

Indirect effects mediate direct effects of climate warming on insect disturbance regimes of temperate broadleaf forests in the central U.S.

Supplementary Materials

Appendix S1. Modeling the vegetation dynamics

We used the LANDIS PRO Succession module to simulate vegetation dynamics. Specifically, we used the LINKAGES 3.0 ecosystem process model (Dijak et al., 2017) to estimate species establishment probability (SEP) and maximum growing space (MGSO) for each land type under different climate change scenarios. Land types are landform units that stratify the heterogenous landscape. We used the SEPs and the MGSOs estimated by LINKAGES 3.0 as model parameters in LANDIS PRO to encapsulate the effects of climate change on vegetation (Fig. 2; Wang et al., 2014; Dijak et al., 2017). In LANDIS PRO, SEP together with species shade tolerance and available growing space determines seedling colonization. Competition is initiated once the MGSO is reached, and competition-caused mortality is simulated by using Yoda's self-thinning theory (Yoda, 1963).

Appendix S2. Modeling the effects of land type and other types of disturbance on insects

The susceptibility of a given site can be adjusted to reflect variation in the susceptibility introduced by site environment (e.g., top-ridges and south-facing slopes with low soil moisture increase site susceptibility to insect outbreaks) and recent disturbance (e.g., recent defoliation or fire may weaken the stressed host). We used the original design of biotic disturbance agent (BDA) proposed by Sturtevant et al. (2004) to adjust susceptibility based on land type modifiers (LTMs) and disturbance modifiers (DMs). The values of LTMs and DMs range between -1 and 1 (the operator sign depended on how DMs and LTMs affect the site susceptibility to insect outbreak) and are added to the site susceptibility index S of all affected sites. It is assumed that the LTMs are constant in the entire simulation, while the DMs with duration decline linearly since last disturbance as shown in Eq. S1:

$$DM_{dist}(t) = DM_{max,dist} \times \frac{DM_{duration,dist} - t_{dist}}{DM_{duration,dist}} \quad \text{Eq. S1}$$

in which $DM_{max,dist}$ is the maximum modifier for the disturbance ($dist$); $DM_{duration,dist}$ is the duration of the disturbance ($dist$), t_{dist} is the time since last disturbance experienced on that site. Disturbances included fire, wind, and other BDA agents. Thus, the site susceptibility index S is modified by LTM and the sum of all DMs as Eq. S2:

$$S_m(t) = S(t) + LTM + (DM_{dist1}(t) + DM_{dist2}(t) + \dots) \quad \text{Eq. S2}$$

Appendix S3. Modeling insect outbreak probability and mortality

Biotic disturbance probability (BDP) was calculated as follows:

$$BDP = \alpha \times \left(\frac{S_m + NE \times W}{1 + W} \right) \times \left(\frac{R + V}{4} \right) \quad \text{Eq. S3}$$

in which α is a user-defined calibration parameter; S_m is the modified susceptibility. NE is the neighborhood effects calculated from the mean S_m of each cell within a user-defined radius by using one of the three radial distance weighting functions: uniform, linear, and Gaussian. NE is calculated for all cells containing host species, W is a parameter designed to define the relative importance between S_m and NE . R is defined as the regional outbreak status ranging from 0 (no outbreak) to 3 (severe outbreak). The severity is estimated based on local outbreak records (Haavik & Stephen 2010). V is the population-level voltinism that can influence the outbreak status. R and V together determine the outbreak status of the entire region. To assess the population development of the temperature-dependent woodborer, we calculated the voltinism—the potential number of generations per year. The voltinism V affects the severity of insect disturbance by changing the value of BDP. Specifically, V is a function of the climate variable (annual degree day sum > 5.5°C) ranging from 0.5 to 1 in the BDP equation. $V=0.5$ indicates poor condition for red oak borer (ROB) population development and the population consists of semivoltine ROB. $V=1$ means the population consists of univoltine borers. Most of the time, borer population consist of both semi- and univoltine borers ($0.5 < V < 1$), the majority of the population depend on the microclimate that an individual borer experienced during its larval stage. The V was derived from and calibrated with field studies and experts' opinions (Aquino et al., 2008).

