# PATTERN AND PROCESS OF TREE REGENERATION AND RECRUITMENT IN MANAGED NORTHERN HARDWOOD FORESTS

By

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#### ABSTRACT

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For managed forests which rely on natural tree regeneration for canopy recruitment, abundance and composition of tree regeneration portend future forest structure and diversity. For northern hardwood forests, a geographically widespread forest type in North America, typical single-tree selection (STS) management relies on natural regeneration to promote new cohorts of canopy trees. Harvesting dispersed, select trees every 15 – 20 years, STS generates low light levels intended to promote sugar maple (*Acer saccharum*) and other shade-tolerant tree species in an uneven-aged system. However, following 60 + years of STS implementation in the Great Lakes region, concerning regeneration trends have emerged, namely low densities of sugar maple and low tree species diversity. Additionally, few studies have analyzed age structure under this system to assess its past efficacy in generating uneven-aged forests. The research presented here characterizes regeneration and recruitment outcomes of STS, analyzing data from a uniquely detailed and geographically widespread research project of 141 northern hardwood stands across northern Michigan.

Given the silvicultural focus on regenerating sugar maple, the first two chapters focus on management outcomes for this key species. First, a flexible Bayesian hierarchical model offers insight on patterns of sugar maple regeneration for key size classes as a function of plot and stand level predictors. Our results indicate that sugar maple regeneration is sparse to absent, particularly for size classes actively browsed by deer and recently escaped from the deer browsing zone. The second analysis characterizes age structure for a subset of 51 stands, drawing on 1499 sugar maple trees > 5 cm diameter sampled via basal discs from recently harvested stumps; this analysis provides insight to past patterns of recruitment and establishment. The results suggest little evidence of sugar maple seedling regeneration and canopy ingrowth over the past 60 + years of STS management; instead, stands have highly suppressed saplings plus aging poletimber and sawtimber classes, which are at or quickly approaching economic maturity. Given declines in sugar maple dominance as evidenced by the first two research analyses, the third analysis assesses stand-level tree species diversity and individual species abundance as a function of landscape predictors and size class to shed light on projected future canopy composition. On average, there are approximately three effective common species for seedlings, saplings, and canopy stems at the stand level, and species less desirable for management are occupying growing space in the sapling layer.

Together, these results indicate that STS has been unsuccessful in regenerating or recruiting sugar maple over the past 60 + years, and stands are characterized by a paucity of tree species. Our results support several potential alternative management strategies, including decreasing basal area via more intense harvests, prohibiting deer browsing via natural browsing barriers, or introducing greater diversity of tree species via direct seeding or planting. These results should be considered to improve current management of northern hardwood forests in the Great Lakes region.

To all who supported me along the way, especially my partner and my parents

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#### **CHAPTER 1**

# **INTRODUCTION**

Forests comprise approximately 30% of the world's total land area (Global Forest Resources Assessment, 2015) and provide a variety of cultural, economic, and ecological services (Brockerhoff et al., 2017). Globally, continued provisioning of these services relies on how forests respond to increasing human-mediated stressors, including increasing forest fire activity (Flannigan et al., 2000), invasive plants and pathogens (Dukes et al., 2009; Ramsfield et al., 2016), and frequency and severity of drought (Allen et al., 2010), plus interactions among factors (Dale et al., 2000) and with climate change (Dale et al., 2001). Given trees are sessile, relatively long-lived individuals with limited seed dispersal distances, migration rates have not matched climatic changes, nor are they projected to in the future (Iverson et al., 2004; Zhu et al., 2012), generating an adaptational lag (Aitken et al., 2008). Therefore, high forest biodiversity, which generally increases resiliency (Thompson et al., 2009), is key to future forest resiliency. Current patterns of low density and diversity of tree regeneration (Ramirez et al., 2018; Miller & McGill, 2019; Vickers et al., 2019) for North American forests are thus significant cause for concern.

Covering over 50 million hectares of northeastern United States (Oswalt et al., 2014), northern hardwood forests are an important forest ecosystem. They are currently ecologically dominated by sugar maple (Schulte et al., 2007), which is also one of the most economically valuable species (Linehan & Jacobson, 2005; Duval et al., 2014). Northern hardwood forests can support a variety of species, including deciduous trees such as maples (*Acer spp.*), American beech (*Fagus americana*), oaks (*Quercus spp.*), basswood (*Tilia americana*), aspen (*Populus spp.*), ash (*Fraxinus spp.*), and elm (*Ulmus spp.*), in addition to conifers such as hemlock (*Tsuga*)

*canadensis*), spruce (*Picea spp.*), and fir (*Abies balsamifera*) (Schulte et al., 2007). However, species richness and structural complexity have declined since European colonization (Schulte et al., 2007).

Management of northern hardwood forests in the Great Lakes region has been dominated by a silvicultural system known as single-tree selection. Development of this system began in the 1920's (Kern et al., 2014), in response to a preponderance of young, even-aged forests established following widespread clearcutting and often subsequent slash-fueled fires in the late 19th and early 20th centuries (Whitney, 1987; Dickmann & Leefers, 2016). Research and development of this silvicultural system (Eyre & Zillgitt, 1953; Metzger & Tubbs, 1971) culminated in development of a popular residual structure marking guide manual (Arbogast, 1957) which, along with other resources (e.g., Tubbs, 1977b), has since dominated the application of selection management (Kern et al., 2014). Under STS, select trees from all size classes are removed in harvests every 10 - 20 years (Neumann, 2015), removing unhealthy or illformed smaller diameter trees alongside larger, economically valuable stems. Single-tree gaps generate low-light understory environments (Beaudet et al., 2004) intended to promote sugar maple (Acer saccharum Marsh) and other shade-tolerant tree species (Crow et al., 2002; Angers et al., 2005; Poznanovic et al., 2013). Single-tree selection theoretically provides many benefits, including a steady stream of timber products and a low-impact harvesting system, which is visually appealing.

However, recent studies suggest single tree selection may be yielding undesirable tree regeneration patterns, including lower diversity (Crow et al., 2002; Neuendorff et al., 2007; Powers & Nagel, 2009).and low density of desirable species, particularly sugar maple (Leak, 2006; Neuendorff et al., 2007; Powers & Nagel, 2009; Matonis et al., 2011). Furthermore, few

studies have analyzed age structure in selection-managed northern hardwood forests, instead relying on diameter distribution as a proxy; but age is vital to assess whether recruitment failures are recent or chronic. Given predicted declines in tree fitness (B. M. Rogers et al., 2017) and climatic changes (Byun & Hamlet, 2018) in the Great Lakes region, characterizing patterns of tree regeneration, recruitment, and associated potential driving factors is vital for promoting long-term forest health and sustainability via management.

Management strategies are interacting with a variety of abiotic and biotic factors to influence forest dynamics. Forests in the Great Lakes are fundamentally distributed by postglacial landforms and their associated nutrient availability and soil-water holding capacity (Zak et al., 1986, 1989; Baribault et al., 2010). Underlying soil type is further refined by a three-fold gradient of annual snow fall driven by the lake-effect, influencing tree species distribution and abundance (Henne et al., 2007). Northern hardwood forests vary in composition across the subset of soil types which support these forests. Interacting with the abiotic landscape, a variety of biotic drivers further filter species, including preferential browsing by white-tailed deer (Odocoileus virginianus) (Curtis & Rushmore, 1958; Horsley et al., 2003; Côté et al., 2004; Kain et al., 2011; Matonis et al., 2011; White, 2012; Bradshaw & Waller, 2016) and competition from woody non-tree shrubs (Royo and Carson, 2006; Walters et al., 2020a) and sedge (Randall & Walters, 2019). Invasive pests and pathogens have altered canopy composition and competitive outcomes for a variety of species, including American beech (Fagus grandifolia), white ash (Fraxinus americana), eastern hemlock (Tsuga canadensis), and American elm (Ulnus americana) (Parker & Leopold, 1983; Forrester et al., 2003; Nuckolls et al., 2009; Klooster et al., 2014). Although causal mechanisms between driving factors and northern hardwood forest dynamics are well researched, few studies have analyzed multiple drivers, across the range of

their values, and their associations with northern hardwood forest regeneration structure and outcomes.

The goal of this dissertation is to characterize regeneration and recruitment outcomes in selection managed northern hardwood forests, with particular focus on a key species, sugar maple. Here, we use regeneration to refer to establishment of new tree seedlings and recruitment to characterize the process of small seedlings surviving and growing into taller size classes which have largely bypassed deer browsing and shrub competition, and therefore have a greater likelihood of reaching the overstory, compared to shorter classes. We assess regeneration and recruitment largely by stem densities. This dissertation relies on data from 141 selection-managed northern hardwood forests across northern Michigan, which cover a gradient of deer density, site quality, and canopy structure and composition. These stands are part of a larger research project analyzing regeneration outcomes to alternative silvicultural systems (Walters et al., 2020). This dissertation focuses on pre-harvest vegetation data and stand dynamics to characterize outcomes of decades of selection management in influencing forest structure and function.

