

# Measuring and modeling waterlogging tolerance to predict the future for threatened lowland ash forests

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**Abstract.** Emerald ash borer is an invasive pest causing widespread mortality of ash trees (*Fraxinus spp.*) across the U.S. Black ash is a major component of lowland hardwood forests within their range, and their extirpation has the potential to induce conversion of such forests to a non-forest state. Broad-scale models can help identify management strategies to maintain lowland ash ecosystems. Simulating lowland forest dynamics in landscape models has been problematic because lowland hydrology is extremely complex, making most hydrology algorithms intractable at landscape scale. A succession extension (PnET-Succession) of the LANDIS-II forest landscape model was recently updated to include simple algorithms to approximate lowland hydrology, but estimating parameters of tree species' waterlogging tolerance is difficult. We describe empirical experiments conducted to generate such estimates and illustrate their behavior in single-cell and landscape simulations. Simulated water stress mimicked two critical characteristics of the empirical experiment: 1) there was little difference in simulated stress variables between the well-drained and intermediate flooding treatments and 2) simulated water stress of species aligned with empirical waterlogging tolerance. We used the landscape model to scale the empirical experiment to landscape scales of space and time. When the simulation experiment was extended to 90 years, species productivity plateaued or peaked at a level that could be supported by the precipitation inputs and rooting zone depth. In a virtual experiment testing the competition outcomes between two species, the more waterlogging tolerant species did much better under the flooding treatment, but also tended to do better under the drained treatment because it never produced droughty conditions. When the updated waterlogging parameters were applied at landscape scale under future climate change and assisted migration (AM) scenarios, the mean biomass density of native species declined, and the introduced AM species increased as climate gradually changed and introduced cohorts thrived. Species that are waterlogging tolerant were able to persist under all Assisted Migration-Climate Change scenarios, and to a limited extent were able to colonize (and ephemerally dominate) upland sites. Well-parameterized landscape models provide a powerful tool to conduct simulation experiments involving novel situations

29 such as climate change, invasive (or intentionally migrated) tree species, invasive insects or diseases, and proposed  
30 management strategies.

## 31 **1. Introduction**

32 Within the Upper Great Lakes region (Michigan, Minnesota, Wisconsin, USA and Ontario, Canada), lowland and  
33 riparian hardwood forests can be dominated by black ash (*Fraxinus nigra* Marsh) and green ash (*Fraxinus pennsylvanica*  
34 Marshall); >95% of all stems can be ash species, creating a monoculture (Kolka et al. 2018). Lowland ash forests provide  
35 multiple ecological benefits, including cavities for nesting birds, subnivean habitat for mammals, shade for trout streams,  
36 carbon sequestration, and hydraulic buffering (Grinde et al. 2022, Flower et al. 2018, Fraver et al. 2022) as well as important  
37 cultural benefits to Native American tribes (Constanza et al. 2017).

38 These important ash forests are under threat from Emerald Ash Borer (EAB; *Agrilus planipennis* Fairmaire). EAB is  
39 a wood-boring beetle native to Asia that has caused widespread mortality of ash trees (genus *Fraxinus*; Anulewicz et al. 2014)  
40 across North America since its discovery in Michigan in 2002 (Cappaert et al. 2005, Flower et al. 2013). Once EAB reaches  
41 ash dominated wet lowland forests, the entire ecosystem is expected to convert from lowland forest to wet meadow, thereby  
42 losing the structure and function that the trees provide (Youngquist et al. 2017 but see Windmuller-Campione et al. 2021 for  
43 response to management of a variety of black ash ecosystems).

44 When lowland ash forests are killed by EAB, they lose their ability to transpire water, resulting in higher water tables  
45 (Slesak et al. 2014, Diamond et al. 2018). These changes, combined with more understory light following loss of the ash  
46 canopy, benefit the establishment of sedges and shrubs and inhibit establishment of a new cohort of trees (Looney et al. 2017).  
47 While lowland ash forest types can be seasonally inundated, the timing, seasonality, and impact of EAB varies among ash  
48 forest types. Additionally, climate change may enhance the spread of EAB to more northerly ash populations (Liang and Fei  
49 2014), but it also provides an opportunity for assisted migration (AM) of new tree species (Prasad et al. 2024) to provide  
50 similar ecosystems function under future climate, but without susceptibility to EAB. There have been recent efforts to collect  
51 foundational data on vegetation, hydrology, and soils for these lowland ash forest ecosystems (Benedict and Frelich 2008,  
52 Telander et al. 2015, Looney et al. 2017), as well as testing of new tree species to artificially reforest lowland ash forests (Palik  
53 et al. 2021, Keller et al. 2023, Keller et al. 2024). Such new data can facilitate improvements in the simulation of forest  
54 response to management (including AM) in lowland hardwood ecosystems across the Great Lakes Region to provide insight  
55 into the effectiveness of such management interventions at landscape scales of space ( $>10^5$  ha) and time (centuries) (e.g.,  
56 Gustafson et al. 2023a).

57 Simulation of lowland forest dynamics in landscape models has been problematic because the drivers of lowland  
58 hydrology have complex spatial (horizontal and vertical) components, making most algorithms to simulate such hydrology  
59 intractable at landscape scale (Sulman et al. 2013). The PnET-Succession extension of the LANDIS-II forest landscape model

60 was recently updated (v.5.1, Gustafson et al. 2023b) to include simple algorithms to approximate lowland hydrology that are  
61 tractable, but mimic lowland forest dynamics in a way that is appropriate for studies of landscape scale (space and time)  
62 responses of lowland forests to climate, disturbance, and management drivers (e.g., Gustafson et al. 2020, Gustafson et al. in  
63 review).

64 The objectives of our study were to 1) conduct empirical experiments to allow estimation of waterlogging tolerance  
65 parameters (for PnET-Succession) for lowland tree species currently extant in the upper Midwest (USA) or that are potential  
66 candidates to replace extant species that are under threat from exotic pests or climate change, 2) parameterize and test the  
67 simulated growth and competition behavior of these species using LANDIS-II (modeling experiment), and 3) simulate a case  
68 study of the effect of AM and climate change (using two different assisted migration strategies under two different climate  
69 change scenarios) on lowland forest dynamics in a northern Wisconsin landscape. Our focus was to explore the effects of  
70 improved waterlogging tolerance parameters in a heterogenous landscape over an ecologically relevant time frame.

## 71 **2 Methods**

### 72 **2.1 Model overview**

73 We used the PnET-Succession extension (Gustafson et al. 2023b) of LANDIS-II (Scheller et al. 2007) because it has  
74 reciprocal links between hydrology and tree physiological growth and competition. . LANDIS-II is a spatial forest landscape  
75 model simulating forest succession using the processes of forest development (seed dispersal, tree establishment, growth) and  
76 degeneration (disturbance, competition, senescence) through time. LANDIS-II consists of a library of modules (extensions)  
77 where specific extensions can be activated to simulate specific ecological processes. Landscapes are represented as a grid of  
78 spatially interacting cells on which species composition and vertical structure are assumed to be homogeneous, and these cells  
79 are spatially aggregated into biophysical spatial units (ecoregions) having homogeneous soils and climate. On each cell, forest  
80 composition is represented as age cohorts of one or more tree species whose competition is a function of their vital attributes  
81 (e.g., growth capacity, shade tolerance, drought tolerance, longevity, seed dispersal) and conditions on the cell. This produces  
82 stochastic successional pathways driven by abiotic conditions, competition, and disturbance (Mladenoff 2004). Independent  
83 disturbance extensions (including silvicultural activity) simulate processes that remove some or all of cohort biomass as a  
84 function of disturbance type and severity. PnET-Succession has direct links between climate drivers (CO<sub>2</sub> concentration,  
85 temperature, and precipitation) and tree species cohort net primary productivity, based on physiological first principles (Aber  
86 et al. 1995). PnET-Succession also accounts for life history traits that were necessary for our study (e.g., waterlogging  
87 tolerance, temperature tolerance). A mechanistic approach is critical when evaluating climate change effects because  
88 phenomenological modeling approaches use the past to predict the future and future climate is expected to fall well outside the  
89 domain of the empirically studied past (Gustafson 2013). Tree cohorts (and their propagules) compete for light and water on

each grid cell, with their competitiveness determined as their life-history traits interact with the abiotic environment, and this competition interacts with disturbances to determine successional trajectories.

In PnET-Succession, fWater is a water-stress cohort variable that functions as a reduction multiplier (one of several) that reduces the optimum net photosynthesis rate for the month. fWater is updated monthly from the soil water potential variable of the cell to reflect cohort water stress from either too little or too much water. A water-stress reduction (fWater) value of 1.0 produces no reduction in photosynthesis, and fWater=0.0 reduces photosynthesis to zero. We used the means of these calculated fWater equivalent values to estimate the PnET-Succession waterlogging tolerance parameters (H1 and H2) for each species. H1 is the soil water saturation potential (water potential, units=ABS(pressurehead)) beyond which photosynthesis completely shuts down (due to waterlogging) and H2 is the soil water saturation at which photosynthesis begins to decline. In the model, photosynthesis (i.e., fWater) declines linearly due to waterlogging as soil water saturation increases from H2 to H1. We arbitrarily set H1 of most species to -1.0 (to ensure that supersaturated soil water conditions never shut down photosynthesis for these waterlogging tolerant species) and then set H2 of each species to cause an fWater value equal to its estimated fWater equivalent (between 0.0 and 1.0) when the soil water potential equals 0 (flooded) (Table 1).