Therefore, BDP is an emergent property of forest susceptibility, neighborhood effects, insect population dynamics, and regional outbreak status. ROB disturbance severity is a direct

function of BDP in which $BDP < 0.33$ is slight disturbance; $0.33 < BDP < 0.67$ is moderate disturbance; $BDP > 0.67$ is severe disturbance (Sturtevant et al. 2004). Different disturbance severity causes different cohort mortality, and the mortality probabilities (MPs) of host tree species were derived from empirical field studies (Fan et al. 2008). The mortality at an outbreak site is determined by species' age and host susceptibility classes (Table S1). The susceptibility of each cohort to insect-induced mortality was derived from empirical field studies and experts' opinions (Hay, 1969, 1974; Fan et al., 2008).

Table S1. Host susceptibility classes and mortality probability (MP).

Species	Susceptibility class 3 age (MP)	Susceptibility class 2 age (MP)	Susceptibility class 1 age (MP)
Black oak	10 (0.5)	20 (0.4)	30 (0.08)
Northern red oak	10 (0.6)	20 (0.5)	30 (0.1)
Southern red oak	10 (0.5)	20 (0.33)	30 (0.05)
White oak	20 (0.3)	30 (0.2)	40 (0.02)
Post oak	20 (0.2)	30 (0.2)	40 (0.02)

Appendix S4. BDA diagram

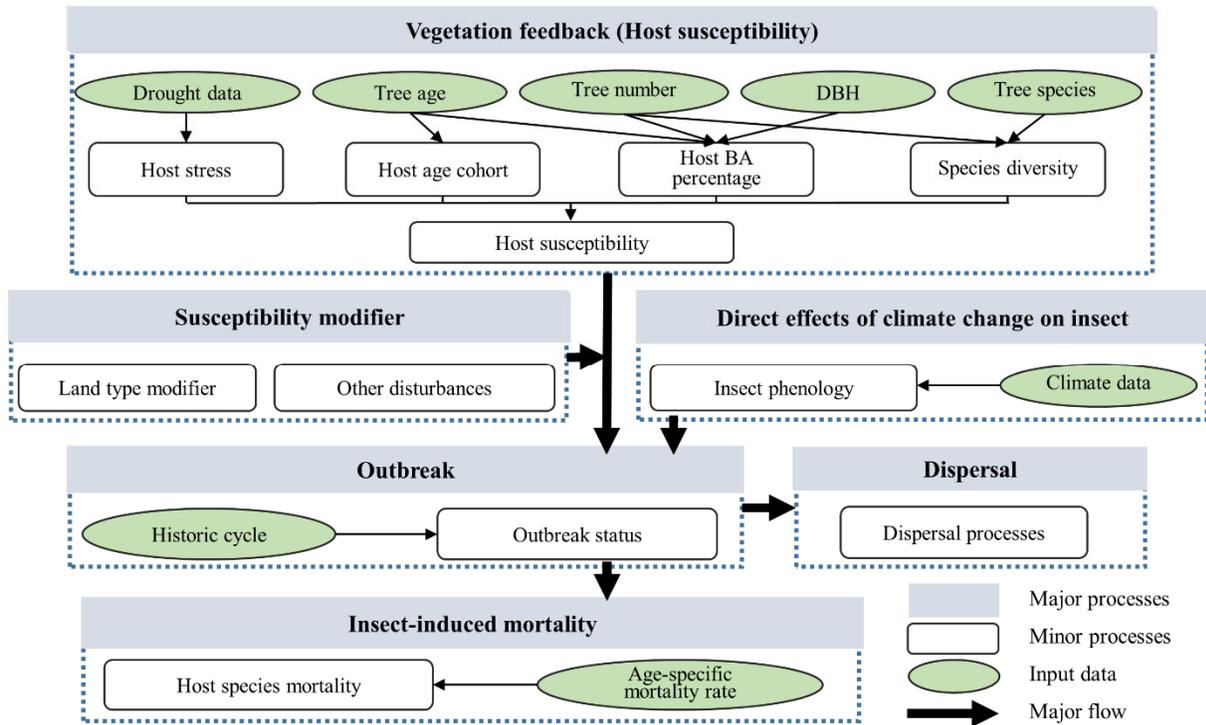


FIGURE. S1 Diagram of the updated BDA module in LANDIS PRO. DBH is tree diameter at breast height (meter), and BA is the basal area (m^2/ha).

