









State and Tribal Capacity Building on Forest Carbon

Forest Carbon and Climate Change in Ohio

This technical briefing summarizes topics such as forest densities and cover types, carbon storage, and climate considerations for the state of Ohio.

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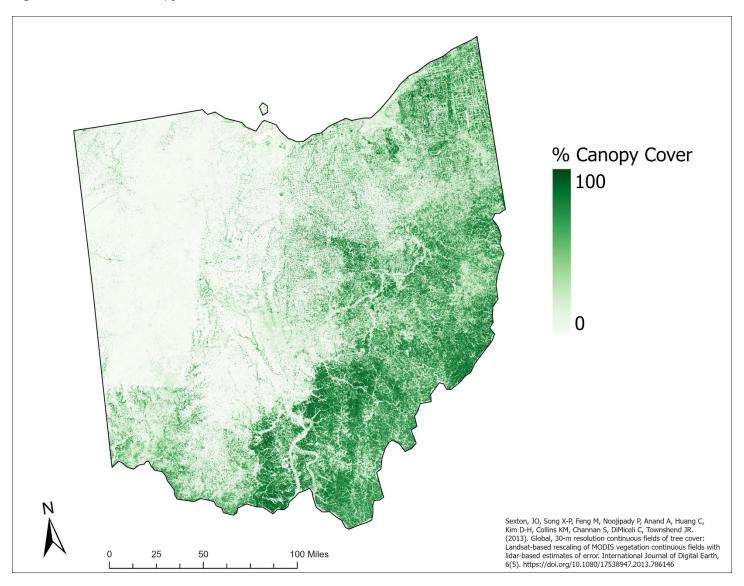
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Ohio Forest Overview

Ohio is situated in the Midwest region of the United States and lies within the US Forest Service's Eastern Region (USFS Region 9). Bordering states include Indiana to the west, Kentucky and West Virginia to the south, Pennsylvania to the east, and Michigan to the northwest, with Lake Erie marking Ohio's northeastern boundary.

A map of percent tree canopy cover in Ohio is shown in **Figure 1**. This state has relatively low forest coverage across much of its extent. However, the southeastern portion of the state has higher levels of canopy cover. This region coincides with the location of protected lands including several state forests and the Wayne National Forest.

Figure 1. Percent tree canopy cover in Ohio.

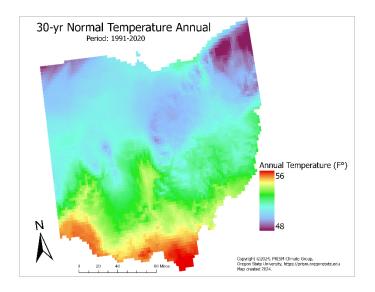


Temperature and Precipitation

Two major factors affecting forest carbon and productivity are temperature and precipitation. **Figure 2** shows normal mean temperatures throughout Ohio between 1991 and 2020. Over this 30-year period, mean annual temperatures varied by about 8 °F across this state. Temperature trends largely follow latitudinal gradients, with warmer mean temperatures occurring in the southernmost portions of the state and giving way to cooler temperatures to the north. The warmest mean annual temperature is around 56 °F and occurs in the central area of Ohio's southern border, while the coolest mean annual temperature is around 48 °F found primarily in the northeast corner of the state.

Figure 2. Normal mean temperature (°F) from 1991–2020 in Ohio.