## 2.2 Empirical experiment

Controlled empirical experiments were designed to estimate the waterlogging tolerance of lowland tree species that might be candidates to replace ash trees following its extirpation by EAB (native and assisted migrants) (Poland and McCullough 2006). Native tree species for the experiment were selected based on initial consultation with Wisconsin Department of Natural Resources (WDNR) silviculturists and from previous work of Keller et al. (2023) (see Table 1). Potential assisted migrant species were included in the testing; these species, such as river birch, bald cypress, and American sycamore, are typically found in more southerly states like Indiana, Illinois, and Iowa. Seedlings were provided by the WDNR state nursery (Boscobel, Wisconsin). Bare-root seedlings were planted in potting soil (Berger BM7) in seedling containers (10 cm x 10 cm x 30 cm), and then placed in 30 cm tall water tanks in which the water level (i.e. flood level) was controlled. The tanks were arranged outside in a split-split plot randomized complete block design with three levels of soil saturation (water level maintained at 0, 14, 27 cm below the soil surface), referred to as flooded, intermediate, and drained, respectively. Additionally, three levels of light availability were manipulated using shade cloth (full light, 40% reduction, 70% reduction). For the purposes of this study, we used only the full light treatment. Flooding levels were checked weekly to ensure that the treatments were accurately applied throughout the experiment (i.e., 14 weeks). Each treatment block was replicated eight times. The details of the empirical experiment along with measurements of photosynthesis, soil water potential, and growth can be found in Keller et al. (2024).

**Table 1. PnET-Succession waterlogging parameters revised from a prior study (Gustafson et al. 2023a) based on the results of the empirical experiment. Species not included in the empirical experiment are not shown.**

Species scientific name	Species common name	New H1 <sup>1</sup>	New H2 <sup>1</sup>	Prior H1 <sup>1</sup>	Prior H2 <sup>1</sup>	AM strategy <sup>2</sup>
<i>Acer rubra</i>	Red maple	-1	0.869	-1.9	2.75	Native
<i>Fraxinus nigra</i>	Black ash	-500	1	-5	0.5	Native
<i>Populus tremuloides</i>	Quaking aspen	-1	0.767	0	4	Native
<i>Quercus macrocarpa</i>	Bur oak	-1	3.37	0	4	Native
<i>Thuja occidentalis</i>	Eastern white cedar	-1	1.704	-4	2	Native
<i>Larix laricina</i>	Eastern larch (tamarack)	-1	0.725	-3.5	1.5	Native
<i>Picea mariana</i>	Black spruce	-1	0.725	-4	2	Native
<i>Acer saccharinum</i>	Silver maple	-1	0.808	-3.1	2	MedRng
<i>Platanus occidentalis</i>	American sycamore <sup>3</sup>	-1	0.5226	-3.1 <sup>3</sup>	2 <sup>3</sup>	MedRng
<i>Ulmus americana</i>	American elm	-1	0.7164	-3.5	1.5	MedRng
<i>Quercus bicolor</i>	Swamp white oak	-1	0.5542	-3.3	2	LongRng
<i>Betula nigra</i>	River birch	-1000	1	-2	3	Both <sup>4</sup>
<i>Quercus bicolor</i>	Swamp white oak	-1	0.5542	-3.3	2	LongRng
<i>Taxodium disticum</i>	Bald cypress	-1	0.3057	-5	1	Both <sup>4</sup>

121 <sup>1</sup> PnET-Succession waterlogging parameters. H1-soil water saturation (units: ABS(pressurehead)) above which  
122 photosynthesis stops; H2- soil water saturation above which photosynthesis begins to slow (waterlogging stress begins).

123 “New” indicates experimentally derived parameter values for this study; “Prior” indicates parameters used by Gustafson et al.  
124 (2023b).

125 <sup>2</sup> Assisted migration strategy (distance of source plantings from study area) in which species were planted in the case study.  
126 See text and Gustafson et al. (2023b) for details.

127 <sup>3</sup> This species was not used in the prior study (Gustafson et al. 2023a). For this comparison, its prior H1/H2 values were  
128 assigned the same as Silver Maple.

129 <sup>4</sup> Atlantic white cedar (*Chamaecyparis thyoides*, not studied in empirical tests and given the waterlogging parameters of *Thuja*  
130 *occidentalis*), was always planted with bald cypress (both AM strategies) in the case study.

131

132 Throughout the empirical experiment , soil moisture was recorded at 10cm depth for a subset of the potted tree  
133 species. However, for the purposes of mimicking the experiment with LANDIS, more specific measurements were needed to  
134 calculate mean soil water potential throughout the soil profile for a given flooding treatment. After the experiment, water  
135 tanks used from the original empirical experiment were each filled to one of the three water levels used during the original  
136 experiment (0, 14, 27, cm below the soil surface). Twelve 10 x 10 x 30cm pots were filled with the same Berger BM7 potting  
137 soil used in the experiment and four pots were placed into each of the three water tanks to replicate the water level and  
138 corresponding soil moisture values. Soil moisture and soil water potential were collected in the saturated portions of the soil  
139 profile for all three treatments and also in the two unsaturated portions of the soil profile for the 14 cm and 27 cm treatments  
140 using Teros 12 and Teros 21 sensors connected to a Z6 datalogger (Meter Equipment Group), respectively. Weighted averages  
141 were calculated for soil moisture and soil water potential for each of the three treatments using the soil moisture data as a  
142 proportion of the soil profile that was either saturated or unsaturated.

143 Although photosynthesis was measured every two weeks throughout the 14-week empirical experiment , we used  
144 only the two measurement cycles from August to parameterize PnET-Succession. We chose August because it represents peak  
145 growing season photosynthesis and allowed eight weeks for the flooding treatments to produce an effect on the seedlings. To  
146 estimate PnET-Succession waterlogging tolerance parameters, we used the empirical photosynthesis measurements to estimate  
147 the reduction in photosynthesis caused by the flooded treatment (0cm of aerated soil) compared to photosynthesis on the  
148 drained treatment (27cm of aerated soil) for each species. This proportional reduction observed is equivalent to the fWater  
149 variable in PnET-Succession.

150 The empirical experiment treatments were thus used to estimate the waterlogging tolerance parameters of PnET-  
151 Succession. In turn, PnET-Succession was used to virtually scale the empirical experiment to longer timeframes and more  
152 diverse species assemblages. PnET-Succession has direct links between climate drivers (CO<sub>2</sub> concentration, temperature and  
153 precipitation) and tree species cohort net primary productivity, using physiological first principles (Aber et al. 1995). PnET-

Succession also accounts for life history traits that are necessary for this study (e.g., waterlogging, shade, and temperature tolerances). Such a mechanistic modeling approach is superior to phenomenological approaches that use the past to predict the future when the future is expected to be exceptionally novel (Gustafson 2013). In PnET-Succession, tree cohorts (and their propagules) compete for light and water on each grid cell as their life-history traits interact with the abiotic environment, and the outcome of competition interacts with disturbances to determine successional rates and trajectories. The model allows some runoff from landscape grid cells to be retained on site (as standing water), the maximum height of which (RunoffCapture) is typically specified only for lowland-specific ecoregions. Additionally, the leakage fraction parameter (default=1.0) can be reduced for specific ecoregions to represent increasingly impermeable soil within or below the rooting zone, also encouraging waterlogged conditions to develop dynamically (monthly) as a function of inputs (precipitation) and outputs (runoff, leakage, evapotranspiration comprised of evaporation, interception, and transpiration). Thus, cohorts on a lowland cell may experience fluctuating water stress (either too dry or too wet) as precipitation and evapotranspiration (as influenced by all living cohorts) raise and lower the soil water potential on a monthly or annual basis. Complete details about PnET-Succession algorithms can be found in Gustafson et al. (2023a).

### 2.3 Modeling the empirical experiment

We first conducted a virtual (simulated) version of the empirical experiment, seeking to mimic as closely as possible the experimental empirical results. The purpose was to assess the difference in simulated outcomes between empirically estimated waterlogging tolerance parameters (H1, H2) and those estimated from other sources for prior studies. Prior to this study, H1 and H2 parameter values were given values with little empirical basis and their relative settings were typically derived from synthetic meta-databases (e.g., Kattge et al. 2020, Niinemets and Valladares 2006). Our modeling exercises tested and evaluated the parameter values estimated from the empirical experiment.