Appendix S5. Model parameterization, calibration, and validation

Species' longevity, maturity, shade tolerance class, seedling dispersal distance, sprouting probability, maximum stand density index, and maximum diameter at breast height (DBH) were parameterized based on previous studies and literature (Burns & Honkala, 1990; Wang et al., 2018) (Table S S2, S3, S4). The initial forest composition map in 2000 for the LANDIS PRO was directly derived from 1995-2005 Forest Inventory and Analysis (FIA) data. We used Landscape Builder to stochastically assign a representative FIA plot to each raster cell based on size class, land cover, and landform (Dijak, 2013). Each raster cell on the map contained the initial tree species distribution (absence/presence), abundance (number of trees and DBH by age cohort), and land type (e.g., forest, urban, water body, and agricultural land) (Wang et al., 2014).

The majority of the model calibration was done in previous studies (e.g., Wang et al., 2018, 2019). A data splitting approach was employed to utilize FIA data for both calibration and validation of forest growth and succession (Wang et al., 2014). Discrepancies between model predictions and FIA data at the landscape scale were relatively small: both mean error (ME) and relative root mean square error (RMSE) were less than 10%. Changes in forest basal area, tree density, and quadratic mean diameter in years after simulated harvest were compared with field studies and were in general agreement (Fraser, He, Shifley, Wang, & Thompson 2013). We compared the predicted basal area and tree density from selected raster cells to the observed values in FIA plots and compared the landscape-scale predictions stratified by land type (using 223A Ozark Highlands Section as an example) (Fig. S2 a, b). We compared predicted ROB disturbed area to the dataset of field-based Insect and Disease Detection Survey (IDS) and field studies in Missouri and Arkansas from 2000-2010 to validate the BDA (Fierke et al., 2007; Haavik & Stephen 2010; Fan et al., 2008). Due to the limited temporal and spatial field data of

ROB disturbance, we only compared the simulation values with the observed values in the ecological section or subsection where the data were available. Specifically, we compared the predicted ROB disturbed area (\pm SD) against observed values of 223Aa St. Francois Knobs and Basins Subsection during 2006-2010, 223Ae Meramec River Hills Subsection during 2006-2010, 223Ad Gasconade River Hills Subsection during 2006-2010, and M223A Boston Mountain Section during 2000-2005 (Fig. S2c).

Table S2 Biological traits of individual tree species and parameters used in the forest dynamic landscape model LANDIS PRO in the central US.

Spp. common name	Mature/longevity	Shade tolerance	Fire class	Max dispersal distance (m/year)	Vegetative probability	Min/max sprouting age	Max DBH (cm)	Max SDI (trees/ha)
Sugar maple	20/300	5	5	2160	0.8	20/80	75	570
Shortleaf pine	20/200	2	4	1350	0.8	10/30	70	990
Loblolly pine	20/200	3	4	1350	0.8	10/30	70	1100
Red maple	10/150	4	1	2160	0.9	10/100	65	700
Mockernut hickory	30/200	3	1	1080	0.6	20/200	65	570
White ash	30/250	3	4	1620	0.6	10/150	65	570
Sweetgum	30/200	2	2	1620	0.6	20/200	65	570
Yellow poplar	20/200	2	2	2160	0.5	20/150	70	700
Black cherry	20/150	2	1	1080	0.8	20/100	65	570
White oak	40/300	4	3	1080	0.7	10/100	75	570
Southern red oak	30/200	3	2	1080	0.7	10/100	65	570
Northern red oak	30/200	3	2	1080	0.7	10/120	65	570
Post oak	40/250	3	3	1080	0.6	10/100	70	570
Black oak	20/150	3	2	1080	0.6	10/100	65	570
American elm	40/200	3	2	1620	0.7	10/60	60	900

Species' scientific names are listed in Table S3.