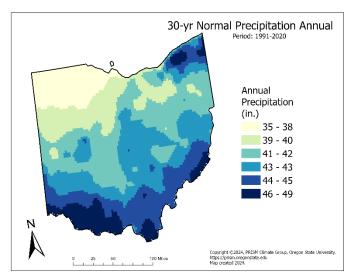
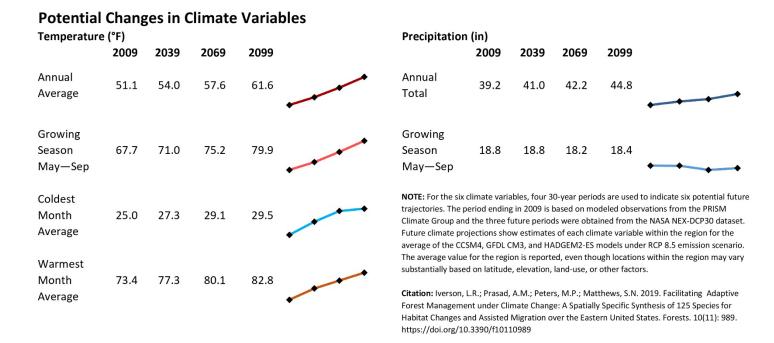


Figure 3 shows normal mean precipitation throughout Ohio between 1991 and 2020 and demonstrates the geographic variation in these trends. Over this 30-year period, mean annual precipitation levels varied by about 14 in. The area with the lowest levels of precipitation (35-38 in.) is the northwest corner of the state. Areas receiving the highest amounts of precipitation (46-49 in.) occur along Ohio's southern border as well as in the northeastern portion of the state.

Projected Future Trends in Temperature / Precipitation

Figure 4. Model results for potential changes in temperature and precipitation trends in Ohio through 2099 under a high emission scenario (RCP 8.5).



Projected future trends in temperature and precipitation for Ohio between 2009 and 2099 are shown in **Figure 4**. Model results suggest average temperatures will continue to increase through the end of the century, a trend which is also projected for the coldest and warmest month averages, as well as throughout the growing season (May – Sep.). Over this 90-year period, average annual temperatures are expected to increase by an estimated 10.5 °F, with the most drastic increases expected to occur during the growing season (+12.2 °F).

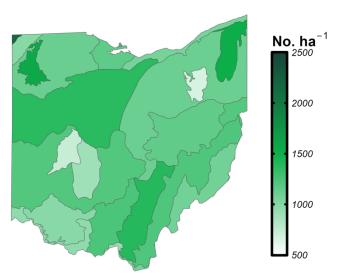
Model results of future precipitation in Ohio follow variable trends, with totals projected to steadily increase through 2099 (**Figure 4**). Over a 90-year period, annual precipitation is expected to increase by an estimated 5.6 in., however, precipitation levels are projected to *decrease* during the growing season by an estimated 0.4 in. This suggests that precipitation in Ohio may increase substantially during the winter months (Oct. – Apr.), while drought events may become more frequent and severe during the growing season.

Forest Density

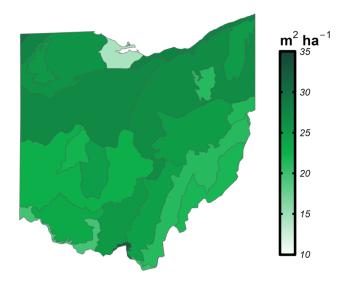
Figure 5. Forest density as live tree density (No. ha-1) in Ohio.

Figure 6. Forest density as live tree basal area (m² ha-1) in Ohio.

Forest Density: Live tree number



Forest Density: Live tree basal area



Forest density¹ is both a structural characteristic of forests and a reflection of forest dynamics. It can be measured as the number of trees per unit area, or it can be measured in terms of live tree area per unit area, known as "basal area". Live tree basal area represents the amount of ground covered by living trees in two-dimensional space. **Figure 5** shows average forest density in terms of live trees per hectare by ecosection² across the state of Ohio, while **Figure 6** represents forest density by ecosection in terms of basal area (m² ha⁻¹).

By comparing these figures we can see that a small inland ecosection near the northwest corner of the state has a higher forest density than surrounding ecosections in terms of number of trees per hectare (**Figure 5**), but a similar density to these ecosections in terms of basal area (**Figure 6**). This suggests that in this ecosection, there may be more total trees per unit area, but on average, these trees tend to be smaller. Meanwhile, an inland ecosection in the northeastern portion of Ohio has a lower forest density than neighboring ecosections in terms of both number of trees and basal area, suggesting a lower overall forest density in this zone.

