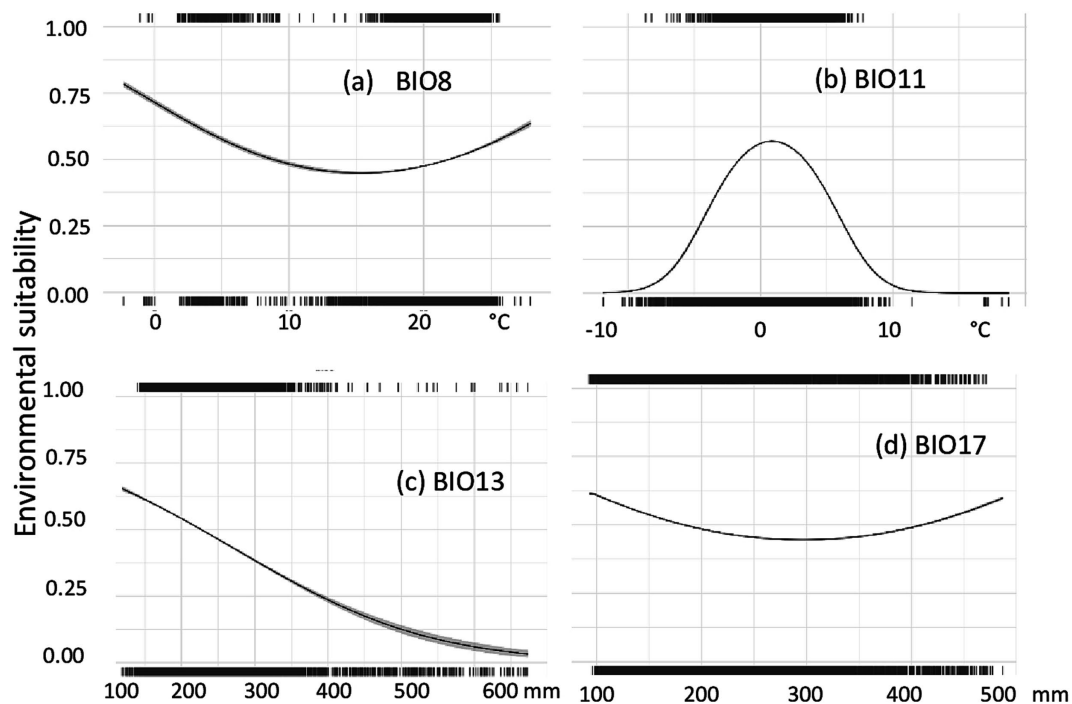


FIGURE 2

Response curves in the Maxent model of habitat probabilities for *Cerasus leveilleana* with respect to different bioclimatic variables. The response curves are presented as the means of 100 replicates with standard deviations shown in gray. (A) BIO8, mean temperature of the wettest quarter; (B) BIO11, mean temperature of the coldest quarter; (C) BIO13, precipitation of the wettest month; and (D) BIO17, precipitation of the driest quarter.

Figure 4 presents the habitat suitability and difference in (future–present) suitability for two species: *C. tschonoskii* and *Q. serrata*. The areas suitable for *C. tschonoskii* shift inland away from the Pacific Coast, while the habitat suitability of *Q. serrata* expands under a changing climate. Habitat suitability maps for the other species are in Appendix 2.

## 4 Discussion

This study identified the climatic variables that seem most important for determining the habitat suitability of major tree species at present, and under two different future climate change scenarios. The majority of species, including *Carpinus. cordata*, *Carpinus. laxiflora*, *A. amoenum*, *S. obassia*, *Carpinus. tschonoskii*, *F. crenata*, *Castanea. crenata*, *Cerasus. leveilleana*, and *Cornus. controversa*, are affected by winter temperatures (BIO06 and BIO11), which is the most important factor affecting the future distribution of the species. Projected future changes for each species centered around the OFR under two SSP scenarios (a scenario implying low greenhouse gas forcing in the future [SSP2] and a fossil fuel development scenario [SSP5] for 2050 and 2,100) suggest that the distributions of tree species in and around OFR will change until the end of the 21st century. Under both climate scenarios, some structural changes are related to the species composition of the assembly of subcanopy species. However, the uncertainties and potential sources of error in model predictions may arise from complex biotic and abiotic interactions within the community, which are often challenging to capture

accurately. Incorporating additional variables, such as terrain and soil characteristics, may only lead to modest improvements in model performance, as these factors might not fully account for the intricate ecological dynamics influencing species distributions.