In Chapter 2 (*Complex drivers of sugar maple* (Acer saccharum) *regeneration reveal challenges to long-term sustainability of managed northern hardwood forests*), I address one of the most concerning trends to forest management: declining sugar maple regeneration patterns. Exploring this question through a Bayesian modeling framework, I analyzed how a variety of plot and stand level factors influenced sugar maple regeneration outcomes, drawing on vegetation surveys collected in all 141 stands. This chapter tests the hypotheses that sugar maple regeneration will: 1) be highly variable, with several regions of unacceptably low sapling densities; 2) negatively associate with historical deer use, particularly for sapling layers in the

deer browsing zone; 3) increase with stand-level density of mature, seed-bearing sugar maple trees, particularly for seedlings; 4) decrease with increasing woody non-tree vegetation density for the seedling class; 5) decline with higher total canopy density; and 6) increase with higher site quality.

In Chapter 3 (*Sugar maple age structure in northern hardwood stands managed by selection silviculture*), I analyzed whether current low regeneration densities of sugar maple reflect a recent regeneration and recruitment failure or whether recent patterns reflect a chronic issue. To address this question, I analyze basal discs from recently harvested sugar maple stems for a subset of 51 stands, which spanned the extent of the larger study as well as landscape and stand-level drivers. With a collection of 1499 sugar maple tree samples, I addressed the following hypotheses: 1) Stands will be dominated by a canopy cohort of sugar maple stems, approximately 100 - 120 years old; 2) Stands will have one or two main age cohorts (less than the expected four age cohorts) which are > 60 years old; 3) Saplings (5 – 10 cm) will be older (> 60 years) than past studies indicate; 4) Older, sparser saplings will associate with areas of historic high deer density and/or lower site quality; and 5) Age and diameter will have a nonlinear relationship, with a wide range of tree diameters having similar ages.

In Chapter 4 (*Characterizing seedling and sapling species diversity in managed northern hardwood forests portends low stand resiliency*), I considered forest resiliency in a broader context, analyzing stand-level tree diversity with a focus on regeneration. In this analysis, I characterized how diversity varies with size class and with key landscape and stand-level drivers. To infer greater nuance to diversity measures, I also analyzed how individual tree species respond to the same suite of drivers for key regeneration size classes. This chapter addresses the following hypotheses: 1) Tree diversity will be highest in the canopy, followed by the seedling

class, and lowest for saplings class, and will be higher on lower quality versus higher quality sites; 2) Canopy composition will associate with regeneration density and diversity via seed source limitation, particularly for large-seeded species; 3 )Highly palatable tree species, like sugar maple, will negatively associate with deer use for the sapling class and will have greater representation in the canopy or seedling layers versus the sapling layer; 4) Shade intolerant and mid-tolerant species will associate negatively with increasing canopy basal area and will have greater representation in the canopy versus the understory; and 5) Small seeded species will face seedling establishment limitations, negatively associating with hardwood litter coverage, and therefore having greater representation in the overstory than in the understory.

In Chapter 5 (*Outcomes and implications of 60+ years of single-tree selection management in northern hardwood forests*), I summarized the overarching findings of my dissertation. I explore my results in light of implications for single-tree selection management, climate change, and ongoing challenges to the long-term health and functioning of northern hardwood forests.

#### **CHAPTER 2**

# COMPLEX DRIVERS OF SUGAR MAPLE (ACER SACCHARUM) REGENERATION REVEAL CHALLENGES TO LONG-TERM SUSTAINABILITY OF MANAGED NORTHERN HARDWOOD FORESTS

## 2.1 Abstract

Single-tree selection silviculture management of northern hardwood forests relies on natural tree regeneration for long-term sustainability, yet current trends in tree regeneration and recruitment elicit concern. Low densities of economically valuable sugar maple (Acer saccharum Marsh.) in understories are often common, likely driven by many factors, including deer browsing, management-dictated stand structure, and site fertility/moisture regimes. However, landscape sugar maple regeneration patterns and relationships with underlying factors are largely unknown. We quantified associations of spatially varying factors with sugar maple regeneration using detailed vegetation and white-tailed deer winter fecal pellet surveys from 141 northern hardwood stands in Michigan, managed for decades with single-tree selection silviculture. We developed models of plot-level sugar maple regeneration counts for three key size classes as a function of plot- and stand-level predictors, including deer use, forest structure, and site quality. Among our 141 stands, sugar maple seedlings (< 50 cm tall) were consistently abundant, averaging 69,000 stems ha<sup>-1</sup> and present in 76% of plots per stand, on average (25 plots per stand, each 2  $m^2$ ). In contrast to seedlings, small (50 - 137 cm tall) and large (> 137 cm tall and < 5 cm DBH) sugar maple sapling densities were much lower, averaging 2,300 and 1,100 stems ha<sup>-1</sup>, respectively, and occurring, on average, in 31% and 32% of plots per stand (25 plots per stand, each 12.6 m<sup>2</sup>). Under a wide range of potential sugar maple stocking criteria, most stands are understocked in saplings, e.g., only 29% of stands had > 2,500 small sugar maple sapling stems ha<sup>-1</sup> and only

35% of stands had > 1,000 large sapling stems ha<sup>-1</sup>. Based on our models, sapling densities negatively associated with deer use and were most abundant on medium quality sites. Across all size classes, negative associations with subcanopy trees and/or shrub densities suggest light limitation, whereas positive associations with sugar maple canopy trees > 25 cm DBH suggest persistent seed limitations. Overall, our study supports a need for alternative forest and/or deer management strategies over much of the range of northern hardwood forests in Michigan to promote higher densities of large sugar maple regeneration for canopy recruitment. Medium quality sites with abundant large sugar maple canopy trees and low deer browse pressure (for example, the deep snow region in the northern portion of the western Upper Peninsula) are the exception; high sugar maple sapling densities suggest forests in this region are thriving under single-tree selection management.

#### 2.2 Introduction

#### 2.2.1 Management and regeneration dynamics

Sustainable forestry depends on regeneration and canopy recruitment of tree species favored by management. However, temperate forests globally exhibit recent trends of declining regeneration (Ramirez et al., 2018; Miller & McGill, 2019). In the northern United States, many naturally regenerated forests are understocked in regeneration size classes, and most are projected to undergo species compositional shifts (Vickers et al., 2019). Northern hardwood forests, which occur from the Great Lakes region to Maritime provinces (Braun, 1950), are one such naturally regenerated forest type with regeneration and recruitment concerns (Jenkins, 1997; R. O. Miller, 2004; Donovan, 2005). Seedling and sapling layers in managed northern hardwood forests may have lower species diversity compared to unmanaged stands (Neuendorff et al., 2007; Powers & Nagel, 2009), which undermines long-term forest resilience (Thompson et

al., 2009). Declining density of sugar maple (*Acer saccharum* Marsh.) regeneration throughout much of its northern range (Leak, 2006; Neuendorff et al., 2007; Powers & Nagel, 2009; Matonis et al., 2011) is particularly concerning as it currently dominates the canopy (Schulte et al., 2007), is important ecologically as a late successional species (Curtis & McIntosh, 1951), and is highly valued economically (Linehan & Jacobson, 2005; Duval et al., 2014). Given that northern hardwood management systems are designed to promote abundant sugar maple regeneration and recruitment, understanding stand and landscape level factors predictive of sugar maple regeneration that managers can manipulate is key for promoting long-term sustainability.

In the Great Lakes region, low sugar maple regeneration densities exist following decades of single-tree selection management. Many of these second growth stands originated 100+ years ago as primarily even aged, following widespread and intense logging in the 19<sup>th</sup> and early 20th centuries (Dickmann & Leefers, 2016). Since stand origination, species richness and structural diversity have declined, in part driven by natural succession and by selective removal of species for economic reasons (Nyland, 1992; Schulte et al., 2007). Single-tree selection was introduced in the 1950's to convert a preponderance of relatively young, even-aged northern hardwood forests to uneven aged stands and/or to maintain uneven-age structure over the limited extents this structure occurred (Arbogast, 1957). In theory, this regime involves partial stand harvesting of dispersed, individual trees from all size classes, generating small gaps at 10–20year harvest intervals (Neumann, 2015). In practice, single-tree selection prevails within stands but with an increased emphasis in the last couple of decades on including some small group harvest areas within stands, which has been implemented to varying degrees. Given the dominance of single-tree selection within and across stands, we hereafter use the term single-tree selection. Regardless of specific pattern, stand basal area (BA) is typically reduced from ~ 25 -

30 m<sup>2</sup> ha<sup>-1</sup> to 17 - 21 m<sup>2</sup> ha<sup>-1</sup> (Pond et al., 2014; Neumann, 2015; OMNRF, 2015). In theory, this mimics frequent, low-intensity disturbances characteristic of unmanaged northern hardwood stands (Frelich & Lorimer, 1991) but contrasts with high intensity disturbances that led to current forest origination (Whitney, 1987). With consistent regeneration and recruitment, single-tree selection generates a continuous stream of timber products, offering economic stability. Since introduction, single-tree selection has been the dominant management paradigm for northern hardwood forests, particularly in the Great Lakes region (Kern et al., 2014).