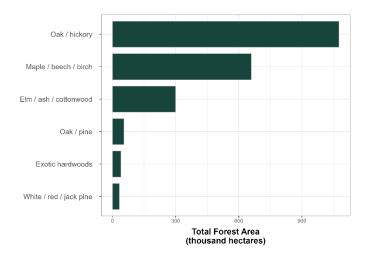
¹ All forest inventory and carbon data were estimated using data from the Forest Inventory and Analysis (FIA) Program which can be accessed through the FIA DataMart (USDA Forest Service, 2024. *Forest inventory and analysis program.* Available at: https://www.fia.fs.usda.gov/) using the rFIA package (Stanke et al, 2020. rFIA: an R package for estimation of forest attributes with the US Forest Inventory and analysis database. *Environ Model Softw.* 127:104664. https://doi.org/10.1016/j.envsoft.2020.104664) in the R programming environment (R Core Team, 2020. *R: A language and environment for statistical computing*, Vienna, Austria: R Foundation for Statistical Computing.

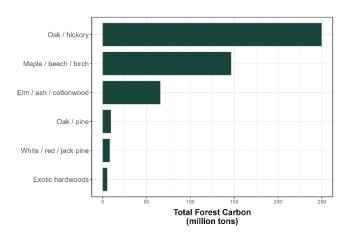
²Ecosection definition can be found at Cleland et al, 2007. Ecological Subregions: Sections and Subsections for the conterminous United States. General Technical Report WO-76D, Washington Office, USDA Forest Service. https://doi.org/10.2737/WO-GTR-76D

Forest Cover Types and Carbon

Figure 7. Total forest area (thousand ha) by forest type in Ohio.

Figure 8. Total forest carbon (million tons) by forest type in Ohio. Total forest carbon is the sum of carbon stored across all aboveground and belowground pools (includes Soil Organic carbon + Live Belowground carbon + Live Aboveground carbon + Litter carbon + Dead wood carbon).



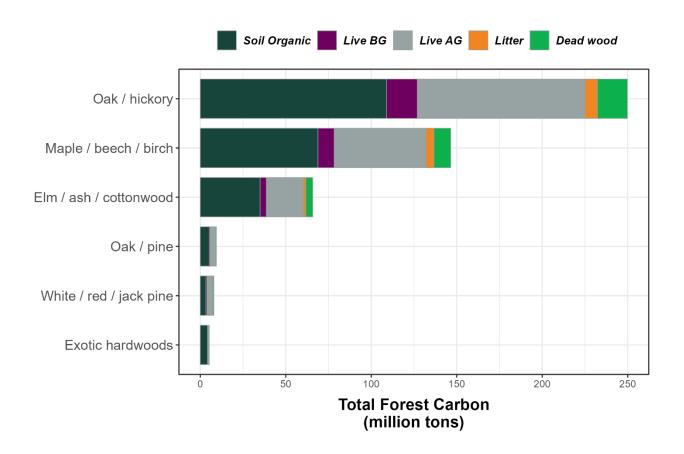


Ohio is dominated by 6 key forest cover types: Oak / hickory, Maple / beech / birch, Elm / ash / cottonwood, Oak / pine, Exotic hardwoods, and White / red / jack pine. **Figure 7** and **Figure 8** show state-level data of total forested area and total forest carbon, respectively, for each of these cover type groups. As these figures show, Oak / hickory is the dominant forest type of Ohio, spanning an area upwards of 1 million hectares and storing 250 million tons of carbon statewide. With coverage levels ranging from <50,000-650,000 hectares, other forest types in this state are less abundant, yet play an important role contributing to enhanced biodiversity and landscape heterogeneity. Comparing trends from **Figure 7** with those in **Figure 8** demonstrates how carbon storage levels vary by forest cover type. For example, Exotic hardwoods cover slightly more land area than White / red / jack pine stands in Ohio (**Figure 7**), yet when it comes to carbon, White / red / jack pine stands store slightly more carbon than their Exotic hardwoods counterparts (**Figure 8**).