The literature suggests that the structural complexity of the canopy in old-growth forests facilitates greater resource use efficiency while maintaining significant carbon storage potential (Hardiman et al., 2013; Obata et al., 2023). Near OFR, the subcanopy species *C. cordata*, *A. rufinerve*, and *S. obassia* will disappear completely by 2,100 under SSP5–8.5 while the remaining two subcanopy species (*Carpinus japonica* and *Carpinus tschonoskii*) will markedly decrease. In the 2,100 climate mitigation scenario, a similar trend is observed for two subcanopy species: *S. obassia* and *C. cordata*. The distribution of the canopy species *F. crenata* will shrink drastically by the end of this century. Another study of the possible future of *F. crenata* in Japan identified vulnerable habitats at low elevations on both the Sea of Japan and Pacific Ocean sides (Matsui et al., 2018). The distributions of the canopy species *Betula grossa*, *A. amoenum*, *C. controversa*, and *K. septemlobus* will shrink remarkably according to projections.

With the potential reductions in the abundance of subcanopy species, the forest structure is expected to change. However, prolonged exposure to unsuitable environmental conditions can lead to the physiological deterioration of trees, disrupting the rhythms of flowering, fruiting, and fertility (Selås et al., 2002; Wright et al., 2022). Consequently, the growth of the next generation of fruiting and juvenile trees may be hindered, resulting in a decline in population density (Niinemets, 2010). Concurrently, competition from species better adapted to warmer