Lower-light conditions characterizing single-tree selection (e.g., at 5 m above the ground, ~10-20% of above-canopy photosynthetic photon flux density for the first few years following harvest, (Beaudet et al., 2004) are intended to promote sugar maple, a shade-tolerant species (Tubbs, 1977a), but recent findings challenge this notion. Harvest regimes which generate low (e.g., single-tree selection) and high (e.g., clearcutting) light conditions tend to have lower sugar maple regeneration densities compared to harvests of intermediate canopy openings (e.g., 50% basal area reduction) and light availability (e.g., ~80% of ambient photosynthetically active radiation for shelterwood cutting, Grayson et al., 2012) (Matonis et al., 2011; Cleavitt et al., 2018; Danyagri et al., 2019). Instead of creating long-term light conditions suitable for sapling recruitment, light levels in single-tree canopy gaps may decline via horizontal expansion of gap edge canopy trees faster than advanced regeneration can recruit (Kern et al., 2013), with light levels returning to pre-harvest conditions within 8-11 years (Beaudet et al., 2004). The notion that small gaps are insufficient for recruitment is supported by observation that successful sugar maple recruitment was more likely in larger single-tree gaps ( $\sim 78 \text{ m}^2$ ) with taller (> 4 m) advanced regeneration (Cole & Lorimer, 2005). Since sugar maple saplings decline in shade tolerance and growth rate as their size/age increases (Donoso et al., 2000; Sendall et al., 2015),

vertical growth may stall and mortality increase unless light levels are maintained or increased in subsequent single-tree selection harvests. Over several cutting cycles, use of harvest gaps too small to recruit saplings to the overstory may lead to a surplus of old, suppressed sugar maple and other shade-tolerant species accumulating in the understory. With limited recruitment from sapling classes over several decades of continued single-tree selection management, canopy/subcanopy tree classes become understocked, first in the smaller classes and ultimately in the largest classes (Millington et al., 2011). Consistent with limited recruitment over decades, but not demonstrating cause and effect, low small tree (pole class 10 - 25 cm DBH) density in northern hardwood stands has been identified in single-tree selection managed northern hardwood stands in northern Michigan (Walters et al., 2020b) and broadly at the landscape level (Hanberry & Abrams, 2019).

#### 2.2.2 Local and landscape factors affecting regeneration

Seedling and sapling competition for light is not only influenced by distant overstory canopies but also by understory shrubs and larger sapling classes (Schwinning & Weiner, 1998; Collin et al., 2017). Shrubs broadly impact regeneration outcomes and shift forest stand dynamics (Royo & Carson, 2006). For sugar maple, shrubs inhibit seedling growth, particularly on high quality sites, though they can improve survival during periods of drought (Berkowitz et al., 1995). Dense, tall tree regeneration can also cast deep shade on smaller regeneration below, inhibiting growth and survival even for shade-tolerant species such as sugar maple (Beaudet et al., 2002). For example, American beech (*Fagus grandifolia* Ehrh.) can establish dense sapling layers that associate with low densities of subordinate sugar maple stems, suggesting competitive effects (Hane, 2003; Nyland et al., 2019; Elenitsky et al., 2020); beech's competitive edge is likely exacerbated by beech bark disease, which triggers dense beech sapling thickets (Forrester

et al., 2003). For single-tree selection managed forests, shrubs and taller, dense regeneration layers present potential competitive challenges for sugar maple regeneration.

Fundamentally, sugar maple seedling regeneration is driven by seed availability and site quality, which affects germination, growth, and survival. Seedling density is closely tied to seed production (Bjorkborn, 1979), with larger trees producing more seed (Garrett & Graber, 1995). Sugar maple is also known to regenerate vigorously via stump sprouting after harvest (Forrester et al., 2014). Sugar maple trees (and therefore seed source/stump sprouting potential) are generally found on high nutrient and moderate to high water holding capacity soils (Burger & Kotar, 2003). Key resources potentially varying and limiting the establishment, growth, and survival of sugar maple regeneration include water, nitrogen, and calcium. Variation in some or all of these resources likely underlie variation in site quality and have been found to scale with ecological classification systems (Zak et al., 1989) and habitat classification categories (Walters & Reich, 1997; Baribault et al., 2010). High quality site conditions promote germination success (Tubbs, 1977a), enhance seedling growth (Walters & Reich, 1997), and reduce regeneration mortality rates (Burns & Honkala, 1990; Caspersen & Kobe, 2001; Kobe, 2006), meaning sugar maple has previously and is expected to thrive on high quality sites. Northern hardwood regeneration exhibits differential abundance among site qualities (Elenitsky et al., 2020), though associations can be confounded at smaller spatial scales (Matonis et al., 2011). Since sugar maple canopy growth rates are higher on higher quality sites (Baral et al., 2016) and individual super-producing masting trees tend to associate with higher nutrient availability (Minor & Kobe, 2017), high quality sites may also have greater seed production, leading to compounding effects of site quality and seed source on seedling density.

Additionally, abundant populations of white-tailed deer (*Odocoileus virginianus* Zimmermann, hereafter called deer) influence regeneration patterns. At broad scales, deer are linked to understocked regeneration throughout northern United States (Vickers et al., 2019) and shifting tree regeneration and herbaceous communities via preferential browsing (Horsley et al., 2003; Côté et al., 2004; Bradshaw & Waller, 2016). Deer exclusion experiments in northern hardwood forests have quantified regeneration density and average height reductions caused by deer (Curtis & Rushmore, 1958; Horsley et al., 2003; Kain et al., 2011; White, 2012). Regarding tree regeneration, high deer densities can mitigate positive effects of disturbance (Nuttle et al., 2013), increase optimal gap size (Walters et al., 2016), and shift shrub coverage from competitive to facilitative (Walters et al., 2016). Browsing pressure particularly suppresses small saplings in the browse at-risk zone (<2 m in height; (Walters et al., 2020a), inhibiting recruitment. While direct impacts of deer on vegetation are well-documented, relative importance of continuous variation in browsing pressure over regional landscapes, with other factors also varying, is largely unknown.

Quantifying landscape level patterns of sugar maple regeneration and identifying predictive factors for managed northern hardwood forests is imperative for continued sustainable management; long-term regeneration failure has enormous impacts, economically and ecologically. Although likely mechanisms from small-scale studies are useful, these studies lack predictive power across broad areas. In contrast, geographically widespread studies using public datasets help identify broad patterns but lack detail and rigor to assess all potential predictive factors. In this study, we bridge the gap between geographic breadth and rigorous detail to characterize landscape patterns and identify factors associated with sugar maple regeneration in single-tree selection managed northern hardwood forests, across northern Michigan. We expect

that sugar maple regeneration will: 1) be highly variable, with several regions of unacceptably low sapling densities; 2) negatively associate with deer use, particularly for sapling layers in the deer browsing zone; 3) increase with stand-level density of mature, seed-bearing sugar maple trees, particularly for seedlings; 4) decrease with increasing woody non-tree vegetation density for the seedling class; 5) decline with larger total canopy density; and 6) increase with higher site quality. We interpret analysis results with consideration of management implications and highlight conditions and areas that are particularly favorable or unfavorable for continuing use of single-tree selection.

#### 2.3 Methods

#### 2.3.1 Study area

We quantified vegetation and deer use in 141 managed northern hardwood stands throughout northern Michigan (Walters et al., 2020b). At the start of the study, stands were considered ready for partial harvest by standard single-tree selection criteria (i.e.,  $> 23 \text{ m}^2 \text{ ha}^{-1}$ BA and well stocked in sawtimber classes; Arbogast, 1957). Most of the 141 stands were State owned (n=119) and managed by the Michigan Department of Natural Resources (MDNR), with private industrial sites (Hancock Timber Resource Group, The Rohatyn Group, n=22) comprising the remaining stands. Stands were generally on upland sites and dominated by sugar or red maple (*Acer rubrum* L.). In addition to sugar and red maple, other common northern hardwood tree species included striped maple (*Acer pensylvanicum* L.), balsam fir (*Abies balsamea* [L.] Mill.), yellow birch (*Betula alleghaniensis* Britton), paper birch (*Betula papyrifera* Marsh.), American beech, white ash (*Fraxinus americana* L.), ironwood (*Ostrya virginiana* Mill.), white spruce (*Picea glauca* [Moench] Voss), white pine (*Pinus stobus* L.), bigtooth aspen (*Populus grandidentata* Michx.), white oak (*Quercus alba* L.), red oak (*Quercus*  *rubra* L.), basswood (*Tilia americana* L.), hemlock (*Tsuga canadensis* L.), and American elm (*Ulmus americana* L.).

Northern hardwood forests of Michigan are broadly found on mesic to wet mesic sites with fertile medium-textured upland soils (Dickmann & Leefers, 2016). In the Great Lakes region, nutrient availability and soil moisture tend to positively covary (Zak et al., 1986; Walters & Reich, 1997), and adjacent Great Lakes drive significant patterns in snowfall and overall annual precipitation. Together, these factors are strong drivers of northern hardwood forest distribution (Burger & Kotar, 2003; Henne et al., 2007). Our study design was intended to maximize variation in site quality among stands (Figure 2.1), which we characterized with a habitat classification system (Burger & Kotar, 2003).