³Forest Types are a classification of forest land based upon and named for the tree species that forms the plurality of live-tree stocking. These forest types used in the briefing align with FIA's definition of Forest type group which are a combination of forest types that share closely associated species and site requirements. Longer definitions of both forest types and forest type groups are found in Appendix D of the Forest Inventory and Analysis Database: Database Description and User Guide for Phase 2 (version 9.1) which can be accessed here: https://research.fs.usda.gov/sites/default/files/2023-11/wo-fiadb_user_guide_p2_9-1_final.pdf

Forest Carbon Pools

Figure 9. Total forest carbon (million tons) by pool and forest type in Ohio.

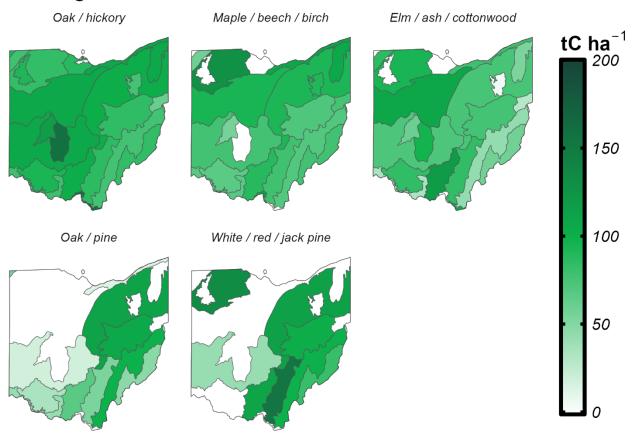


Forest carbon storage can be further assessed by examining how it's distributed across different ecosystem carbon pools. **Figure 9** shows the amount of carbon stored in different carbon pools of key forest cover types in Ohio. These values show how different forest types allocate distinct proportions of forest carbon into soil organic matter, live belowground (BG) biomass, live aboveground (AG) biomass, litter, and dead wood pools. For instance, forests composed of Maple / beech / birch, Elm / ash / cottonwood, and Exotic hardwoods allocate more ecosystem carbon to belowground pools (soil organic matter + live BG biomass), whereas forest types like Oak / hickory and White / red / jack pine tend to distribute stored carbon more evenly between aboveground and belowground pools. Another noteworthy trait shown in **Figure 9** is the magnitude of carbon storage levels across different pools and cover types. Oak / hickory's dominating presence on this landscape means its statewide carbon pools are outsized compared to other groups. For example, leaf litter and dead wood pools of Ohio's Oak / hickory forests on their own contain more stored carbon than the total ecosystem carbon (sum of carbon stored across all pools) contained by the Oak /pine, White / red / jack pine, or Exotic hardwoods groups.

Forest Carbon Density

Figure 9. Aboveground live forest carbon density (tC ha-1) by forest type in Ohio.

Average Forest Carbon Density by Ecosection: Aboveground Live



Forest carbon density can be influenced by many ecosystem traits, such as tree density, stand age, species mix/ cover type, soil fertility, elevation, and a site's management and disturbance history. In **Figure 9**, the carbon density of aboveground living forest biomass is shown for 5 key cover types in Ohio. Of these, Oak / hickory and White / red / jack pine stands hold the highest levels of aboveground live carbon per unit area, represented by the deep shades of green in specific ecosections in the southwest (Oak / hickory) and south-central (White / red / jack pine) portions of the state. Across much of their extent, Maple / beech / birch and Elm / ash / cottonwood stands exhibit relatively even carbon densities, while cover types like Oak / pine and White / red / jack pine show higher levels of variability across ecosections. In these instances, variable carbon densities can be driven by the relative prevalence or absence of each forest type from a given ecosection.