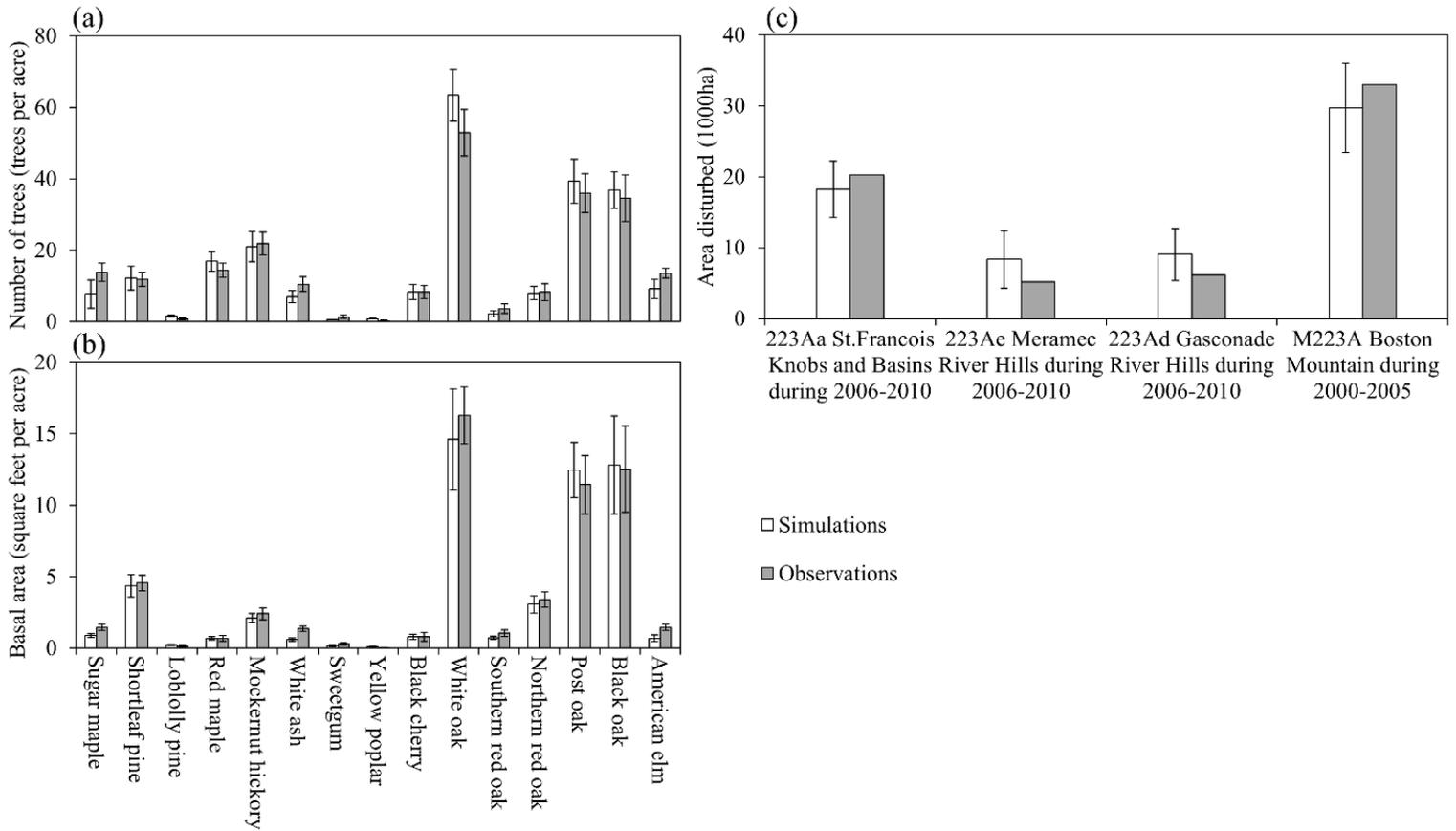


FIGURE. S2 Comparison of the predicted number of trees (a) and basal area (b) by species from the LANDIS PRO model against observed values of 233A Ozark Highlands Section in 2010. Comparison of predicted ROB disturbed area (\pm SD) against observed values of 223Aa St. Francois Knobs and Basins Subsection during 2006-2010, 223Ae Meramec River Hills Subsection during 2006-2010, 223Ad Gasconade River Hills Subsection during 2006-2010, and M223A Boston Mountain Section during 2000-2005 (c).