For these initial tests of the model, we constructed an artificial “landscape” where each individual grid cell represented a specific combination of water-level treatment and tree species of the empirical experiment. To maintain the isolation of these cells (experimental units in this context), seed dispersal and establishment were turned off. Tree seedlings from the empirical study were simulated by establishing a new cohort of the designated species on each given cell (MapCode) as specified in the initial conditions input file. The soil and water-level conditions were controlled by setting the soil parameters assigned to the cell to mimic waterlogging treatment conditions from the empirical study. The ecoregion soil was a novel soil texture (PTNG) parameterized by us to represent the potting soil used in the empirical experiment (Table A1). The model simulates soil water dynamics using a “leaky bucket” conceptual approach, where soil moisture in each month is modified from previous month moisture as a function of current month water inputs (precipitation and snow melt) and outputs (runoff, leakage (drainage out of the rooting zone), and evapotranspiration). Because the model does not track (or control) water table, the water-level treatments (drained, intermediate, and flooded) were tricky to implement. The precipitation inputs required by the model were made artificially high (and distributed equally across all days of the month) to ensure adequate water to maintain the “flooded” treatment. Each water level treatment was implemented by varying LeakageFrac and RunoffCapture

(Table 2) to produce appropriate soil water conditions. The rooting zone was constant across all treatments and was equal to the depth of the pots used in the empirical experiment (280 mm). Monthly temperature and other abiotic conditions (PAR, [CO<sub>2</sub>]) inputs were held constant year to year in the model and reflect the conditions observed at the site of the experiment (Table A2). Thus, our model did not explicitly mimic all the components of the empirical experiment (e.g., weather), but attempted to mimic the experimental factors explicitly held constant or manipulated.

**Table 2. Ecoregion parameter values used to mimic the three empirical soil water treatments (see text).**

Parameter name	Drained	INT	Flooded
RootingDepth (mm)	280	280	280
LeakageFrac <sup>1</sup>	0.9167	0.125	0.0
RunoffCapture (mm) <sup>2</sup>	0	0	10

<sup>1</sup> Leakage fraction; fraction of soil capacity water that percolates out of the rooting zone (i.e., slow leakage)

<sup>2</sup> Height above ground surface of basin outlet; allows standing water on the cell up to this depth

**2.4 Temporally extending the empirical experiments with the model**

To mimic the empirical experiment, we used the PnET-Succession species parameters estimated from the empirical results (Table 1) and parameters representing the potting soil used in the empirical experiment (Table A1). Other LANDIS-II and PnET-Succession parameters took values used in other studies in northern Wisconsin (e.g., Gustafson et al. 2023a). We virtually extended the length of the empirical experiment using the calibrated and validated model parameters to provide insights into the likely long-term outcomes of such treatments. We ran these simulations for 90 years, matching the longevity of the shortest-lived species. We also conducted a virtual experiment to project the outcome of a hypothetical empirical experiment in which two competing species were grown together (for 90 years) under the experimental treatments.

We also evaluated a simple case study on a real landscape in northern Wisconsin (Fig. 1) to project the outcome of using AM species to replace keystone tree species that are expected to be lost from this landscape in the near future (both upland and lowland). We simulated the extirpation of black ash from lowland hardwood stands by EAB in the next decade (using the Biological Disturbance Agent extension, Sturtevant et al. 2004) and ash replacement through AM. To this end, we simulated two AM scenarios (involving many tree species) for 300 years under two climate futures (RCP6.0 (mean monthly precipitation = 79.14 mm/mo) and RCP8.5 (91.67)) to allow the treatments time to respond to the climate signal and to



210 overcome the ecological inertia of the initial conditions. The less aggressive AM strategy (MediumRange) planted species  
211 with ranges centered to the south of the range of endemic species on the study landscape, and the more aggressive strategy  
212 (LongRange) planted species having ranges centered even further south. We used the simulation methods of Gustafson  
213 et al. (2023b), using their initial landscape conditions and parameters for the disturbances (including their AM strategies) that  
214 shape successional dynamics in this landscape. We also used their ecoregion and species parameters, except substituting the  
215 H1 and H2 parameters estimated for the species studied in the empirical experiments described above. Because lowland bogs  
216 exist in this landscape, we revised black spruce H1 and H2 values to those of tamarack to make its waterlogging tolerance  
217 equal to its primary competitor on lowland bogs (Table 1). We compared the outcomes produced with revised waterlogging  
218 parameters (H1, H2) to those generated using the parameter values of Gustafson et al. (2023b) (i.e., Prior-H1 and Prior-H2 in  
219 Table 1). We simulated two replicates of each scenario combination for 300 years because variation in outcomes was low and  
220 we were not testing hypotheses.

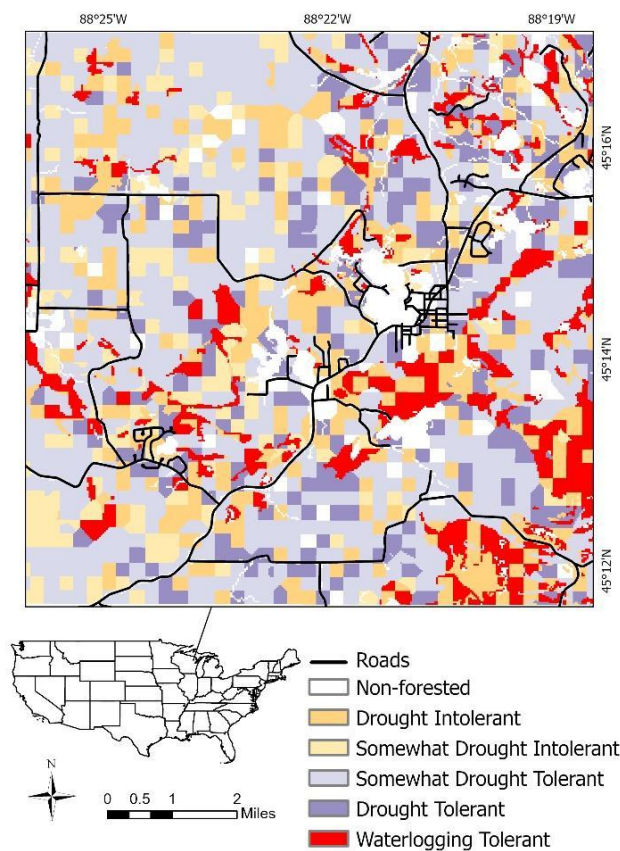


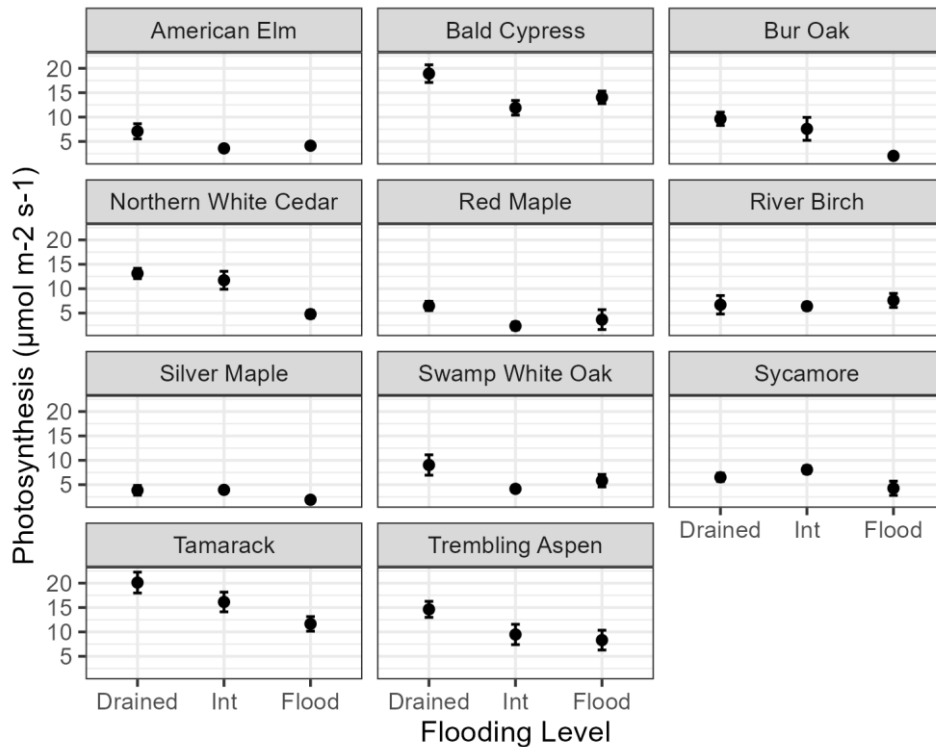
Figure 1. Case study landscape (9,608 ha) located in Oconto County (Wisconsin, USA), showing soil type (left) and Time Zero dominant tree species types (right) defined by tolerance of soil water stress. Waterlogging tolerant tree species are currently dominant mostly on lowland (MUCK) sites.

**3. Results**

**3.1 Empirical experiment**

The empirical measurements resulted in revised H1 and H2 parameter settings that in some cases varied considerably from prior settings (Table 1). The common setting of  $H1 = -1$ , caused H2 to be inversely related to waterlogging tolerance. H1 values  $< -1$  produce very high waterlogging tolerance by ensuring fWater values near 1.0 when pressurehead=0 (i.e., photosynthesis is little reduced by flooding).