Species-Specific Considerations for Climate Adaptation

Climate change is expected impact the distribution of species into the future. Predictive modeling of potential future changes that incorporate species interactions, dispersal mechanisms, demography, physiology, and evolution is needed to assist in adaptive forest planning. The USDA Forest Service Climate Change Tree Atlas, Version 4, provides modeled potential suitable habitat for 125 species in the eastern US, with an additional 23 species. https://www.fs.usda.gov/nrs/atlas/tree/

Core Climate Change Atlas components:

- DISTRIB-II: Species habitat suitability model
- SHIFT: Migration model (when combined with DISTRIB-II, estimates colonization potential (HQCL) of future suitable habitats
- Adaptability Ratings: Species adaptability ratings (species traits not included in DISTRIB-II and SHIFT models)

In addition to the modeled potential suitable habitat for individual tree species, the Climate Change Atlas includes Current and potential future habitat, capability and migration for individual tree species and potential changes in climate variables summarized by the following spatial extents:

Geographic Area	Description
National Forest Summaries	Results summarized for 55 national forests
National Park Summaries	Results summarized for 78 national parks
HUC6 Watershed	Results summarized by hydrologic unit codes level 3 (HUC 6) which are hierarchical classifications based on surface hydrologic features in which level 3 maps watershed basins (Seaber et al, 1987) https://pubs.usgs.gov/wsp/wsp2294/
Ecoregional Vulnerability Assessments (EVAS)	Results summarized by ecoregions used in the USDA Climate Hub Regional Vulnerability Assessments https://www.climatehubs.usda.gov/assessments
USDA Forest Service EcoMap 2007 Sections	Results summarized by ecological sections that delineate ecosystems with distinctive vegetation and other unique ecological characteristics (Cleland et al, 2007, McNab et al, 2007)
National Climate Assessment (NCA) 2015 Regional Summaries	Results summarized by National Climate Assessment Region which include the Midwest, Northeast, Northern Plains, Southeast, and Southern Plains
1 x 1° Grid Summaries	Results summarized by 1x1° latitude and longitude
State Summaries	Results summarized for 38 states
Urban areas	Results summarized for 185 urban areas across the eastern US

Additional background on this tool can be found at: https://research.fs.usda.gov/centers/ccrc along with short video tutorials on the Climate Change Atlas website.

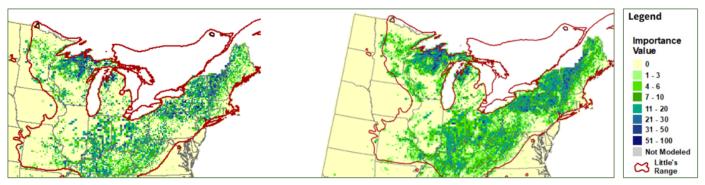
Habitat Suitability and Migration Models

Model Reliability: High

Key Species Example: Modeled potential suitable habitat for Sugar Maple (Acer saccharum) through 2100

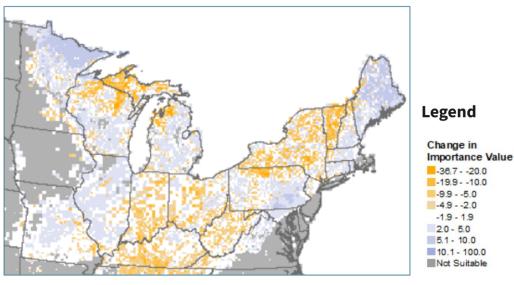
Current habitat quality and distribution (DISTRIB-II)

Potential migration (SHIFT) and colonization likelihood (CL)