Appendix S6. Temporal changes in mean annual temperature

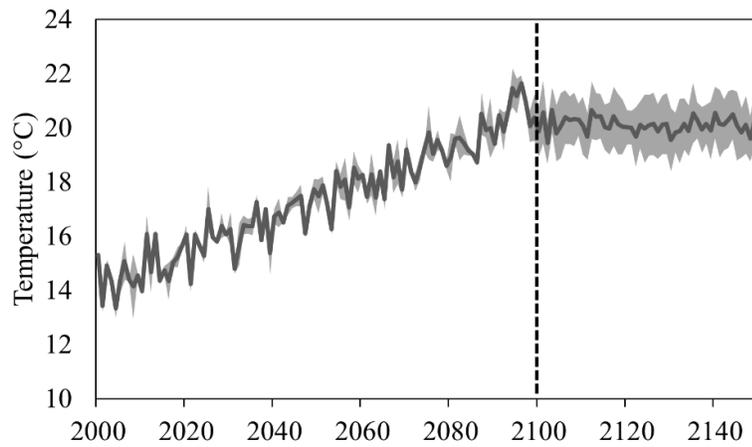


FIGURE. S3 Temporal changes in mean annual temperature under ACCESS1-0 scenario from 2000 to 2099. From 2100 to 2150, the mean annual temperature was randomly sampled from 2080-2099 based on the zero-order approximation that climate will stabilize toward 2150.

Appendix S7. Sensitivity analysis for the insect model

We used three general circulation models (GCMs) (ACCESS1.0, CanESM2, and GFDL-ESM2M) in the RCP 8.5 emission scenario used in the IPCC Fifth Assessment Report as climate change scenarios and employed one historic climate observation (1980-2009) as the baseline climate scenario to conduct the sensitivity analysis. Compared to the baseline, the three GCMs projected mean annual temperature in 2070-2100 increased by 5.6°C, 4.8°C, and 3.4°C respectively. There was a great variation in region-wide precipitation projections among these three GCMs. Precipitation on average decreased 40 mm in ACCESS1-0, and increased 60 and 26 mm in CanESM2 and GFDL-ESM2M, respectively. We set the model stimulating for 300 years from 2000 to 2300 and presented the mean values of voltinism, total susceptibility, and primary host biomass under each climate condition (Fig. S4). The results showed that the predictions of our insect module could capture the variation among different climate scenarios.