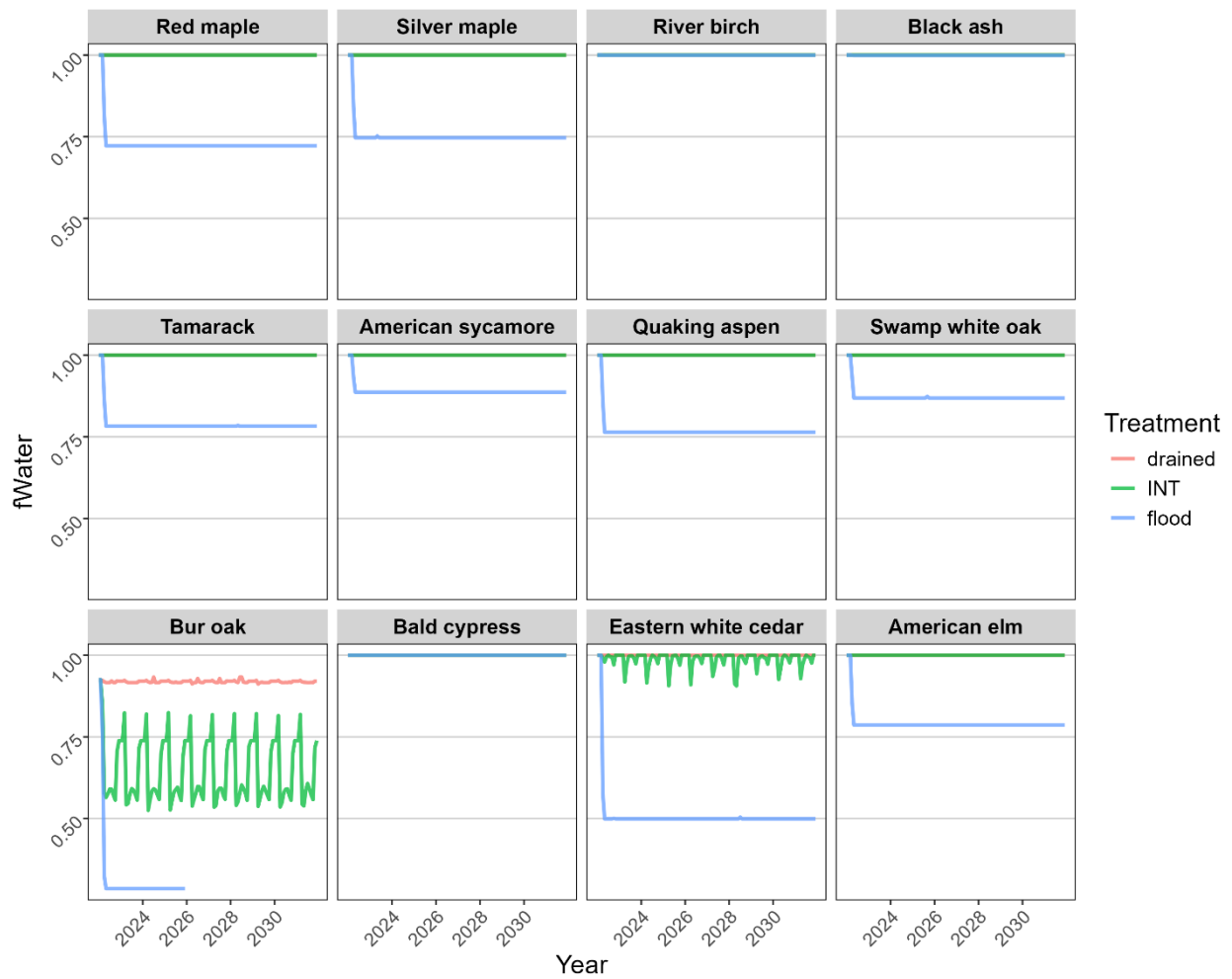
In the empirical experiment, the waterlogging stress of seedlings of most species under the “drained” and “intermediate” treatments was quite similar because at least half of both soil profiles was well-drained. Species exhibited three general responses to the treatments (Fig. 2). Most species showed a response similar to bur oak, where the drained and Intermediate treatments were relatively unstressed compared to the flooded treatment. Trembling aspen represents species that were relatively easily waterlogging stressed, but flooding was not dramatically worse. Bald cypress represents a highly water tolerant species that responded equally well to both the Intermediate and Flooded treatments and performed very well under the drained treatment.



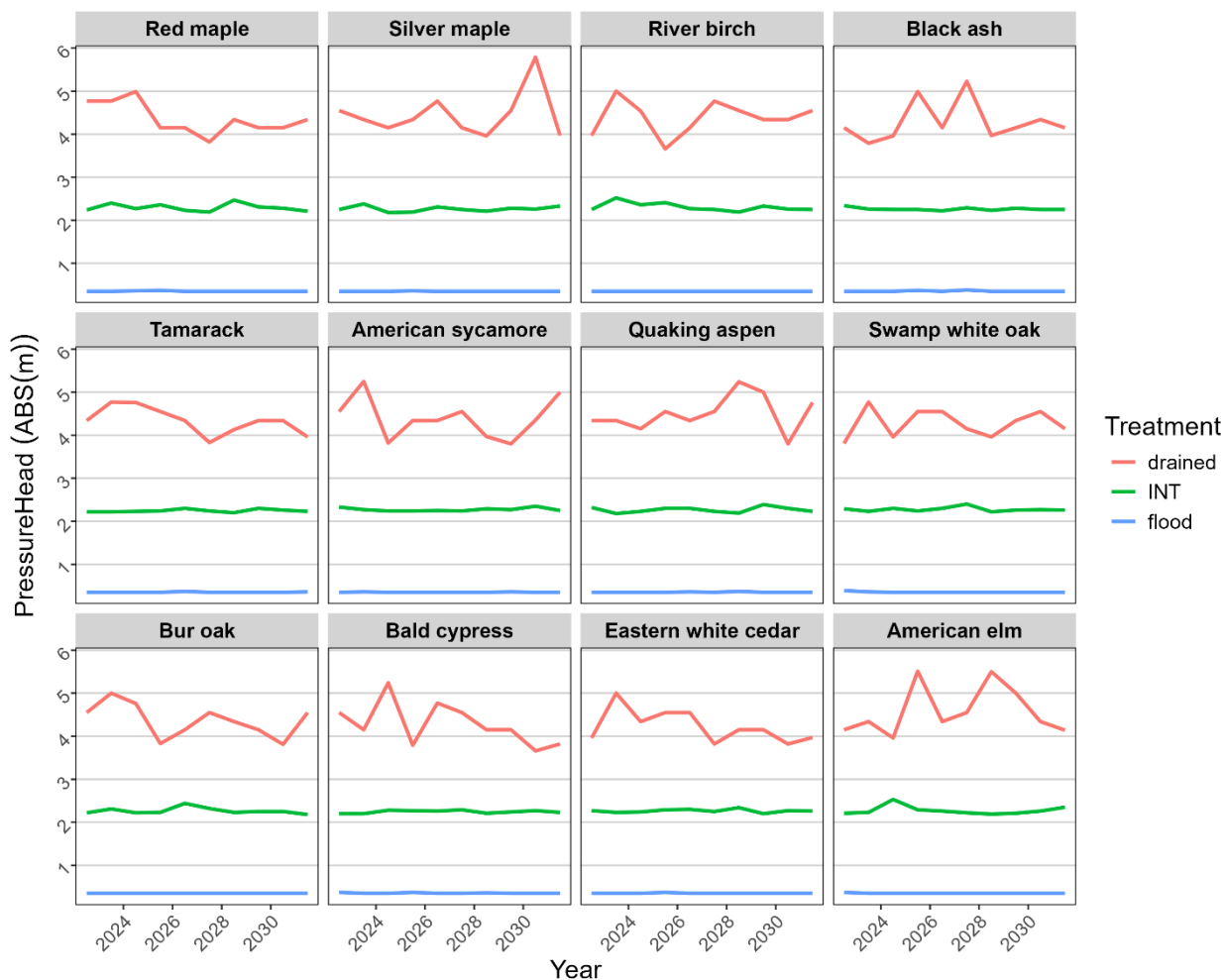
**Figure 2. Effects of flooding level on photosynthesis for potted seedlings grown in full sunlight in the empirical experiment. Error bars show standard error.**

### 3.2 Modeling the empirical experiment

We were successful in parameterizing the model to mimic the soil water conditions of the empirical experiment. Simulated water stress (fWater) mimicked two critical characteristics of the experiment: 1) there was very little difference between the well-drained and intermediate flooding treatments and 2) water stress of species (Fig. 3) aligned with the empirical waterlogging tolerance (Table 1). This was achieved with the average simulated soil water potential of the INT treatment (across entire soil profile) falling midway between field capacity and flooded (Fig. 4), mimicking expectations for the intermediate flooding level.