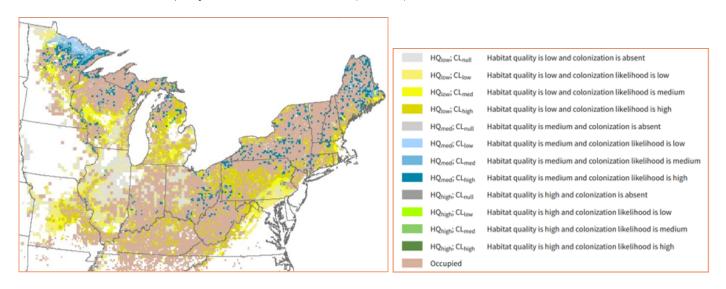
Importance value is a measure of abundance that accounts for both tree basal area and number of stems, ranging from 0-100.

Colonization potential of future habitats under a high emission scenario (RCP 8.5)



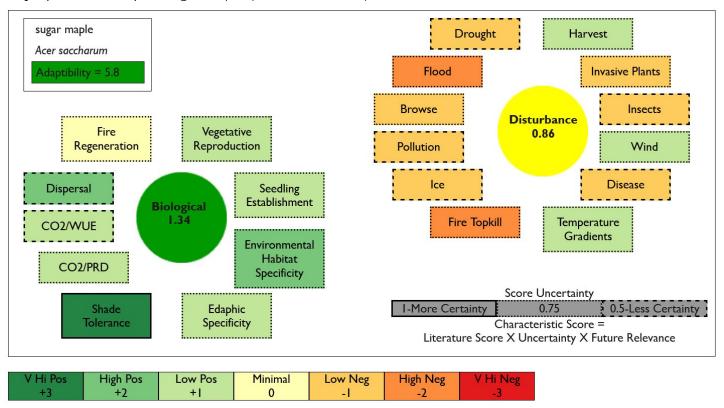
Colonization is limited to range margins and infill (Blue) which is derived from habitat quality (DISTRIB) and migration model (SHIFT) utilizing the colonization likelihood model (CL). Orange shading represents current species' distributions where abundance is predicted to decrease due to loss of habitat suitability.

DISTRIB-II + SHIFT: Habitat quality and colonization likelihood (RCP 8.5)



Adaptability Ratings

Key Species Example: Sugar Maple (*Acer saccharum*)



The Adaptability score, which assesses 21 variables to assign adaptability ratings to tree species in the eastern US, reflects a species' potential adaptability to climate change-driven stressors and disturbances at range wide scale. Adaptability ratings provide broad insights into factors that cannot be directly included in the Climate Change Tree Atlas species migration models. Two types of species traits are evaluated: 1) biological and 2) disturbance, each with their own set of factors to help characterize species' traits and responses to disturbance. Uncertainty is also included for each trait or factor assessed. When coupled with other modeled projections, adaptability ratings can support future planning under a changing climate.

The Adaptability variable is single score derived from the Modification Factors which encompass scores for the 12 disturbance and 9 biological factors. The Adaptability results can be considered relative to other tree species. For example, a species with a low Adaptability variable likely does not have life history characteristics to allow it to thrive under most conditions whereas a high Adaptability variable will likely do better under the climate change outputs from the DISTRIB-II and SHIFT Models.

Climate Change Atlas Summary for Sugar Maple

Sugar maple is widely distributed (21.3% of area), dense, and with high IV across much of the northern 2/3 of the Eastern US. It ranks fourth in overall abundance across the eastern US, behind loblolly pine, red maple and sweetgum. It rates as highly adaptable although under persistent drought or other stresses, it would likely decline. In contrast to our earlier models which showed substantial habitat decline in the south under harsh climate change, the species is modeled to decline only modestly, so we rate it with a very good capacity to cope, and to be a good infill species (according to SHIFT).