Figure 2.1. Locations of 141 single-tree selection-managed northern hardwood stands sampled for sugar maple regeneration and recruitment across northern Michigan by site class. SQ1 stands are poor to poor/medium, SQ2 medium, SQ3 medium/rich, and SQ4 rich to very rich (Burger and Kotar, 2003).

# 2.3.2 Field methods

In summer 2016, we established a permanent 12.14 ha square within each of our 141 sites, which we heretofore refer to as stands, with 25 survey points laid out in a systematic grid; survey points were spaced 69.5 m apart in the grid. We surveyed vegetation in these stands in summer 2017. At each survey point, we established two 1 m<sup>2</sup> quadrats centered 3 m east and west of the survey point, which together represent one plot  $(2 \text{ m}^2)$ . In those quadrats, we counted all tree seedlings, by species, shorter than 50 cm, and ocularly estimated percent cover of seedlings and shrubs. Using the survey point as a common center, we used a 2 m radius circular plot  $(12.6 \text{ m}^2)$  to tally tree regeneration (by species) 50 cm to 137 cm tall (small saplings), and tally and measure DBH (by species) for stems > 137 cm tall and < 5 cm DBH (large saplings). In a 6 m radius circular plot  $(113 \text{ m}^2)$ , we tallied and measured DBH for stems > 5 cm DBH. As we

traversed stands, we recorded herbaceous plant assemblages and key indicator species to assign a dominant site-quality index (Burger & Kotar, 2003) at the stand level.

To assess winter deer use, we conducted pellet surveys in spring of 2017 at a subset of 50 sites, stratified spatially to sample our entire study region. In spring of 2019, entering the first growing season following timber harvest (unrelated to this analysis) of the sites, we surveyed 139 sites using the same protocol. Although pellet surveys are error-prone, they are cost-effective, feasible at large scales, and provide a reasonable approximation of winter deer use (Forsyth et al., 2007; Urbanek et al., 2012; Brinkman et al., 2013). We surveyed linear transects (6 m wide) that totaled 628 m in length, spatially dispersed throughout the stand (Appendix A). Transects were subdivided into 103 segments ~ 6 m long in which we recorded deer pellet occupancy. We conducted surveys from 22 April to 20 May in 2017 and 14 April to 29 May in 2019, prior to emergence of spring ephemerals, and assumed leaf-off to be 01 November for all stands, for both winters. With pellets atop last-years leaf litter, the method assumes pellets were deposited between leaf-off and date of survey. We developed a stand-level deer use model which incorporated climatic and landcover predictors to provide point estimates of deer use for stands which were not surveyed in 2017 (methods in Appendix A).

# 2.3.3 Sugar maple regeneration density modeling framework

Our goal was to estimate parameter values and associated uncertainty for a set of carefully selected ecological factors potentially predictive of sugar maple regeneration count data in key size classes. All analyses and modeling were conducted in R (R Core Team, 2018). We developed a multilevel model (predictors at plot- and stand-level, Table 2.1) to predict plot-level sugar maple counts, which we ran separately for sugar maple counts in three size classes: seedlings (0 – 50 cm tall), small saplings (50 cm – 137 cm tall), and large saplings (> 137 cm tall)

and less than 5 cm DBH). These size classes represent key stages of sugar maple size and canopy recruitment probability with respect to deer browsing pressure (active driver for small saplings, potential legacy effects on large saplings escaped from the deer browse zone, and less tied to seedlings), seed source (anticipated greatest association with seedlings). Large saplings are considered to have highest potential for canopy recruitment, seedlings the lowest. Plot sizes were  $2 \text{ m}^2$  for seedlings and 12.6 m<sup>2</sup> for small and large saplings.

Plot-level predictors in our model included percent shrub coverage (SHRUB, all non-tree woody species, generally < 100 cm tall in our stands) and total stem count of relatively taller regeneration classes, all species (SAP1, small saplings; SAP2, large saplings; and SUB, subcanopy trees, 5 - 10 cm DBH). Stand level predictors included modeled average estimated likelihood of deer use (DEER, see Appendix A for deer modeling details; Figure 2.2), four ordinal categories of site quality (SQ1-4, 1 being poorest, see Appendix B for details), average total BA trees > 10 cm DBH (BA, proxy for forest canopy effect on light availability and microclimate), and average BA of sugar maple trees > 25 cm DBH (SMBA, proxy for seed availability, size criteria determined using methods in Appendix C). BA and SMBA were averaged at the stand level to better reflect management scale and because a 6m radius plot may not fully capture canopy competition or seed availability.

	Name	Description	Mean (Range)
Plot level	SAP1	Total stem counts of small saplings (> 50 cm tall and < 137 cm tall) per plot (12.6 m <sup>2</sup> ), all species	6 (0 – 116)
	SAP2	Total count of large saplings (> 137 cm tall and < 5 cm DBH) per plot (12.6 m <sup>2</sup> ), all species	4 (0 – 38)
	SUB	Total count of subcanopy trees $(5 - 10 \text{ cm DBH})$ per plot (12.6 m <sup>2</sup> ), all species	3 (0 – 33)
	SHRUB	Ocular estimate of percent ground obstructed by shrubs, viewed from above $(0 - 100\%)$	2 (0-93)
Stand level	BA	Basal area all species $> 10$ cm DBH (m <sup>2</sup> ha <sup>-1</sup> )	26 (19 – 44)
	SM.BA	Basal area of sugar maple > 25 cm DBH ( $m^2ha^{-1}$ )	13 (0 – 27)
	DEER	Modeled deer use $(0 - 100\%)$	25 (4 - 76)
	SQ1	Poor to poor/medium site quality $(n = 39)$	-
	SQ2	Medium site quality $(n = 22)$	-
	SQ3	Medium/rich site quality $(n = 47)$	-
	SQ4	Rich to very rich site quality $(n = 33)$	-

Table 2.1. For our multilevel models of the plot-level count of sugar maple regeneration (seedlings, small saplings, and large saplings), we included predictor variables at both plot and stand levels. The table lists a brief description, the abbreviation used, the mean, and the range of parameter values.


Figure 2.2. Model estimated stand-level deer use, winter of 2017. Zero represents a stand with no estimated winter deer use (0 segments had fecal pellets), while 100 indicates a high use stand where all 103 transect segments (each segment  $\sim 6m \times 6m$ ) had fecal pellets.

We checked for interactive effects between site quality and three broad regions of our study (Elenitsky et al., 2020) and between site quality and deer use (selective browsing pressure and/or deer occupancy within region or site class may be altered via differences in species assemblages of potential browse species); analyses did not support adding any interactive effects (Appendix C). Multicollinearity among the predictor variables was assessed using variance inflation factors (via vif in the car package (Fox & Weisberg, 2011)); at a threshold of 2, none warranted removal. To understand how stands covary in their characteristics (e.g., do high quality stands tend to have more canopy sugar maple trees?), we calculated a correlation matrix based on Kendall's tau statistic (n=141), averaging plot-level predictors to analyze all at the stand level; p-values (< 0.05 significance threshold) were calculated using cor\_pmat from the rstatix package (Kassambara, 2020). We also calculated summary statistics to characterize our

response data (sugar maple regeneration), including occurrences, average densities, relative abundance, and boxplots of regeneration density vs. site quality.

For each regeneration size class, we developed five candidate models (Appendix C for details on model formulation and associated diagnostic checks), including a null model and various inclusions of quadratic terms to account for non-linearity, a common approach in regeneration modeling (e.g., Horsley et al., 2003; Schwarz et al., 2003; Danyagri et al., 2019; specifically for deer use effects: Horsley et al., 2003; Tremblay et al., 2006; Bradshaw & Waller, 2016). We selected models based on deviance information criterion (DIC) comparison, which reflects model complexity and fit (Spiegelhalter et al., 2002). For each size class, we used the following model specification to estimate sugar maple stem counts *y* in stand *i* and plot *j*:

Plot level
$$y_{ij} \sim NegBin(\alpha_i + \beta_p X_{pij}, \delta_i)$$
Stand level $\alpha_i \sim Norm(\mu + \beta_s X_{si}, \sigma^2)$ 

 $y_{ij}$  is sampled from a negative binomial distribution with mean  $\alpha_i + \beta_p X_{pij}$ , where  $\alpha_i$  is a stand-specific intercept and  $\beta_p$  and  $X_{pij}$  represent the vector and model matrix for plot-level predictors SHRUB, SAP1 (for seedling model), SAP2 (for seedling and small sapling model), and SUB plus any included quadratic terms (Table 2.1).  $\delta_i$  is a stand-specific, plot-level dispersion parameter.  $\alpha_i$  is drawn from the normal distribution with mean  $\mu + \beta_s X_{si}$ , where  $\mu$  is the grand mean and  $\beta_s$  and  $X_{si}$  represent the vector and model matrix for stand-level predictors BA, SM.BA, DEER, SQ1, SQ3, and SQ4 plus any included quadratic terms (SQ2 included in the intercept, Table 2.1).  $\sigma^2$  represents stand-level variation. We estimated stand-specific dispersion parameters to account for variation among stands in plot-level variance. We mean centered all predictors and divided by two standard deviations to enable comparison (Gelman & Hill, 2007).