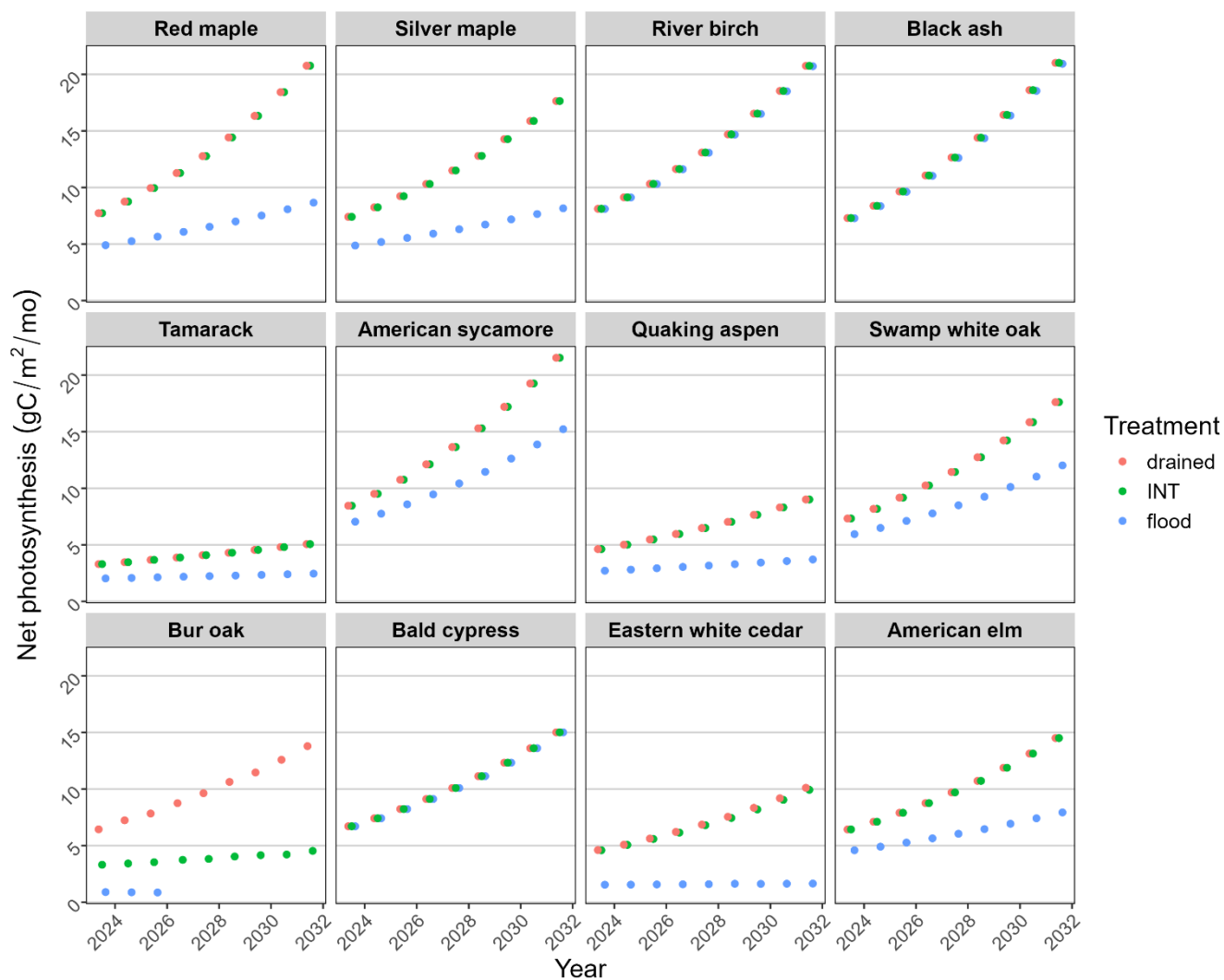


**Figure 3. Simulated monthly water stress (fWater) for each species under three water-level treatments. fWater=1.0 indicates no stress; fWater=0.0 causes photosynthesis to completely shut down. In most cases, the “INT” line obscures the “drained” line.**

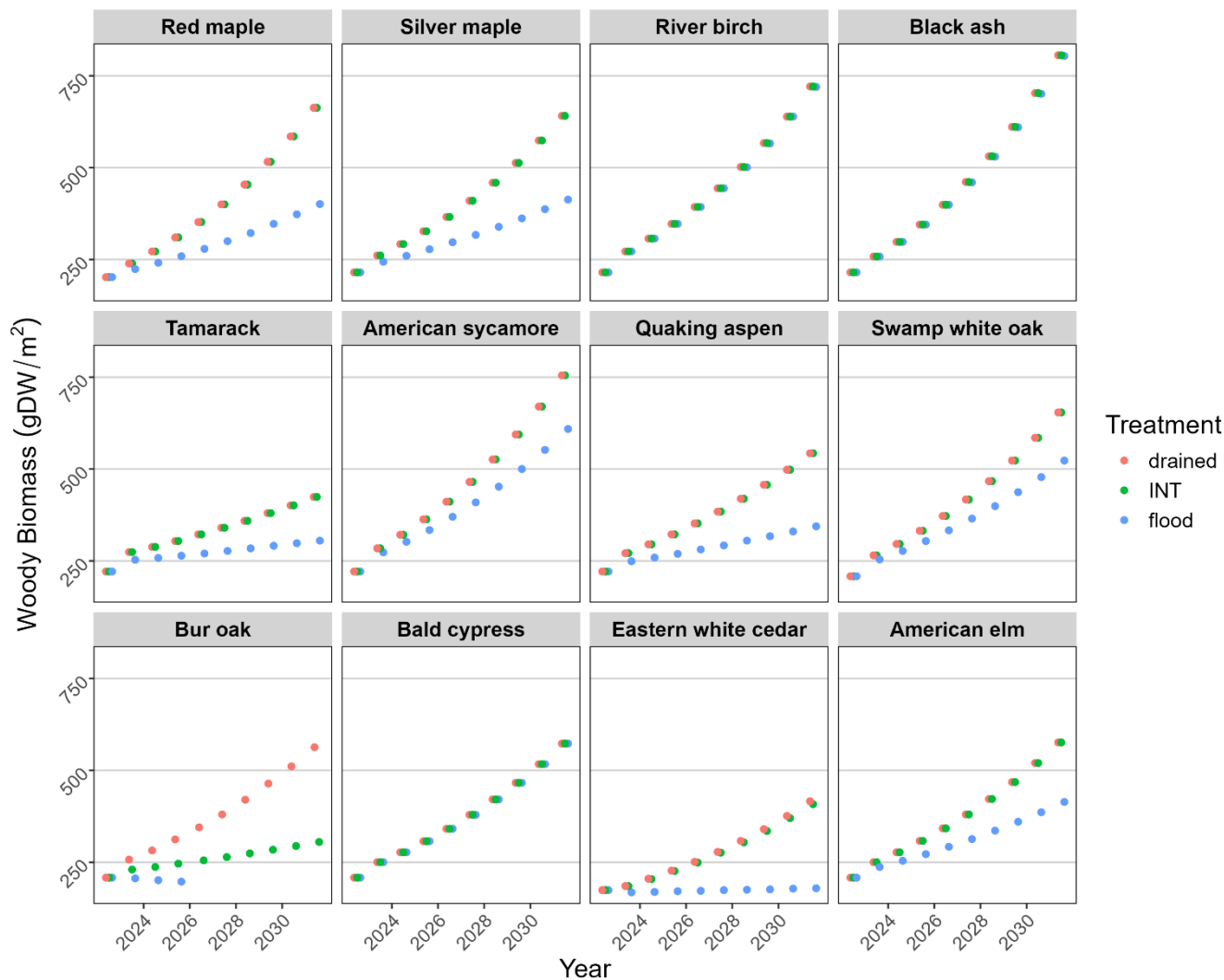


**Figure 4. Simulated mean July soil water potential (absolute value of m pressure head; higher values represent less moisture), accounting for transpiration by seedlings.**

In the empirical experiment, the growth of seedlings of most species under the “drained” and “Intermediate” treatments were quite similar because at least half of the soil profile was well-drained (Fig. 2). Growth was reduced under flooded conditions, with the amount of reduction reflecting the species’ waterlogging tolerance. Growth (NetPsn, biomass accumulation) was also simulated by the model as generally lower under flooded conditions, and identical under drained and INT conditions (Fig. 5, Fig. 6).