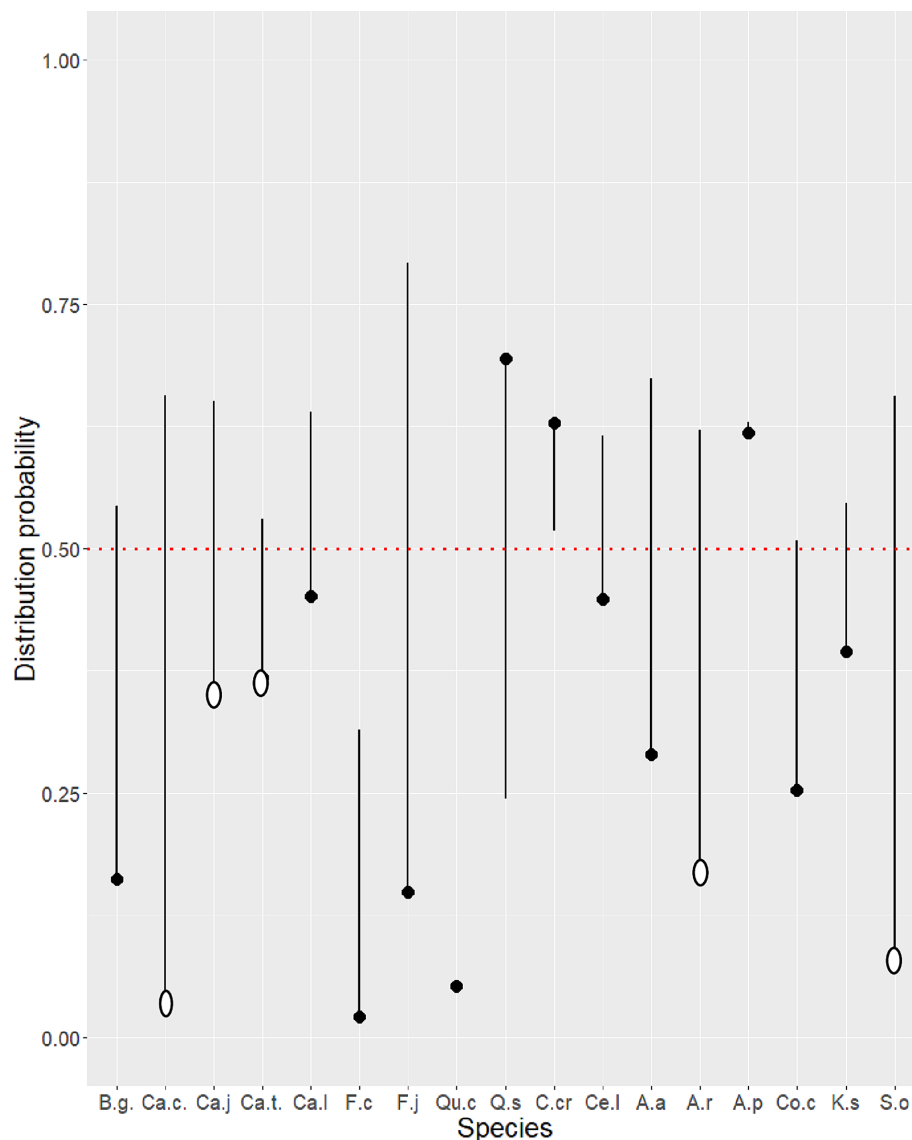


FIGURE 3

Expected values of distribution probabilities of 16 major species near OFR from the present to 2,100 are shown by the black dots (canopy) and white ovals (subcanopy). The ends of the lines extending from the dots indicate the present values. The red dotted line is the threshold level. Here, only the SSP5–85 scenario is considered.

environments (e.g., evergreen broadleaved species) may induce changes in forest species composition and structure, accompanied by increased invasion and establishment from external sources, such as the growth of warm-temperate tree species (Lindner et al., 2010).

Stand structure has important implications for forest ecosystems, including the maintenance of wildlife habitat, sustaining biodiversity, nutrient cycling, regulating climate via carbon storage, and affecting forest regeneration (Brassard and Chen, 2006). In this sense, changing climate will likely affect the stability of the plant assembly and the habitats of various mammals and birds will also change. Accordingly, species diversity in the ecosystem surrounding OFR is expected to decrease. Based only on our findings, however, we cannot make a general statement regarding forest ecosystem stability throughout Japan. There may be places where climate change is working against *Q. serrata* and favoring subcanopy species.

Overall, *Q. serrata* is projected to remain stable and even expand its distribution, even under extreme climate conditions in 2100. Notably, it is the only one of the 17 species that expands its distribution area under both the climate adaptation and mitigation scenarios (Figure 4). These results suggest a simplified species composition of the canopy layer and a reduced abundance of subcanopy species in OFR in the future. *Q. serrata* occurs in the intermediate temperate forest zone and is widely distributed in lower and warmer areas, particularly in less snowy environments (Matsui et al., 2018). A demographic study of the life history of OFR indicated that *Quercus* species may continue to dominate the future landscape, primarily due to the higher survivorship at diameters at breast height of 10–40 cm (Wijenayake et al., 2023). The canopy species *A. pictum* and *C. crenata* will also remain stable under SSP5 in 2100.

Taking proactive steps to understand how trees respond to climate, through well-planned tests in specific areas, could bring