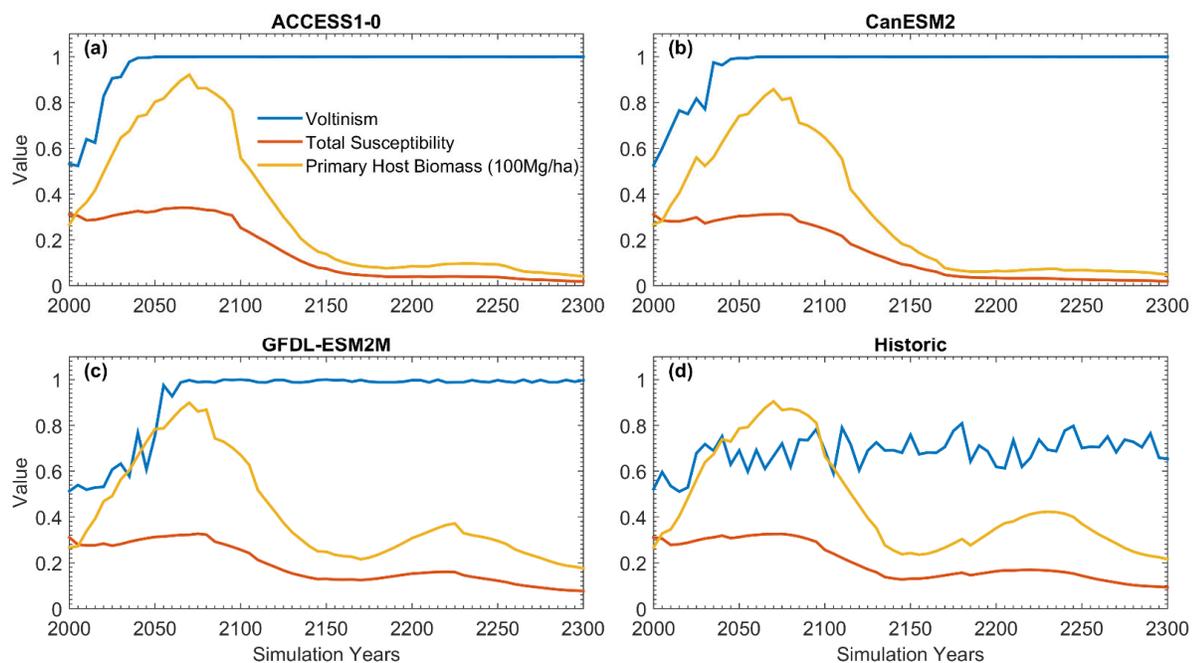


FIGURE. S4 The mean values of voltinism, total susceptibility, and primary host biomass under four climate scenarios from 2000 to 2300 in the Central Hardwood Forests, U.S.

Appendix S8. Species host type, common name, and scientific name

Table S3 Species host type, common name, and scientific name

Host type	Common name	Scientific name
Non-host	Sugar maple	<i>Acer saccharum</i> Marshall
Non-host	Red maple	<i>A. rubrum</i> L.
Non-host	Loblolly pine	<i>Pinus taeda</i> L.
Non-host	Shortleaf pine	<i>P. echinata</i> Mill
Non-host	Mockernut hickory	<i>Carya tomentosa</i> Nutt.
Non-host	White ash	<i>Fraxinus americana</i> L.
Non-host	Sweetgum	<i>Liquidambar styraciflua</i> L.
Non-host	Yellow poplar	<i>Liriodendron tulipifera</i> L.
Non-host	Black cherry	<i>Prunus serotina</i> Ehrh.
Secondary host	White oak	<i>Quercus alba</i> L.
Primary host	Southern red oak	<i>Q. falcate</i> Michx
Primary host	Northern red oak	<i>Q. rubra</i> L.
Secondary host	Post oak	<i>Q. stellate</i> Wangenh.
Primary host	Black oak	<i>Q. velutina</i> Lam
Non-host	American elm	<i>Ulmus americana</i> L.

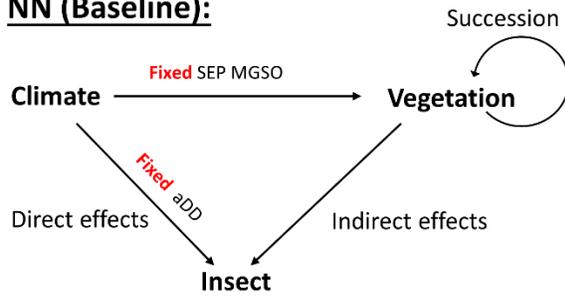
Appendix S9. Biological traits of individual tree species used in LINKAGES

Table S4 Biological traits of individual tree species and parameters used in LINKAGES 3.0 in the central U.S. (Drought tolerance: the maximum proportion of growing season that species can withstand drought; Frost tolerance: minimum January temperature species can withstand)