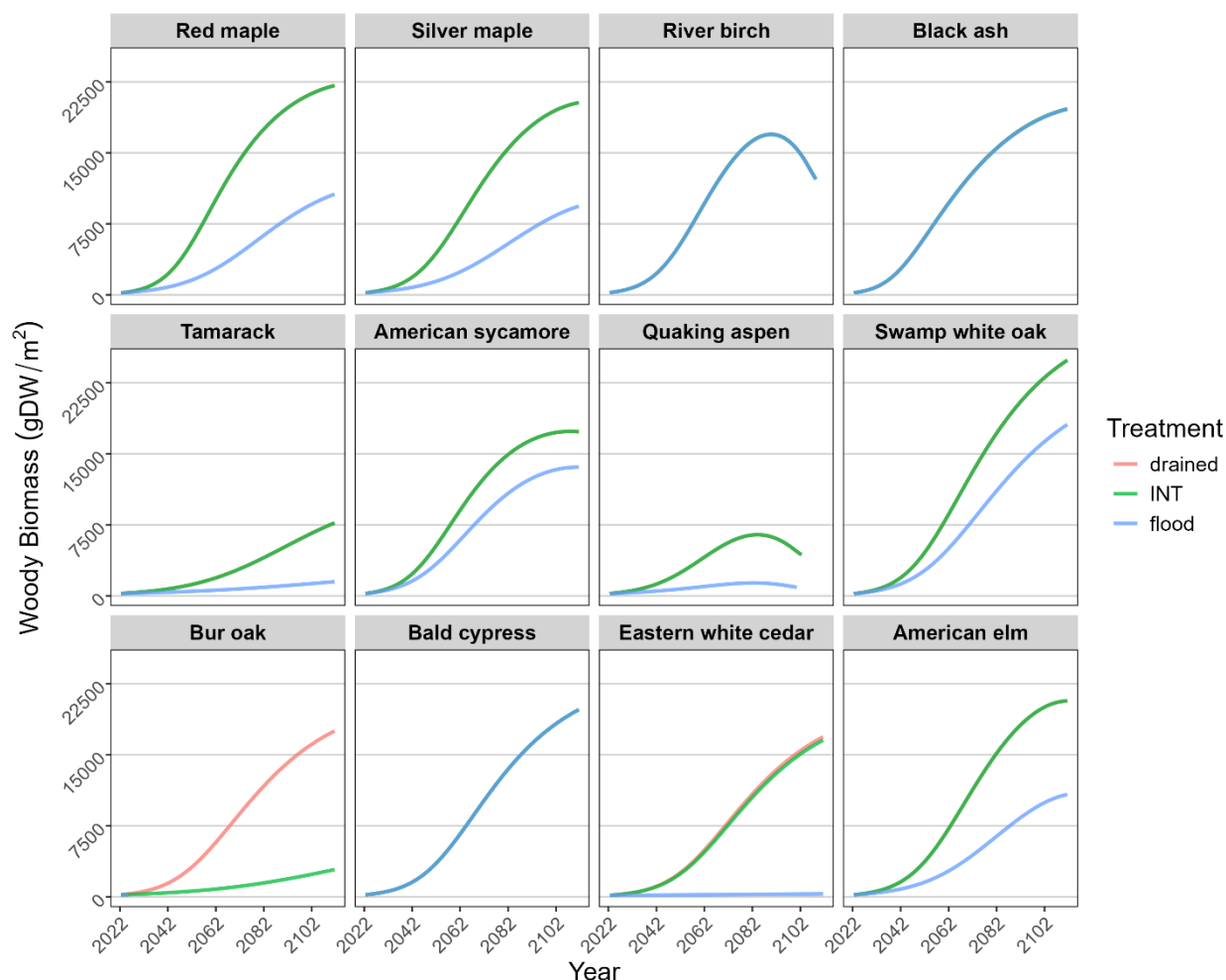
**Figure 5. Simulated mean July net photosynthesis for each species (beginning at age 1) under three water-level treatments. In most cases, the “INT” points obscure the “drained” points.**



**Figure 6. Simulated biomass accumulation for each species (beginning at age 1) under three water-level treatments. Lines ending prematurely indicate the death of the cohort. In most cases, the “INT” points obscure the “drained” points, and in some cases, it obscures both other points.**

### 3.3 Temporally extending the empirical experiment

When the model extended the empirical experiment for 90 years, species productivity and biomass accumulation plateaued or peaked at a level that could be supported by rooting zone depth and (the artificially high) precipitation (Fig. 7). Note that the units reported for biomass density are scaled up from planting pots to a full m<sup>2</sup> of area. Decline occurs when a cohort nears longevity age, and it may die before reaching longevity if it is stressed (e.g., quaking aspen).

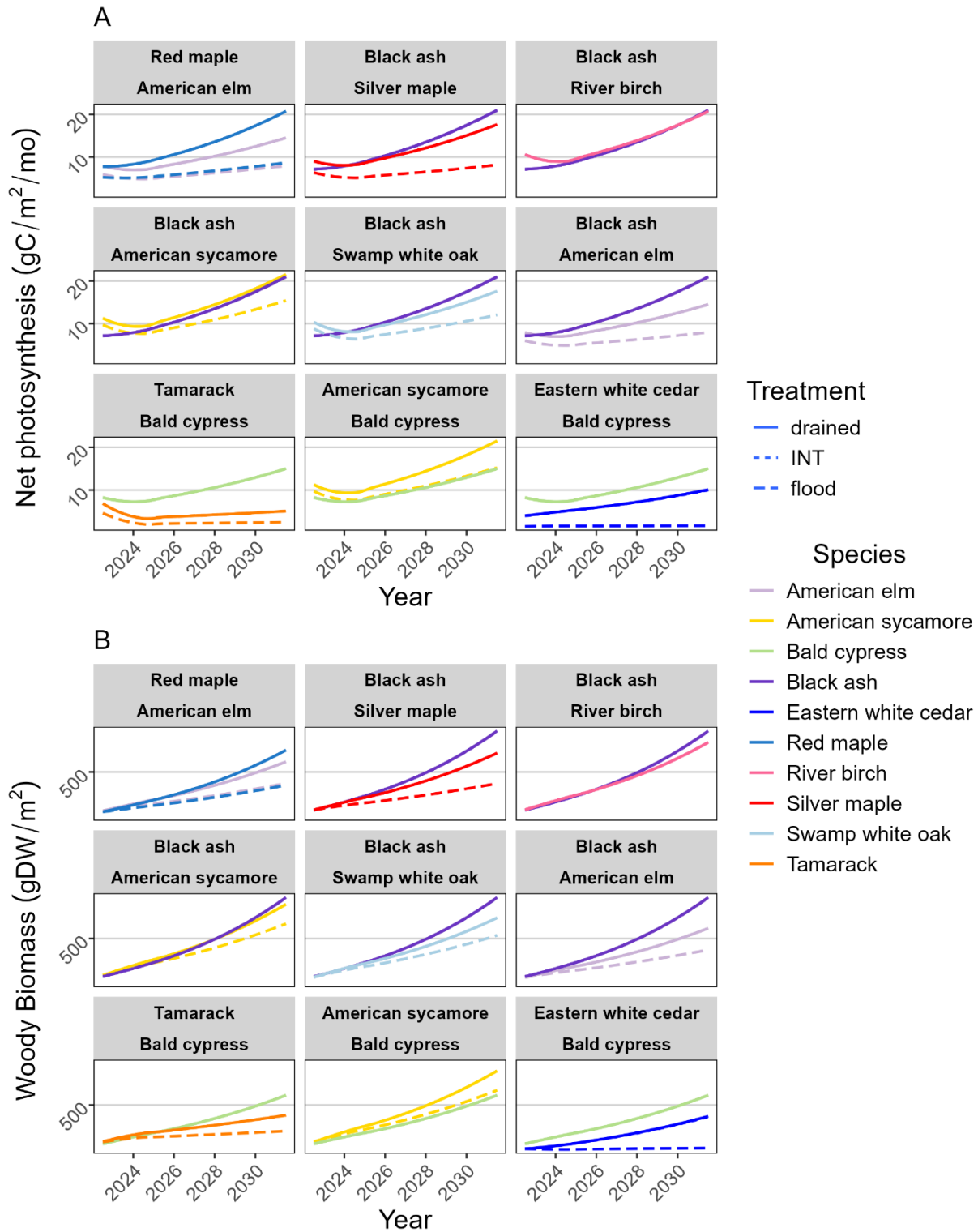


**Figure 7. 90-year extension of the simulation showing simulated biomass accumulation through time for each species (beginning at age 1) under three water-level treatments. In most cases, the “INT” line obscures the “drained” line, and in some cases, it obscures both other lines.**

In the experiment extension that tested competition outcomes between two species (Fig. 8) we expected more waterlogging tolerant species to be less productive than less waterlogging tolerant species because such species generally have less drought tolerance, suggesting more susceptibility to dry soil. While more waterlogging tolerant species did much better under the flooding treatment (as expected), they also tended to do well on the drained treatment. This is because the drained treatment did not produce dry soil, but a very moist soil kept constantly near field capacity by the precipitation inputs (Fig. 4). No species combination tested resulted in the death of a species in the first 10 years, although bald cypress severely suppressed northern white cedar under flooded conditions. We note that even though American elm is slightly more waterlogging tolerant



285     than red maple, red maple was more productive than American elm except perhaps under flooded conditions, likely because  
286     red maple is slightly more shade tolerant than American elm.



288 **Figure 8. Simulated 10-year trajectories of (A) net photosynthesis and (B) woody biomass when two cohorts are established in the**  
289 **same cell (planter pot). Cohorts are represented by color, and often the treatment lines overlap, consistent with empirical**  
290 **observations.**

291 **3.4 Landscape case study**

292 When the updated waterlogging parameters were applied at landscape scale under climate change and AM, the  
293 landscape mean biomass density ( $\text{g/m}^2$ ) of native species generally declined, and the AM species increased as climate gradually  
294 changed and planted cohorts of the most waterlogging tolerant species thrived (Fig. 9). American elm and bald cypress thrived  
295 under the RCP6.0 climate, while river birch and the other MediumRange AM species merely established a presence. However,  
296 under the RCP8.5 climate, American elm, bald cypress and river birch thrived, while the other LongRange AM species barely  
297 survived (Fig. 9).