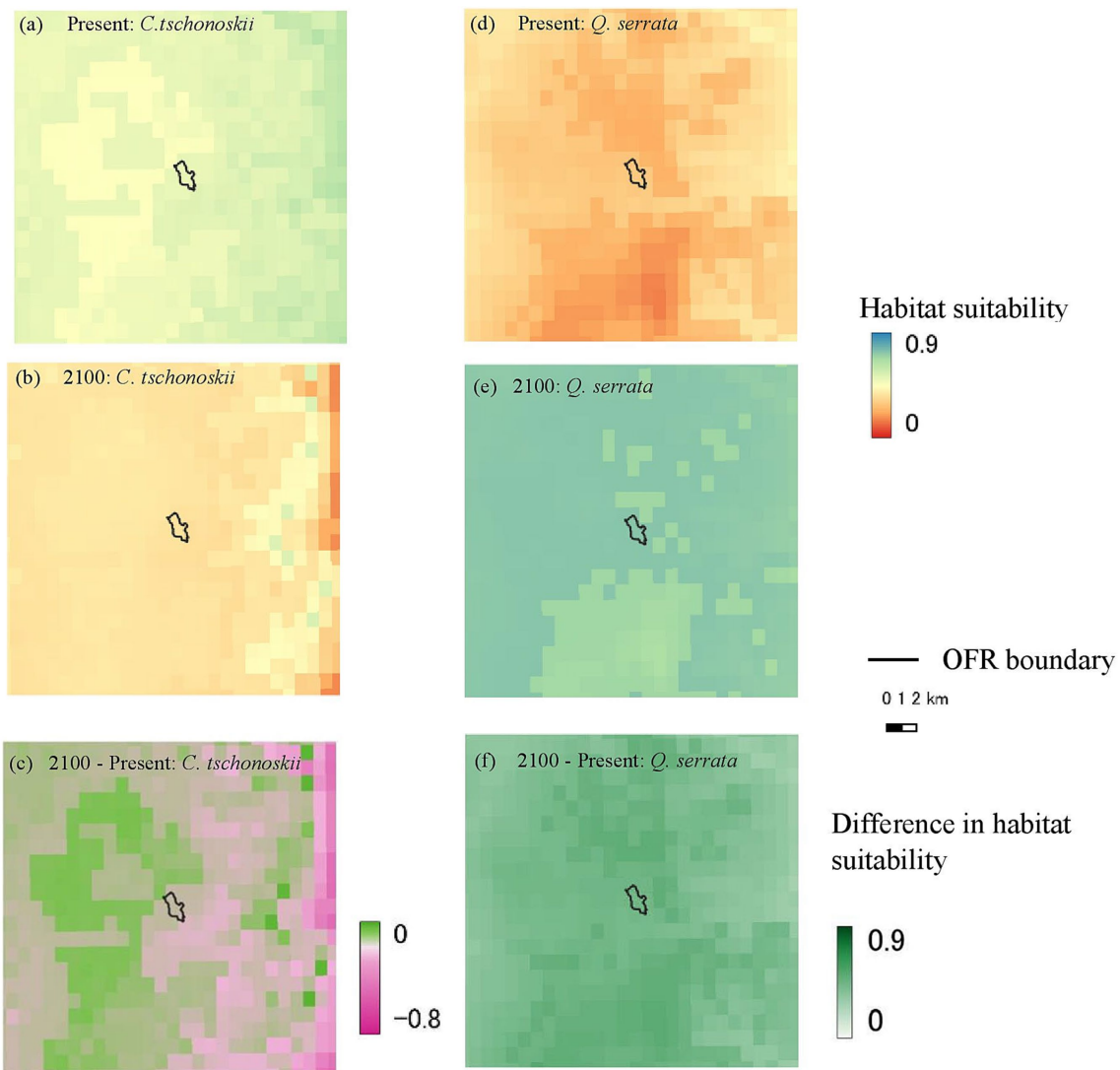


FIGURE 4

Habitat suitability and difference of habitat suitability maps in and around OFR created using Maxent under the current climate and predicted climate change (SSP5-8.5 pessimistic prediction characterized by high population growth and high levels of greenhouse gas concentrations by the end of 2,100) in 2100 for two species: *C. tschonoskii* and *Q. serrata*.

tangible advantages in the future (Rehfeldt et al., 2014). The findings in Appendix 2 can help identify suitable conservation strategies for the tree species examined here. For example, for the subcanopy species that are expected to undergo considerable loss in the future, efforts could focus on facilitating range expansion to the northwest of OFR, which may enhance their stability facing the consequences of rapid climate change. Furthermore, the northern range limits of evergreen broadleaved tree species are projected to expand to higher elevations and northwards into areas dominated by *Quercus* (Nakao et al., 2011; Matsui et al., 2018). Our study also suggests that the Fagaceae will show resilience and expand their distributions, particularly *Q. serrata*, and especially under extreme climate conditions in 2100.

The species *C. controversa* and *C. crenata* had lower AUC values, which is a study limitation. *C. controversa* is a fast-growing, shade-intolerant pioneer species that may have lower correlations with climate variables. *C. crenata* is a valuable food resource related

to agroforestry and it would be affected by past socioeconomic activities rather than bioclimate variables (Freitas et al., 2022). Therefore, it is important to note that the distribution changes are not only affected by climate factors but also by socioeconomic factors such as anthropogenic disturbances (Koyama et al., 2021), and land use history, which could have a lasting impact on forest resilience. For species such as *C. controversa* with potentially long seed dispersal distance (Wijenayake et al., 2024), the incorporation of spatially dynamic processes such as seed dispersal models should also be important to improve model prediction from landscape perspective. Long-distance dispersal of seeds significantly influences plant population dynamics via enabling vulnerable species to colonize more suitable distant habitats; however, our understanding of its implications at a landscape scale remains limited. For this reason, we excluded seed dispersal factors to minimize complexity. Climate change will increase disturbances



in forests, but these disturbances should not be seen only as a threat to stability. They have the potential to enhance tree species diversity, with positive impacts on productivity and other ecosystem functions (Silva Pedro et al., 2016), while leading to changes in habitat quality (Hovenden and Williams, 2010). These vulnerabilities are not necessarily a realistic picture of the area around OFR. Note, however, that the future changes in this area may not be as severe due to topographic features. The current dominant tree species will remain similar, although their relative dominance will change with the climate.