Citations:

Habitat suitability models on trees:

Peters et al. (2020). Climate change tree atlas, Version 4. U.S. Forest Service, Northern Research Station and Northern Institute of Applied Climate Science, Delaware, OH. https://www.nrs.fs.fed.us/atlas;

Iverson, L.R, Peters, M.P., Prasad, A.M., & Matthews, S.N. (2019). Analysis of Climate Change Impacts on Tree Species of the Eastern US: Results of DISTRIB-II Modeling. Forests, 10(4), 302. doi: 10.3390/f10040302 https://www.fs.usda.gov/treesearch/pubs/57857

Peters, M. P., Iverson, L. R., Prasad, A. M., & Matthews, S. N. (2019). Utilizing the density of inventory samples to define a hybrid lattice for species distribution models: DISTRIB-II for 135 eastern U.S. trees. Ecology and Evolution. doi: 10.1002/ece3.5445 https://www.fs.usda.gov/treesearch/pubs/58353

Iverson, L. R., Prasad, A. M., Peters, M. P., & Matthews, S. N. (2019). Facilitating Adaptive Forest Management under Climate Change: A Spatially Specific Synthesis of 125 Species for Habitat Changes and Assisted Migration over the Eastern United States. Forests, 10(11), 989. doi: 10.3390/f10110989 https://www.fs.usda.gov/treesearch/pubs/59105

Prasad, A. M., Iverson, L. R., Matthews, S. N., & Peters, M. P. (2016). A multistage decision support framework to guide tree species management under climate change via habitat suitability and colonization models, and a knowledge-based scoring system. Landscape Ecology, 31(9), 2187–2204. doi: 10.1007/s10980-016-0369-7 https://www.fs.usda.gov/treesearch/pubs/50748

Prasad, A. M., Gardiner, J. D., Iverson, L. R., Matthews, S. N., & Peters, M. (2013). Exploring tree species colonization potentials using a spatially explicit simulation model: implications for four oaks under climate change. Global Change Biology, 19(7), 2196–2208. doi: 10.1111/gcb.12204 https://www.fs.usda.gov/treesearch/pubs/43705

Iverson, L. R., A. M. Prasad, S. N. Matthews, and M. Peters. 2008. Estimating potential habitat for 134 eastern US tree species under six climate scenarios. Forest Ecology and Management 254:390-406. http://www.treesearch.fs.fed.us/pubs/13412

Adaptability of tree species:

Iverson, L. R., S. N. Matthews, A. M. Prasad, M. P. Peters, et al. (2012). Development of risk matrices for evaluating climatic change responses of forested habitats. Climatic Change 114(2): 231-243. doi: 10.1007/s10584-012-0412-x. https://www.fs.usda.gov/treesearch/pubs/41221

Matthews, S. N., L. R. Iverson, A. M. Prasad, M. P. Peters, and P. G. Rodewald. 2011. Modifying climate change habitat models using tree species-specific assessments of model uncertainty and life history factors. Forest Ecology and Management 262:1460-1472. http://www.fs.usda.gov/treesearch/pubs/38643

Climate summary definitions:

McNab, W.H.; Cleland, D.T.; Freeouf, J.A.; Keys, Jr., J.E.; Nowacki, G.J.; Carpenter, C.A., comps. 2007. Description of ecological subregions: sections of the conterminous United States [CD-ROM]. Gen. Tech. Report WO-76B. Washington, DC: U.S. Department of Agriculture, Forest Service. 80 p. https://research.fs.usda.gov/treesearch/48669

Cleland, D.T.; Freeouf, J.A.; Keys, J.E.; Nowacki, G.J.; Carpenter, C.A.; and McNab, W.H. 2007. Ecological Subregions: Sections and Subsections for the conterminous United States. Gen. Tech. Report WO-76D [Map on CD-ROM] (A.M. Sloan, cartographer). Washington, DC: U.S. Department of Agriculture, Forest Service, presentation scale 1:3,500,000; colored. https://research.fs.usda.gov/treesearch/48672

Seaber, Paul R., F. Paul Kapanos, and George L. Knapp (1987). Hydrologic Unit Maps. United States Geological Survey Water-Supply Paper 2294: i–iii, 1–63.