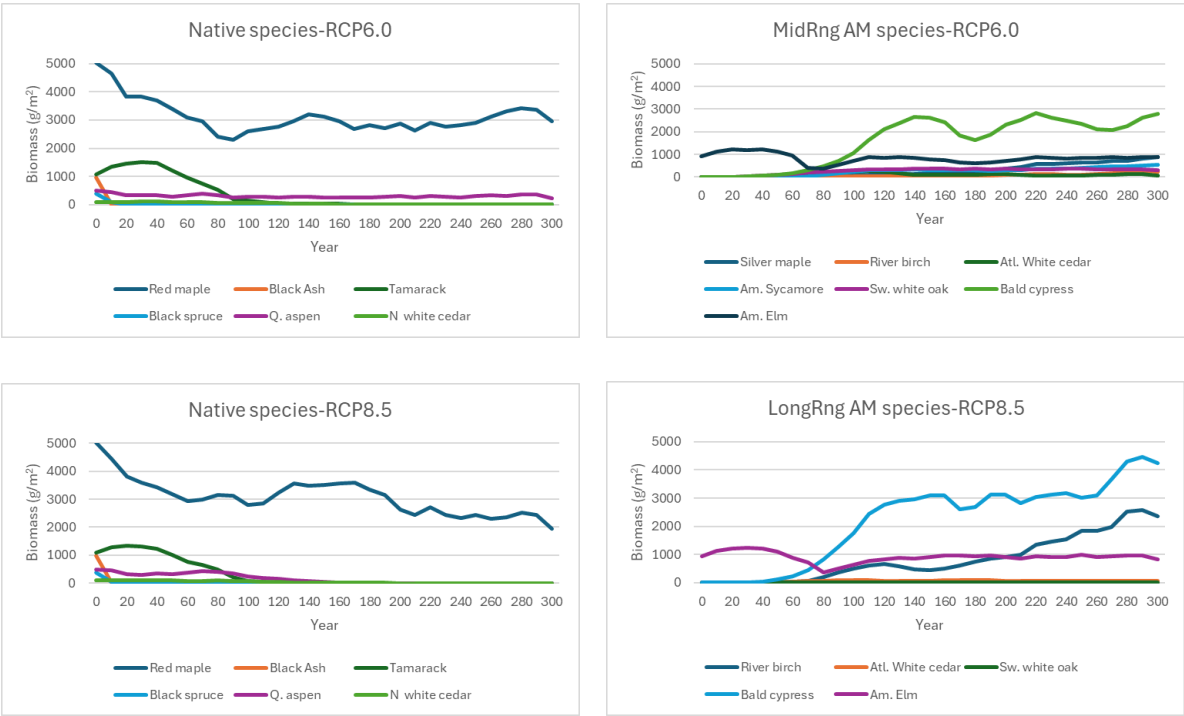
298

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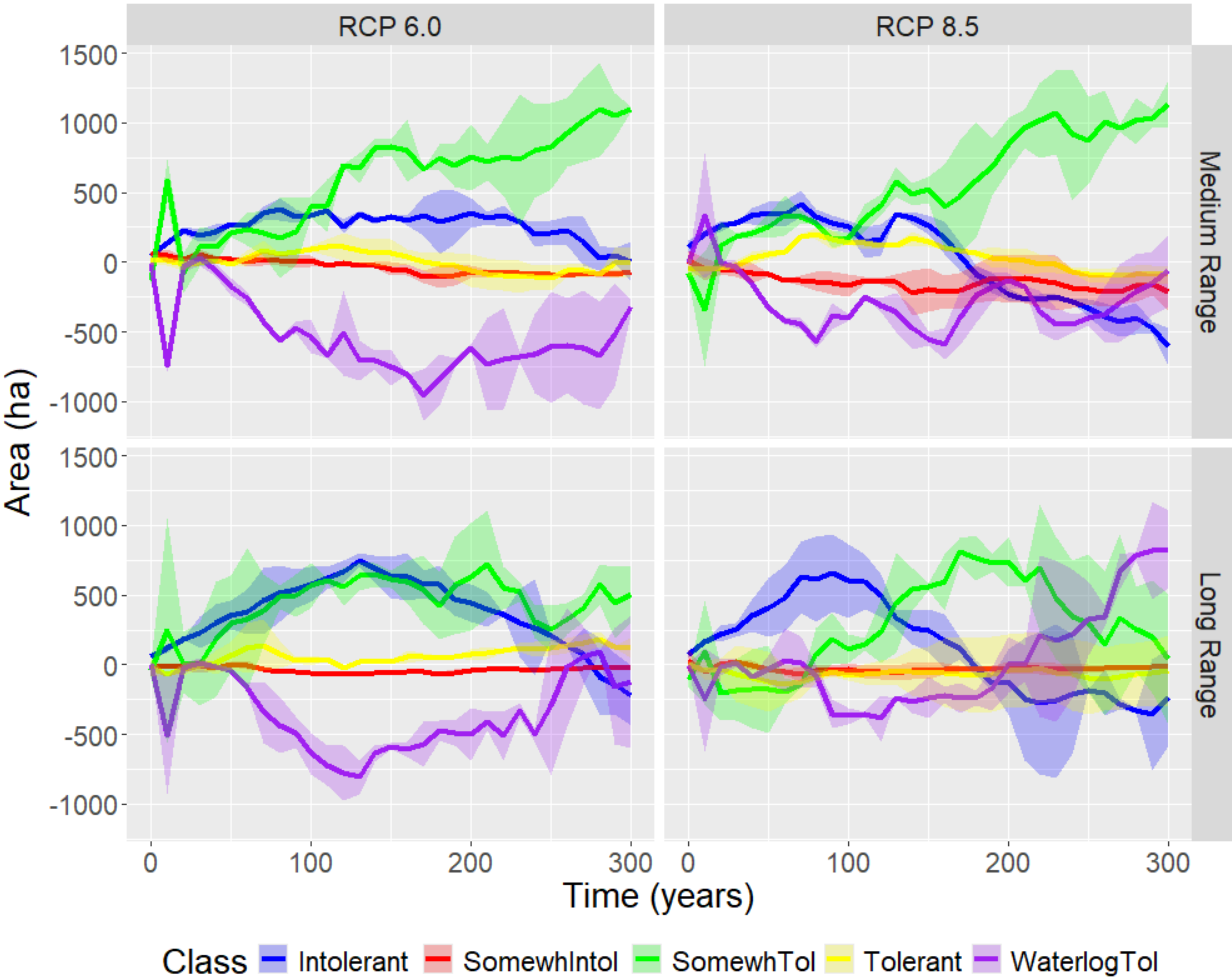
300

301 **Figure 9. Mean woody biomass (across active landscape cells and replicates) of lowland tree species under the AM-Climate Change**  
302 **scenarios on the case study landscape.**

303 The revised waterlogging tolerance parameters (Table 1) resulted in some differences in the area of the landscape  
304 dominated by waterlogging tolerant species, especially under severe climate change (Fig. 10). The most obvious effect of the  
305 (usually only slightly) revised waterlogging tolerance parameters was that a decline in abundance of waterlogging tolerant



species tended to mirror an increase in abundance of drought-intolerant species, which tend to have only slightly less  
 waterlogging tolerance than hydrophilic species. Thus, the ability to thrive with waterlogging appears to be a step-function  
 with a very abrupt threshold (i.e., a very small difference in waterlogging tolerance may be the difference between living and  
 dying). Lowland types (waterlogging tolerant species) were able to persist under all AM-Climate Change scenarios and to a  
 limited extent were able to colonize (and ephemerally dominate) upland sites (not shown). The Long-range waterlogging  
 tolerant AM species thrived under both climates (especially severe climate change) and became more abundant than the  
 Medium-range waterlogging tolerant species under both climates. The productivity of many species was overestimated with  
 the prior waterlogging parameters, and some were grossly underestimated (Table 3).