Predictive inference was based on 10,000 post burn-in samples from three Markov chain Monte Carlo (MCMC) chains using the RJAGS package (Plummer, 2018), which is an interface to JAGS (Just Another Gibbs Sampler) software. We assessed model convergence using Gelman Rubin diagnostics (coda package), trace plots, and residuals vs. predicted plots. Histograms of predicted zeros compared to true data zeros for each model indicated that the models adequately predicted absences, indicating that the negative binomial was an appropriate fit. We checked for spatial autocorrelation in model residuals, both plots within individual stands and among stands, using variograms and found no residual spatial autocorrelation in the model (Appendix C). To visualize the model results, we simulated model predictions for 1,000 iterations across the range of predictor values for each significant continuous predictor, setting all other predictors at their mean and using average dispersion parameters and stand-level intercepts. To plot the model predictions, we applied a loess smoothing curve on predicted means and 95% lower and upper confidence intervals.

## 2.4 Results

# 2.4.1 Sugar maple regeneration characteristics

Stand-level sugar maple regeneration was highly variable within and among size classes (Table 2.2). For all three size classes, frequency distributions of plots with sugar maple were right skewed with wide ranges (Table 2.2), indicating that most plots had few or no stems and a small percentage had high stem counts (Table 2.2). Sugar maple seedlings (0 – 50 cm tall) were widespread (present in 99% of stands and 76% of all plots) and abundant (average 14 stems per 2  $m^2$  plot, or 70,000 stems ha<sup>-1</sup>) (Table 2.2). Small and large sugar maple saplings were also widespread, occurring in 79 and 90% of stands but only occurred in 31 and 32% of total plots, respectively (Table 2.2). In 8% of our stands, our surveyed area failed to capture either a single

small or large sugar maple sapling. Average counts for small and large saplings were 3 and 1 stems per 12.6 m<sup>2</sup> plot, respectively, or approximately 2300 and 1090 stems ha<sup>-1</sup>. Median densities for small and large sapling stems were lower than the mean at 600 stems ha<sup>-1</sup> and 500 stems ha<sup>-1</sup>, respectively (Table 2.2). We observed regional variation in sugar maple stand sapling densities (seedling densities were more uniformly dispersed, Figure 2.3A). For small saplings, the northern Lower and southern half of Upper Peninsula had low densities, with values  $\geq$  10,000 stems ha<sup>-1</sup> only found in the northern half of the Upper Peninsula (Figure 2.3B). Large sapling followed a similar, but somewhat weaker pattern as higher densities did exist in some stands in the southern half of the Upper Peninsula as well as in the Lower Peninsula (Figure 2.3C).

Plot level	S	tems per p	olot	% re	lative abu	Plot occurrences			
	Mean	Median	Range	Mean	Median	Range			
Seedlings	14	5	0-245	60	75	0 - 100	2675 / 3524 (76%)		
Small saplings	3	0	0 - 100	32	0	0 - 100	1082 / 3524 (31%)		
Large saplings	1	0	0 - 34	30	0	0 - 100	1130 / 3524 (32%)		
Stand	Thous	sand stems	s per ha	% re	lative abu	ndance	Stand occurrences		
Stand level	Thous Mean	sand stems Median	s per ha Range	% re Mean	lative abu Median	ndance Range	Stand occurrences		
Stand level Seedlings	Thous Mean 69	sand stems Median 47	<b>5 per ha</b> <b>Range</b> 0 - 434	% re Mean 64	lative abu Median 71	ndance Range 0 - 99	<b>Stand occurrences</b> 140 / 141 (99%)		
Stand level Seedlings Small saplings	Thous   Mean   69   2.3	Median 47 0.6	<b>5 per ha</b> <b>Range</b> 0-434 0-22	% re Mean 64 34	lative abu Median 71 25	<b>Range</b> 0 - 99 0 - 98	Stand occurrences 140 / 141 (99%) 111 / 141 (79%)		

Table 2.2. The mean, median, and range of sugar maple regeneration densities for seedlings, small saplings, and large saplings at the plot and stand level. Seedlings are 0-50 cm tall, small saplings are 50 cm - 137 cm, and large saplings are 137 cm tall or greater and up to 5 cm DBH. Densities are presented in stems per plot ( $2 \text{ m}^2$  plot for seedlings,  $12.5 \text{ m}^2$  plot for small and large saplings). Plot occurrences indicates the number of surveyed plots which contained at least one sugar maple. Stand occurrences indicates the number of sites which contained at least one sugar maple stem in any of the plots surveyed in that stand. The percent relative abundance refers to the percentage of sugar maple regeneration stems relative to the total density of stems in that size class; for plot percentage relative abundance, plots with zero total stems in a size class were omitted.



Figure 2.3. Average stand-level density of sugar maple seedlings (0 - 50 cm tall; A), small saplings (50 - 137 cm tall; B), and large saplings (>137 cm tall and <5 cm diameter breast height) based on data collected for 141 northern hardwood stands, summer 2017.

## 2.4.2 Stand characteristic covariation

Among variables used to predict sugar maple regeneration densities, we found that deer use, shrub cover, sugar maple canopy BA, and site quality positively covaried in our stands (Figure 2.4). These variables generally negatively covaried with density of total understory vegetation (including large saplings and subcanopy trees). Additionally, total stand BA positively covaried with canopy sugar maple BA and negatively covaried with shrub cover.



Figure 2.4. Kendall's tau correlation matrix of predictor variables used to model stand level sugar maple regeneration for Michigan (non-significant (p>0.05) correlations are labeled with an x, positive associations are blue, and negative associations are red with parallel lines). Variables include total stem density of small saplings (SAP1), total density of large saplings (SAP2), total density of trees 5 – 10 cm DBH (SUB), deer usage (DEER), site quality (SQ, treated as continuous from 1-4), shrub coverage (SHRUB), basal area of sugar maple trees > 25 cm (SM.BA), and total stand basal area (BA).

## 2.4.3 Sugar maple regeneration models

For sugar maple seedling counts, the model intercept was ~2.5 on the log scale,

predicting occurrence more often than absence. At the plot level, seedling counts negatively

associated with shrub cover, and exhibited quadratic relationships with trees 5-10 cm DBH, and

small and large sugar maple saplings (Figure 2.5). Sugar maple seedlings generally increased with total small sapling counts (eventually plateauing) but generally decreased with total large saplings counts, total sub-canopy tree counts, and percent shrub cover (Figure 2.6). At the stand level, sugar maple seedling count increased with BA of canopy sugar maple trees until ~18 m<sup>2</sup> ha<sup>-1</sup>, then declined (Figure 2.6). There were no clear overarching associations with site quality (Figure 9), though stand-level densities of sugar maple seedlings on medium/rich sites (SQ3) were significantly lower than on medium sites (SQ2, included in the intercept) (Figure 2.5). Deer use and total stand basal area were not significantly associated with seedling counts. Our top-ranked sugar maple seedling model ranked higher than the null model (Appendix C).



Figure 2.5. Model parameter estimates for predictors of sugar maple regeneration, with 95% Bayesian credible intervals. Variables include total stem density of small sugar maple saplings (SAP1), total density of large sugar maple saplings (SAP2), total density of trees 5 - 10 cm DBH (SUB), shrub coverage (SHRUB), deer usage (DEER), total stand basal area (BA), basal area of sugar maple trees > 25 cm (SM.BA), site quality (SQ1 is poor, SQ3 medium/rich, SQ4 rich to very rich, and SQ2 medium included in the intercept), and model intercept (MU), which represents average plot-level stem count, on the log scale, when all predictors are at average values. Quadratic terms (indicated by superscripted 2) included in the final models portrayed with dashed lines.



Figure 2.6. Model estimates for the relationship of sugar maple seedlings per plot  $(2 \text{ m}^2)$  with the plot-level densities of small saplings (SAP1), large saplings, (SAP2) and all sub-canopy trees 5 - 10 cm DBH (SUB); the stand-level basal area of sugar maple canopy trees larger than 25 cm DBH (SM.BA); and the plot-level density of shrubs (SHRUB). Mean (black line) and 95% Bayesian credible intervals (gray lines) are displayed, based on the average dispersion parameter estimate.

Small and large sugar maple sapling counts associated similarly with predictor variables. Overall intercepts were not significantly different from zero, meaning both models more often predicted absence than a count (Figure 2.5). Counts of small and large sugar maple saplings decreased with increasing deer use, increased with basal area of sugar maple trees > 25 cm DBH at lower basal areas, and were unassociated with overall stand BA (Figures 2.5, 2.7, 2.8). Sugar maple saplings were most abundant on low (SQ1) or medium quality sites (SQ2, included in the intercept) and lowest on medium/rich (SQ3) and rich sites (SQ4) (Figure 2.9). Small sugar maple saplings were positively related to total large saplings (weakly at low density), and large sugar maple saplings showed no association to subcanopy trees (Figure 2.7). Similar to seedlings, small saplings declined with increasing counts of subcanopy trees. Small and large sugar maple saplings differed in that shrub density was weakly negatively related to large sapling counts (Figure 2.8) and unassociated with small saplings. Both models ranked higher than the null model (Appendix C).