To verify the generality of findings from previous related studies, we figured out sharply contrasting outcomes. Hansen et al. (2001) examined the potential impacts of human-induced climate change on forest biodiversity in the eastern U.S., revealing that habitat suitability for tree species varies based on their individual responses to environmental changes. Their study predicted a northward shift in habitat for many species, ranging from 100 to 530 km, with some species extending their optimal habitats beyond the U.S. border. This contrasts with our findings, where a high-resolution ecosystem-based climate envelope model predicts climate change impacts on forests in British Columbia, emphasizing climate as the primary driver of plant distribution (Hamann and Wang, 2006). By 2050, a significant proportion of present-day beech and sessile oak forests in zonal positions at low altitudes in Central Europe may fall outside their current bioclimatic niches. However, in our case the current dominant tree species are expected to remain, their relative dominance is likely to shift with changing climatic conditions. We believe that our study expanded the generality of knowledge of this field.

## 4.1 Future aspirations

We used a species distribution model (SDM) approach to analyze the compositional changes in species of old-growth forest landscapes under future climate conditions, but the accuracy of the projected changes needs to be checked. It is necessary to continue monitoring actual changes of OFR and adjust the model accordingly to improve the model prediction. Furthermore, this study did not consider surrounding land use. In OFR, the primary landscape transitioned from expansive grasslands and broadleaved forests to a mosaic of fragmented secondary forests and coniferous plantations (Miyamoto et al., 2011). The canopy trees in plantations remain evergreen, creating consistently darker conditions compared to the varying light conditions in natural old-growth forests, where the light conditions change seasonally (Gonzales and Nakashizuka, 2010). This difference would be a barrier to regeneration by limiting recruitment, particularly for canopy species. It is ideal to integrate studies of the seed dispersal distances of each species to detect barriers to regeneration (Wijenayake et al., 2024). Furthermore, for local forest planning, it may be useful to designate corridors based on climate change scenarios and land-use changes. It is essential to explore strategies that balance biodiversity conservation with climate change mitigation and to validate their effectiveness (Hirata et al., 2024). By integrating models that connect the economy, land use, and biodiversity, we can understand how different socioeconomic factors affect biodiversity directly or indirectly (Ohashi et al., 2019). Incorporating nature-based solutions

is essential for understanding the impacts of climate change on species distributions and forest ecosystem dynamics.

## 5 Conclusion

This study demonstrated the utility of species distribution models for addressing knowledge gaps regarding changes in major tree species under varying climate change scenarios. We project that the canopy species *Q. serrata* will become more dominant than other canopy species by 2,100 under the baseline climate scenario, potentially leading to shifts in the species composition of the OFR. Consequently, the stability of forest ecosystems and local species diversity may be compromised under future climate change scenarios. This information is crucial for developing vegetation monitoring, ecosystem management, and climate change adaptation pathways. Further refinements should consider factors such as topography and land use changes in the surrounding landscape.

## Data availability statement

Publicly available datasets were analyzed in this study. This data can be found here: Japan Biodiversity Mapping Project (J-BMP) database.

## Author contributions

PW: Conceptualization, Data curation, Formal analysis, Methodology, Visualization, Writing – original draft. TS: Data curation, Methodology, Software, Validation, Writing – review & editing. AH: Methodology, Software, Validation, Writing – review & editing. TeM: Methodology, Software, Validation, Visualization, Writing – review & editing. YK: Data curation, Resources, Validation, Writing – review & editing. TaM: Conceptualization, Funding acquisition, Investigation, Project administration, Resources, Supervision, Writing – review & editing.

## Funding

The author(s) declare that financial support was received for the research, authorship, and/or publication of this article. The JSPS KAKEN supported this work under grant 21H04946.

## Acknowledgments

The authors thank Dr. Kazuaki Tsuchiya, for his various support, guidance, and encouragement.

## Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

## Generative AI statement

The authors declare that no Gen AI was used in the creation of this manuscript.

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