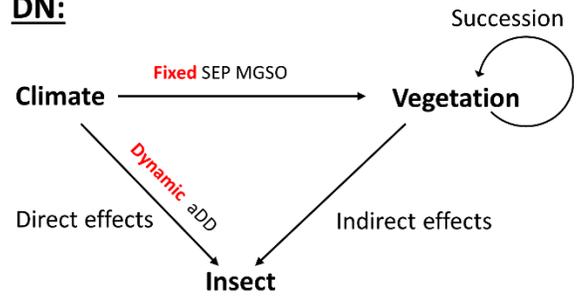
Sp. Common name	Growing degree days		Growth parameters		Drought tolerance	Frost tolerance	Parameters to calculate nitrogen growth factors					Leaf weight
	maximum	minimum	b3	b2			cm1	cm2	cm3	cm4	cm5	
Sugar maple	3528	984	0.1272	38.17	0.08	-25	2.94	117.52	0.00234	-1.2	1.3	440
Shortleaf pine	5314	1892	0.2863	57.26	0.423	-11	2.79	219.77	0.00179	-0.6	1	440
Loblolly pine	6090	2714	0.2912	58.23	0.36	-6	2.79	219.77	0.00179	-0.6	1	440
Red maple	6891	900	0.2863	57.26	0.23	-25	2.79	219.77	0.00179	-0.6	1	440
Mockernut hickory	5400	1900	0.2663	53.26	0.425	-6	2.94	120	0.002	-0.9	1.2	248
White ash	5866	1056	0.2863	57.26	0.28	-23	2.99	207.43	0.00175	-5	2.9	440
Sweetgum	6142	2100	0.2152	53.81	0.3	-9	2.99	207.43	0.00175	-5	2.9	440
Yellow poplar	6000	1588	0.1495	44.84	0.16	-16	2.99	207.43	0.00175	-5	2.9	440
Black cherry	8258	1169	0.2863	57.26	0.3	-21	2.99	207.43	0.00175	-5	2.9	173
White oak	5314	1169	0.3363	67.26	0.33	-19	2.94	117.52	0.00234	-1.2	1.3	440
Southern red oak	5800	2100	0.3363	67.26	0.423	-9	2.94	117.52	0.00234	-1.2	1.3	440
Northern red oak	4766	984	0.2863	57.26	0.225	-25	2.94	117.52	0.00234	-1.2	1.3	440
Post oak	6142	1892	0.4201	63.01	0.555	-11	2.79	219.77	0.00179	-0.6	1	440
Black oak	5435	1437	0.2863	57.26	0.3	-15	2.79	219.77	0.00179	-0.6	1	440
American elm	2200	279	0.2912	58.23	0.401	-22	2.87	180	0.00207	-0.9	1.2	440

Appendix S10. Experimental design

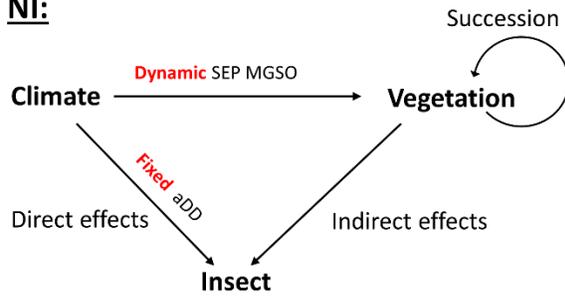
NN (Baseline):



DN:



NI:



DI:

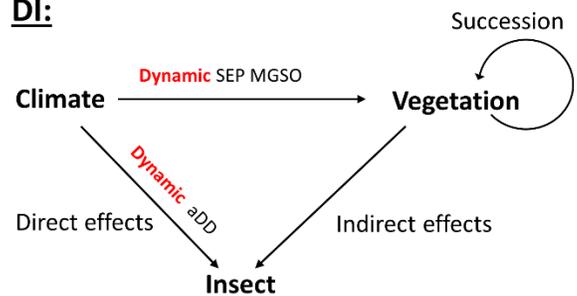


FIGURE. S5 The simulation design for NN, DN, NI, and DI scenarios. Specifically, simulations in the NN scenario were based on baseline climate with fixed forest growth parameters (i.e., SEP and MGSO) and no changes in the direct effects on ROB; simulations in DN scenario in with fixed forest growth parameters while the direct effects on ROB were updated according to RCP 8.5; simulations in the NI scenario included dynamic growth parameters according to RCP 8.5 while no change in the direct effects on ROB and simulations in the DI scenario with both dynamic changes in growth parameters and the direct effects on ROB.

Appendix S11. Harvest and wildfire

We simulated the current tree harvest regime with the LANDIS PRO Harvest Module (Fraser et al., 2013). We used US Forest Inventory and Analysis (FIA) units as the management units to capture the variations in harvest practices across the regions by simulating different harvest strategies for each unit. We simulated two types of harvest in each management unit consisting of high-grading and clear-cutting. We varied the percentage of the unit harvested and the preferred species for harvest to capture similar removals to those reported since 1995–2015 (Fraser et al., 2013). Also, we simulated the current natural fire regimes by using the LANDIS PRO Fire Module (Fraser et al., 2019). We did not include any direct effects of climate change on fire regimes in this study. We parameterized the size and frequency of wildfire based on fire records from 1980 to 2014 from LANDFIRE (2012) and Monitoring Trends in Burn Severity (Eidenshink et al., 2007).

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