315

316 **Figure 10. Difference (Revised minus (-) Prior) through time in landscape area dominated by species in each (of four) drought**  
 317 **tolerance class and (one) waterlogging tolerance class caused by differences in waterlogging tolerance parameters (Table 1) by**  
 318 **climate scenario and AM strategy. The waterlogging tolerance class contains all species having at least some waterlogging tolerance**  
 319 **(H2<3.37). Ribbons show 1 STD of two replicates.**

320  
 321 **Table 3. Mean (active landscape cells, two replicates) biomass density at year 300 of the landscape case study of cohorts of selected**  
 322 **species under selected AM and climate scenarios , comparing landscape outcomes when prior or revised waterlogging**  
 323 **parameters were used .**

	Prior parameters	New parameters	Prior parameters	New parameters
AM/Climate Scenario	Medium Range-SSP6.0		Long Range-SSP8.5	
Species	Biomass (g/m <sup>2</sup> )		Biomass (g/m <sup>2</sup> )	
Red maple	823.10	2945.07	1109.30	1655.57
Silver maple	764.40	854.49	0.0 <sup>1</sup>	0.0 <sup>1</sup>
River birch	1.01	245.97	208.98	1921.97
Black ash <sup>2</sup>	0.0	0.0	0.0	0.0
American sycamore	258.45	531.84	0.0 <sup>1</sup>	0.0 <sup>1</sup>
Quaking aspen	0.02	239.78	0.0 <sup>3</sup>	0.0 <sup>3</sup>
Swamp white oak	343.54	340.55	0.95	0.27
Bur oak	11.19	72.56	0.32	0.52
Bald cypress	3152.49	2704.17	4366.78	4946.72
Eastern white cedar	5.68	2.11	0.36	0.002

American elm	1097.15	834.18	1095.85	643.22
Eastern larch (tamarack)	6.83	1.50	0.0 <sup>3</sup>	0.0 <sup>3</sup>
Black spruce	0.0 <sup>3</sup>	0.0 <sup>3</sup>	0.0 <sup>3</sup>	0.0 <sup>3</sup>

324 <sup>1</sup> not planted in this AM scenario

325 <sup>2</sup> species extirpated by EAB in all scenarios

326 <sup>3</sup> species unable to survive in this climate scenario

327 **4. Discussion**

328 Using empirical experiments, we were able to estimate waterlogging tolerance parameters needed by the PnET-  
329 Succession forest landscape model for specific tree species that are being considered as replacements for species that are  
330 threatened by exotic pests and/or climate change in the upper Great Lakes region (Keller et al. 2024). These revised parameter  
331 values represented both absolute and relative waterlogging tolerance rather than just the relative tolerance given by other  
332 sources (e.g., Niinemets and Valladares 2006, Kattge et al. 2020) and as used in other studies (e.g., Gustafson et al. 2023a).  
333 They allow for greater photosynthetic productivity to occur in model simulations for trees growing in lowland sites according  
334 to their waterlogging tolerance. Three species (river birch, black ash, bald cypress) tested were as productive in flooded  
335 conditions as in well-drained (field capacity) conditions (Fig. 5). The landscape model projected that many potential lowland  
336 forest replacement species can be expected to physiologically thrive under both future climates tested (Fig. 9), although we  
337 did not simulate every possible future disturbance or pest scenario. Pests can have considerable impacts for forest successional  
338 dynamics and carbon cycling (Flower and Gonzalez-Meler 2015). Such model capabilities enable more robust evaluation of  
339 proposed lowland forest management strategies to mitigate forest stressors such as climate change and invasive pests, and  
340 model results can help forest managers to select which species should be tested first given the unique site characteristics on  
341 their landscape.

342 A major impetus for our work is the impending loss of extant black ash stands to the EAB in upper Great Lakes  
343 lowland forests. It is feared that the abrupt and catastrophic elimination of all pole- and larger-sized ash trees from such stands  
344 will cause chronically high water-levels that will greatly reduce regeneration potential and shift ecosystems towards open  
345 meadows (Flower and Gonzalez-Meler 2015, Diamond et al. 2018, Kolka et al. 2018). While our results produce some  
346 indication that other species might be able to replace the water-transpiration function that black ash currently provides, there  
347 remain many uncertainties that must be resolved with tree- and stand-level empirical studies. For example, the model makes  
348 many simplifying assumptions about the establishment of cohorts via planting, and many of these have not yet been verified  
349 under lowland conditions at the stand scale. Furthermore, planting of new species before vs. after the mortality of black ash

probably requires very different silviculture methods and likely has very different outcomes. The development of methods for establishing new lowland tree species is a very new and difficult line of research because water levels can rapidly and unpredictably fluctuate. However, the results presented in this study show that a variety of different species have the ability to exist in poorly drained lowland forests under future climate. Forest managers might consider including these species in future enrichment plantings to diversify lowland forests and create greater resilience to future pests and hydrologic changes. Our simulation results should not be viewed as definitive, but at best, hopeful. That is, if the challenges surrounding regeneration and establishment (of productive lowland cohorts) can be addressed, the long-term prospects are promising.

Our case study simulations were conducted using the same model on a subset of the landscape studied by Gustafson et al. (2023b), allowing landscape-scale results obtained using their (prior) waterlogging parameter values to be compared to ours. Although we simulated one species (American sycamore) not simulated by Gustafson et al. (2023a), we found that lowland forest types were generally able to persist under either AM strategy (Fig. 10). Because simulation of lowland forests has historically been a weakness of forest landscape models, there are few other studies that are directly comparable.

Our study illustrates the power of LANDIS-II, specifically PnET-Succession, as a scaling tool. PnET-Succession uses mechanistic algorithms based on physiological first principles to simulate cohort growth (photosynthesis), competition for limited resources, and response to stressors, making it robust to novel drivers (including climate and CO<sub>2</sub>), novel species assemblages, and novel management strategies. We were easily able to use the model to extrapolate the one-year empirical experiment to 90 years and scale up that experiment from planting pots to a real landscape under different climate futures. Our landscape case study provides evidence that AM of lowland forest species may indeed be capable of conserving lowland forest function and services.

While data, scientific studies, and management within lowland hardwood forests are increasingly prevalent, the vast majority of such work has been done within the last 10 years. Black ash forest communities have typically been long-lived and are varied in age structure, so 10 years represents a very small window of data about stand development and stand dynamics (Erdmann 1987). Additionally, climate change and the likely expansion of EAB into extensive black ash forest ecosystems will create novel conditions that have yet to be studied; EAB is a top concern among managers in Minnesota (Windmuller-Campione et al. 2020). It will be extremely important to monitor the expanding mortality from EAB to quantify how EAB impacts ash dominated systems compared to ash in mixed hardwood systems across the eastern US, because ash-dominated stands are most likely to convert to wet meadows and the overall hydrologic changes will be much greater than in mixed ash stands. Additionally, modelers and managers could partner around an adaptive management framework that uses model projections of proposed management strategies to either promote resilience to EAB impacts or actively restore stands after EAB impacts.

380 **5. Conclusions**

381 Our study allowed us to draw several conclusions. 1) Empirical studies can generate critical observational data that  
382 can be used to robustly estimate waterlogging tolerance parameters for simulation models, which improves their projections.  
383 Our results suggest that a very small difference in waterlogging tolerance parameters may determine whether a species lives  
384 or dies in simulated scenarios, underscoring the importance of our study. 2) Mechanistic forest landscape simulation models  
385 can be used to scale up such empirical studies to longer temporal scales and broader spatial scales. 3) AM may be an effective  
386 step toward maintaining the function of wetland hardwoods in the face of EAB. 4) The wetland AM species parameterized  
387 and tested demonstrated substantial survival and biomass accumulation in the landscape case study . 5) Landscape  
388 models thus parameterized provide a powerful tool to conduct simulation experiments involving highly novel situations such  
389 as climate change, invasive (or intentionally migrated) tree species, invasive pests, pioneering management strategies, or all  
390 of these combined.

391  
392 **Appendix A.**

393 **Table A1. Soil texture parameters (Saxton and Rawls 2004) used in the simulations. PTNG represents the potting soil used in the**  
394 **empirical experiment and MUCK is an *ad hoc* lowland soil type used in the landscape case study.**

Soil type code	Full name	Sand	Clay	% Organic matter	Density Factor
SAND	Sand	0.85	0.04	2.08	1
LOSA	Loamy Sand	0.8	0.05	2.33	1
SALO	Sandy Loam	0.63	0.1	2.52	1
LOAM	Loam	0.41	0.19	3.06	1
SILO	Silty Loam	0.15	0.18	3.05	1
MUCK	Muck	0.25	0.14	15	0.9
PTNG	Potting Soil	0.2	0.1	9	1



395

396 **Table A2. Mean monthly weather parameter values used. Each year was the same to avoid confounding annual weather variation**  
397 **with treatment effects.**

Month	Tmax (°C)	Tmin (°C)	Precipitation (mm)	PAR (μmol/m <sup>2</sup> /s)	CO <sub>2</sub> (ppm)
January	-4.64	-16.39	226.89	32.6	412
February	-1.81	-14.53	320.0	23.0	412
March	3.921	-8.42	405.4	45.8	412
April	12.25	-1.16	472.6	65.1	412
May	19.85	4.814	585.7	120.0	412
June	24.62	10.04	632.6	120.0	412
July	26.93	12.73	534.5	120.0	412
August	25.54	11.80	376.6	120.0	412
September	20.47	7.23	233.2	120.0	412
October	14.05	1.68	150.2	120.0	412
November	5.02	-4.56	120.4	55.0	412
December	-2.10	-12.01	148.3	36.9	412

398

**Code availability:** LANDIS-II code is freely available from [www.landis-ii.org](http://www.landis-ii.org).

**Data availability:** Data are available from the authors upon request.

**CRedit author statement**

**Eric Gustafson:** Conceptualization, Methodology, Software, Writing-Original Draft. **Dustin Bronson:** Conceptualization, Methodology, Writing-Review & Editing. **Marcella A. Windmuller-Campione:** Methodology, Writing-Review & Editing. **Robert Slezak:** Methodology, Writing-Review & Editing. **Deahn Donner:** Funding acquisition, Writing-Review & Editing.

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