Figure 2.7. Model estimates for the relationship of sugar maple small saplings per plot (12.6 m<sup>2</sup>) with large saplings per plot (SAP2), subcanopy trees per plot (SUB), deer use (DEER), and basal area of sugar maple canopy trees larger than 25cm DBH (SM.BA). Mean (black line) and 95% Bayesian credible intervals (gray lines) are displayed for the average dispersion parameter estimate.



Figure 2.8. Model estimates for the relationship of sugar maple large saplings per plot (12.6 m<sup>2</sup>) with deer use (DEER), basal area of sugar maple canopy trees larger than 25cm DBH (SM.BA), and shrub coverage (SHRUB). Mean (black line) and 95% Bayesian credible intervals (gray lines) are displayed for the average dispersion parameter estimate.



Figure 2.9. For each sugar maple regeneration class, boxplots of stand-level average density by site quality category (*sensu* Kotar and Burger 2003), northern Michigan, 2017. Site quality categories include poor (SQ1), medium (SQ2), medium to rich (SQ3), and rich to very rich (SQ4) in the model. Middle lines represent median values, the lower and upper hinge represent the first and third quartiles, and the whiskers represent values no further than 1.5 of the interquartile range (distance from first to third quartile); individual points are outliers.

# 2.5 Discussion

Widespread and growing evidence suggests understocked sugar maple regeneration in single-tree selection-managed northern hardwood forests (Jenkins, 1997; Miller, 2004; Donovan, 2005; Leak, 2006; Neuendorff et al., 2007; Powers & Nagel, 2009; Matonis et al., 2011), with uncertainties regarding full extent and associated predictive factors; this information is crucial for long-term management success. Detailed studies which have elucidated mechanism tend to be limited in geographic scope and stand characteristics, while broad-scaled analyses gain power but may lack the detailed information needed to identify ecological drivers. Drawing on established theory and previous insight, we bridged depth and breadth by surveying 141 singletree selection managed northern hardwood stands across northern Michigan. Our results confirm widespread and concerning trends in sugar maple regeneration and recruitment. Sugar maple regeneration densities were variable for key size classes and low across wide geographic ranges and stand characteristics. While densities of sugar maple seedlings were generally high (though potentially limited on some sites by low mature sugar maple seed sources and/or competing taller vegetation), densities of small and large sugar maple saplings were overall low, particularly in portions of the northern Lower Peninsula and the central southern Upper Peninsula. Densities in the sapling classes were approximately three orders of magnitude less than for the seedling class. In 20% of our stands, our surveyed area (totaling 315 m<sup>2</sup> per stand) did not include a single sugar maple small sapling (50 cm - 137 cm tall).

We found evidence of potential sugar maple regeneration limitation by both deer browsing and seed availability in our stands. Negative associations of deer use with small saplings suggest proximal/recent deer browsing impacts; effects for large saplings that have largely escaped deer browse pressure may indicate long-term legacies of deer browsing

inhibiting sapling canopy recruitment, which aligns with previous studies (Horsley et al., 2003; Kain et al., 2011; Matonis et al., 2011; Bradshaw and Waller, 2016). We found no association between deer use and sugar maple seedling count, unsurprising given that shorter regeneration is often less browsed (perhaps protected by winter snow) and/or maintained in that size class by persistent browse pressure (Saunders & Puettmann, 1999; Randall & Walters, 2011). Our results also suggest persistent legacies of seed limitation in larger size classes. As expected (Bjorkbom, 1979), seedling count increased with sugar maple canopy BA (up to ~  $20 \text{ m}^2 \text{ ha}^{-1}$ ); the decline of seedlings at higher sugar maple canopy BAs may point to potential light limitation at high densities of canopy sugar maple, as sugar maple trees cast deep shade (Canham & Burbank, 1994). However, general positive associations of small and large sugar maple saplings with sugar maple canopy BA suggest that despite other pressures, such as deer, stands with greater seed source tend to have more sugar maple saplings. Promoting abundance of sugar maple trees > 25 cm DBH, up to densities of 20 m<sup>2</sup> ha<sup>-1</sup>, may lead to greater densities of sugar maple saplings even in regions of high deer populations.

Light limitation from shrubs (Matonis et al., 2011; Kern et al., 2013) and larger tree regeneration (Schwinning & Weiner, 1998) influences regeneration outcomes in northern hardwood stands. Our results support previous work, with negative associations between sugar maple regeneration with competing larger regeneration (excluding the relative next largest size class) and/or shrub density. We attribute positive or neutral associations between a given sugar maple size class and total counts of the relative next largest size class (e.g., sugar maple seedlings positively associated with total small saplings, but negatively with large saplings and subcanopy trees) to categorizing a continuous distribution of regeneration heights for a species of generally high relative importance (also found in Elenitsky et al. 2020). We did not find evidence

supporting light limitation from canopy BA, previously identified as limiting to sugar maple regeneration (Matonis et al., 2011; Cleavitt et al., 2018; Danyagri et al., 2019). However, though not significant, all three mean stand BA parameter estimates were negative in our model; we may have failed to detect a significant association due to the distribution of stand BA included in our study with fewer stands at very high BA (range  $19 - 44 \text{ m}^2 \text{ ha}^{-1}$ , mean 26 m<sup>2</sup> ha<sup>-1</sup>). Stand BA also does not represent light reaching the forest floor well when subcanopy or understory vegetation is dense or when composition of canopies (with different canopy light transmissivities among species) vary among sites (Beaudet et al., 2004).

In our study, sugar maple sapling densities diverged from traditional affinity for high quality sites (Burger & Kotar, 2003) (Figure 2.9). Although high quality sites generally enhance regeneration growth (Walters & Reich, 1997) and reduce mortality rates (Burns & Honkala, 1990; Caspersen & Kobe, 2001; Kobe, 2006), small and large sugar maple sapling abundance was generally lowest on rich sites. Seedlings, in contrast, did not demonstrate much significant variation by site quality; thus, abundant seedlings on high quality sites may not translate to abundant saplings. One possible explanation is higher quality sites had higher deer use (Figure 2.4), likely due to the spatial distribution of site qualities (e.g., in the western Upper Peninsula, site quality and deer use both go from low to high along a north-south gradient) (Figure 2.1). Deer and site quality were also confounding in Matonis et al. (2011) which comprised a subset of our study area (also in the western Upper Peninsula). Lower sapling counts on high quality sites could also be associated with deeper shade, driven by the higher leaf area index (LAI) generally associated with higher quality sites that is largely driven by species with high LAI on those sites (e.g., sugar maple) (Canham & Burbank, 1994; Fassnacht & Gower, 1997). More rapid sugar maple growth on high quality sites (Baral et al., 2016) could result in more rapid closure of small

canopy gaps, reducing canopy recruitment potential and increasing understory mortality. It is also possible that out categorical site quality indicators were too coarse to adequately capture variation in resources actually underlying seedling establishment, growth, and survival over sites (e.g., nitrogen, calcium, water); furthermore, site-quality as a stand-level predictor may fail to capture finer scale nutrient variability within stands which could affect sugar maple regeneration.

While our study suggests that some combination of deer use, seed availability, light competition, and site quality drive sugar maple regeneration dynamics, our ability to draw conclusions on causal mechanisms is limited because our data are from a natural, snapshot in time experiment (Diamond, 1986). We addressed this potential pitfall by carefully choosing model predictors grounded in ecological theory. Due to the large number of stands and plots, our plot sizes were also relatively small, which may lower precision of our density estimates for a single plot. We overcame deer use sampling limitations by developing a deer use model (Appendix A) and using evidence-based statistical modeling approaches to account for nonlinearities. Despite potential limitations, our study bridges the gap between broad-scale and detail-oriented, with a wide geographic extent and broadly varying site characteristics complemented by a rigorous and fine-scaled sampling design. This affords rare insight based on statistical power in understanding dynamics of managed northern hardwood forests.

## 2.6 Implications for Management

Although Arbogast's (1957) manual is considered definitive for stocking of trees > 5 cm DBH, no single definitive standard exists for stems < 5 cm DBH necessary to maintain canopy ingrowth under single-tree selection of northern hardwoods. Older studies frequently evaluated stocking of regeneration irrespective of height class (Leak & Wilson Jr., 1958; Metzger & Tubbs, 1971) or as a total density of stems weighted by height class (McWilliams et al., 1995).

Recognizing deer browsing as a formidable barrier to recruitment in some areas, size class must be included when analyzing regeneration success (Walters et al. 2020a). Furthermore, applying standards that were generally developed for even-aged forests, is likely inappropriate for unevenaged managed northern hardwood stands. Even-aged stocking guides suggest 2850 stems ha<sup>-1</sup> for saplings 25 - 137 cm tall, and 950 stems ha<sup>-1</sup> for saplings > 137cm tall to < 5 cm DBH (Elenitsky et. al. 2020), or 2500 stems ha<sup>-1</sup> for saplings 1.5 - 9.5 cm DBH (Arseneault et al., 2011). Adequate uneven-aged stocking for Arbogast's smallest size class (~ 4-11 cm DBH) is 494 stems ha<sup>-1</sup> after harvest. Within a range of likely stocking guidelines, managers use varying criteria based on harvesting timetables and site-specific factors.

When using density of sugar maple saplings to evaluate stocking levels, we found that our stands were largely understocked across conservative criteria (Figure 2.10). For example, based on criteria in Elenitsky et al. (2020), only 29% of stands had > 2500 sugar maple small sapling stems  $ha^{-1}$  and 35% of stands had >1000 large sapling stems  $ha^{-1}$  (Figure 2.10). For a majority (> 50%) of stands to be considered adequately stocked in sugar maple regeneration, thresholds would need to be ~700 stems  $ha^{-1}$  for small saplings and ~500 stems  $ha^{-1}$  for large saplings, the latter is equivalent to Arbogast's guideline for density of total stems 4 -11 cm DBH. In a silviculture system intended to produce ample sugar maple regeneration, single-tree selection silviculture appears to be failing for many areas in Michigan. It is important to note that since our study did not incorporate unmanaged forests as controls, our results do not necessarily conclude that single-tree selection stands are faring worse than unmanaged northern hardwood stands or that single-tree selection alone has driven these issues. Our finding of widespread understocking differs from Vickers et al. (2019) which found that northern hardwood forests in the Great Lakes region were well stocked. Vickers et al. (2019) included all tree species to evaluate stocking

thresholds, including American beech (which cannot presently recruit in high numbers to the canopy due to beech bark disease).



Figure 2.10. Percentage of surveyed stands which would be considered stocked in small (50-137 cm tall) and large (137 cm tall to 5 cm DBH) sugar maple saplings, based on average stand densities, is plotted as a function of stocking criteria thresholds (100% stocking at a threshold of 0 stems ha<sup>-1</sup>, not shown). Red dashed lines provide an example of figure interpretation: at stocking criteria of 2500 stems ha<sup>-1</sup> for small sugar maple saplings, 29% of surveyed stands are stocked, and at 1000 stems ha<sup>-1</sup> for large sugar maple saplings, 35% of stands are stocked.

Given pervasive understocking of sugar maple sapling regeneration and disparate densities of seedlings vs. saplings, it is unsurprising that relative importance of sugar maple saplings is lower than in seedling (Table 2.2) or canopy tree (Table 2.1) classes. Our count data furthermore do not distinguish stems based on vigor or age; old, suppressed regeneration may never recruit to the canopy, even with release from competition (Donoso et al., 2000), artificially inflating sugar maple regeneration density or importance. This may be most pronounced in our large sapling class where many of the stems appeared old and not vigorous (C. Henry personal observation). Altered patterns of relative abundance can lead to compositional shifts (Vickers et al., 2019) which has significant impacts on ecological structure and economic value of northern hardwood forests. Sugar maple replacement with non-viable canopy species, such as beech bark disease infected American beech saplings and emerald ash borer impacted white ash, or by sub-canopy species, such as ironwood, could portend changes in forest structure and loss of a closed canopy (Bohn & Nyland, 2003; Matonis et al., 2011; Bannon et al., 2015; Danyagri et al., 2019; Elenitsky et al., 2020). American beech saplings have particularly been demonstrated to impact understory structure in our region (Elenitsky et al., 2020) and, as one of the most common species in our surveyed stands, may pose future challenges to management. These trends represent significant management challenges in many areas should single-tree selection be maintained as the de facto management system in northern hardwood forests and these trends continue.

Our model results highlighted the importance of deer use, site quality, light limitation, and seed availability on sugar maple regeneration and recruitment, and there are several potential management solutions. First, more intensive canopy harvesting regimes that increase understory light availability may improve sugar maple (as well as overall tree diversity) regeneration outcomes (Kern et al., 2017; Webster et al., 2018). Silvicultural systems to potentially achieve this outcome include even-aged shelterwood and seed tree systems, or uneven-aged systems that emphasize larger group selection openings and patch-cuts (Sage et al., 2003; Walters et al., 2016; Hupperts et al., 2020; Walters et al., 2020a). Notably, however, more open overstories are likely to promote greater shrub densities (Walters et al., 2016) and growth of undesirable (for management) advance regeneration (e.g., beech, ash) which compete with sugar maple seedlings and saplings. Thus, it may be necessary to combine more intense overstory harvests with understory treatments aimed at controlling shrubs and undesirable regeneration.

In addition to novel silviculture in these systems, treatments may need to include tactics that decrease deer use or browsing pressure on regenerating trees. Although direct management of the deer herd at large scales is difficult to implement and often socially unacceptable, physical barriers, such as felled treetops, may protect regeneration from deer browse (Grisez, 1960;

Pellerin et al., 2010; Hagge et al., 2019). Our research also suggests that maintaining larger, seed-bearing sugar maple trees in the stand also positively influences stocking of sugar maple regeneration. Individual managers may select different criterion for designating successful regeneration, but a wide range of thresholds for sugar maple sapling stocking indicates widespread regeneration failure in Michigan. We recommend managers critically examine regeneration in multiple height classes and consider underlying regional drivers to assess potential long-term trends of forest composition and structure; for many regions and forests, continuing with business as usual may lead to fundamental ecosystem shifts.

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#### **CHAPTER 3**

# SUGAR MAPLE AGE STRUCTURE IN NORTHERN HARDWOOD STANDS MANAGED BY SELECTION SILVICULTURE

## 3.1 Abstract

Single-tree selection (STS) silviculture has dominated management over the last six decades in northern hardwood forests of the Great Lakes region. Over time, periodic partial harvests are assumed to promote well stocked natural regeneration, resulting in balanced uneven age stand structure with sustainable harvest volumes. However, recent studies indicate understocked tree sapling size classes for species desirable for management, including dominant, economically valuable sugar maple (Acer saccharum) suggests STS in not working, at least recently, in some regions. However, few studies have analyzed sugar maple size-age structure across all size classes to determine whether STS has chronically failed to recruit new age cohorts over the last 60 years or is a recent event. Here, we analyzed stand-scale age structure for 51 managed NHF stands located in the Upper and Lower Peninsulas of Michigan via aging basal discs of 1499 sugar maple trees > 5 cm diameter at breast height (DBH). Our results indicate most stand canopies are dominated by 90 - 120-year-old stems, with low densities of relatively old (> 60 years) saplings (stems 5 - 10 cm DBH). Most cohorts of trees < 90 years old are understocked by uneven-aged stocking guides. For most stands, age and diameter are non-linearly related with a less diverse age structure than assumed when using diameter as a proxy for age. Among stands, by traditional stocking diameter classes, Arbogast's sapling size class (5 - 11.4 cm DBH)averages 66 years old, poletimber (11.4 – 24 cm DBH) 91 years, and sawtimber (> 24 cm DBH) 106 years. Areas with greater January precipitation (snowfall) and lower annual temperatures, where deer populations have been historically low, generally have younger saplings. Our results

offer little evidence to suggest STS has promoted well stocked ingrowth of sapling recruits over 60 + years, i.e., current STS management is unsustainable in the long term in some regions. Our results add a new dimension to the growing body of literature highlighting the need for alternative management of northern hardwood forests in the Great Lakes region.

## 3.2 Introduction

Single-tree selection (STS) has been the predominant management regime for northern hardwood forests (NHF) in the Great Lakes region for nearly six decades (Arbogast 1957, Kern et al., 2014). STS harvests remove individual trees from all diameter classes every 10 - 20 years (Neumann, 2015), creating mostly small harvest gaps (e.g., the width of a dominant tree's canopy) with the regimen expected to maintain or create (from forests initially even-aged) then maintain a balanced uneven-aged structure. Partial harvesting generates low-light understory environments (Beaudet et al., 2004), which are intended to promote economically valuable (Linehan & Jacobson, 2005; Duval et al., 2014) and ecologically dominant (Schulte et al., 2007; Walters et al., 2020b; Henry et al., 2021) sugar maple (Acer saccharum Marsh) and other shadetolerant tree species (Crow et al., 2002; Angers et al., 2005; Poznanovic et al., 2013). Despite long-term research and development (Eyre & Zillgitt, 1953; Metzger & Tubbs, 1971), current tree sapling-class patterns in NHF suggest STS may be failing to recruit well stocked stems of desirable species, particularly sugar maple (Leak, 2006; Neuendorff et al., 2007; Powers & Nagel, 2009; Matonis et al., 2011; Henry et al., 2021). What is not known is if sapling-class regeneration failure is a recent phenomenon or indicative of longer-term failure of STS to develop new cohorts following partial harvests.

A means of probing this question is to compare existing diameter structures with those expected under uneven-aged STS management and assume that age and diameter are related.

However, while diameter structure is important (especially as the currency of stocking guides), shade tolerant sugar maple can persist for decades as suppressed, small diameter stems (Gravel et al., 2011), such that age may vary little among size classes. Employing diameter as a proxy for age may therefore lead to overly optimistic assessments of the uneven-aged character of managed stands. Assessing both diameter and age, and their relationship across size classes, is key for assessing validity of size-based structural guides for STS and the efficacy of STS in promoting regeneration cohorts and recruitment among size/age classes. Here, we use regeneration to refer to the process of establishing a new tree cohort in response to partial harvesting, recruitment as the process of trees transitioning to larger size classes and/or life history stages, and cohort to characterize the group of seedlings/saplings which move into stems > 5 cm DBH per harvesting event. Few studies have directly analyzed sugar maple age structure (Table 3.1), and most are limited in geographic scope and/or diversity of stand structures and underlying landscape-scale drivers. Such analyses, conducted over large regions and considering factors potentially driving pattern, are necessary to determine the extent and reasons for (or associations with) STS successes and failures.

Source	Location	Stand history	DBH (cm) range	Sampling height (cm)	<b>I</b> lstands	<b>R</b> samples	Age-Diam.	Age at 10 cm DBH	${f R}^2$
Tubbs (1977a)	Michigan, USA	OG	5-81	-	1	60	С	100	N/A
Leak (1985)	New Hampshire, USA	OG	4 – 55	S	1	47	Q	50	0.47
Goldblum and Rigg (2002)	Ontario, Canada	OG	5 - *	30	8	*	*	*	*
Tubbs (1977a)	Michigan, USA	SEL	2.54 - 63	30	1	60	L	48	0.94
Kenefic & Nyland (1999)	New York, USA	PAR T	2-74	137	1	96	Q	27	0.81
Dey <i>et al</i> . (2017)	Wisconsin, USA	SEL	35 - 85	S	4	60	L	N/A	0.64
Harmala (2021)	Michigan, USA	VAR	> 30 cm tall - 12.7 cm DBH	137	8	50	L	42 - 70	0.82 - 0.99
Harmala (2021)	Michigan, USA	UM	> 30 cm tall - 12.7 cm DBH	137	1	5	L	79	0.86
Odom and Ford (2021)	New York, USA	*	*	137	*	61	L	24	0.42

Table 3.1. Summary of published sugar maple age-diameter analyses, including source, location, stand history, sampled DBH range, sampling height, the number of stands (n<sub>stands</sub>), the number of trees sampled (n<sub>samples</sub>), the form of the age-diameter relationship, estimated age of 10 cm DBH stems, and the reported R<sup>2</sup>. Stand history codes include old growth (OG), selection management (SEL), partial cuttings (PART), variable histories, including SEL and PART (VAR), and unmanaged (UM). For sampling height, S refers to stump height (unspecified). Age-diameter relationships are linear (L), quadratic (Q), or free-hand curve (C, which is not a statistical model and therefore does not have an R2 (N/A)). Tubbs (1977a) contained two snapshots of the same stand, before and after harvest treatment, and so is included twice. Harmala (2021) reported stand models separately; for brevity, we have summarized the unmanaged control from the harvest-managed stands, but additional details can be found in Harmala (2021). Information not reported is marked with a \*, and information not applicable is marked N/A.

The age structure of sugar maple under successful STS management in the Great Lakes region is expected to be influenced by stand history. This includes stand history prior to STS establishment plus the development of subsequent sugar maple cohorts in response to multiple partial harvests over 60+ years of active STS management. Many current NHF in the region established, likely as even-aged stands, 100+ years ago following widespread and intense exploitative logging in the 19th and early 20th centuries (Dickmann & Leefers, 2016), sometimes followed by slash fueled fires (Whitney, 1987). Intense disturbances promoted sprouting species, such as maple, and also early successional species (Whitney, 1987), with the latter since declining in relative abundance due to short lifespans, species-targeted partial harvests, insects and disease (bronze birch borer (Agrilus anxius Gory) impacting paper birch (Betula papyrifera Marsh.), emerald ash borer (Agrilus planipennis Fairmaire) impacting white ash (Fraxinus americana L.)), and limited high intensity disturbance following initial stand establishment (Marquis, 1967; White & Mladenoff, 1994). For most state-owned NHF stands managed by the Michigan Department of Natural Resources (MDNR), partial harvests began in the 1960s with subsequent harvests every 15-20 years. Most partial harvest were STS-motivated, plus many stands in the 1970s were subject to timber stand improvement partial harvests, with similar residual basal area to STS harvests (i.e., 70-90 ft<sup>2</sup> acre<sup>-1</sup>) (Bernie Hubbard, MDNR retired, personal communication). Since the 1960s, most stands likely had 3-4 partial harvests to present.

Anticipated age structure of harvest-established cohorts is influenced by sugar maple silvics. Periodic production of large seed crops (2 – 6-year intervals; Garrett & Graber, 1995; Cleavitt & Fahey, 2017) determines initial age distribution of seedlings, which can then persist as seedlings (and eventually saplings) for decades (Marks & Gardescu, 1998; Gravel et al., 2011). However, vigor declines with age/length of low light suppression, such that younger stems

demonstrate greater growth in response to harvest than older stems of a comparable size (Donoso et al., 2000), and seedling and sapling populations decline in abundance via accumulated low light mortality slowly with age (Hett, 1971; Gravel et al., 2011). In STS stands, sugar maple can grow to 5 cm DBH in ~ 25 - 40 years (Tubbs, 1977a; Harmala, 2021). Taken as a whole, seedlings-saplings responding to partial harvesting which recruit to a new cohort of stems > 5 cm DBH could include a relatively broad range of ages (several decades) but would likely be numerically dominated by younger individuals from the seedling/sapling class (~ 30 years old).

Thus, successful STS in NHF, over the past 60+ years, would be characterized by 3 to 4 harvest-established cohorts, perhaps of relatively broad ages, plus a dominant canopy ~100 - 120 years old. Cohorts of trees recruiting into > 5 cm diameter sapling classes in the last 60 years might be expected to range from (according to Tubbs' age size relationship) 60-year-old, 15 cm DBH trees to 30-year-old, 5 cm DBH trees. Given sugar maple's silvics (i.e., as described previously), each partial harvest cohort may vary in age but would likely have a detectable density age peak. Evidence of such age cohort peaks have even been identified in unmanaged NHF, suggesting pulses of recruitment are typical in forests under both managed and unmanaged dynamics (Goldblum & Rigg, 2002). Based on the results in Tubbs (1977a), STS should increase growth rates of poletimber and sawtimber, as evidenced by reducing average age of stems across all size classes; for example, age of a 30 cm DBH stem dropped from 194 years to 104 years, following STS management intervention. Contrastingly, failed regeneration recruitment over the same timeframe would result in a dominant age cohort of stems ~100 – 120 years old occupying a wide range of diameters, from suppressed to dominant stems, with limited younger cohorts.

Several factors could contribute to persistent, long-term recruitment failure. Browsing by white-tailed deer (*Odocoileus virginianus*) could contribute to regeneration failure, with deer

populations and browsing pressure varying spatially and temporally and regeneration recruitment failure possibly reflecting these patterns. Temporally, deer populations in Michigan were only  $\sim$ 45,000 statewide in 1914, increased to a peak of  $\sim 1.5$  million in the late 1940s, declined to 0.5 million by 1972, increased to a new peak of 2.2 million in 1995, and have declined somewhat but remained relatively high ever since (MDNR, 2016). Thus, we might expect that, with the exception of several years around 1972, deer browsing pressure on young/small trees has been high since the 1940s. Studies have confirmed that high deer browsing pressure is associated with low sugar maple sapling densities (Curtis and Rushmore, 1958; Horsley et al., 2003; Kain et al., 2011; Matonis et al. 2011; White, 2012, Henry et al. 2021) suggesting that high long-term deer densities could chronically limit the recruitment of saplings beyond their reach. In addition to State-scale changes in deer populations over time, deer populations also vary spatially at both local and regional scales due to several factors, some of which vary temporally (Shi et al.2006; Beyer et al., 2010). In Michigan, there is a strong and consistent pattern of low and seasonally migrating deer populations in regions characterized by a deep extended snowpack close to Lake Superior (Beyer et al., 2010). Areas with consistently low deer populations might, all else equal, be expected to have regeneration recruitment less negatively impacted by deer.

In addition to deer browsing, several other factors may contribute to persistent sugar maple recruitment failure. The low-intensity harvests of STS may limit recruitment of seedlings to larger classes. STS results in only modest and ephemeral increases in light, with residual canopy tree crowns quickly filling in the small, single-tree canopy gaps and stalling seedlings/sapling growth (Caspersen & Saprunoff, 2005; Kern et al., 2013). Recruitment can also be hampered by light competition from shrubs (Royo and Carson, 2006; Walters et al., 2020a), sedge (Randall & Walters, 2019), and non-sugar maple sapling layers from species like *O*.

*virginiana* and *F. grandifolia* (Elenitsky et al., 2020). Finally, although not likely a factor causing failure, site quality as driven by nutrient/water availability may affect the density of sugar maple seedling/sapling populations and the density and composition of its competitors (Elenitsky et al., 2020; Henry et al., 2021). Seedling and canopy tree growth rates are also influenced by site quality, with sugar maple generally having higher growth rate on better quality sites (Kobe, 2006; Baral et al., 2016). Comparing a low vs. high quality NHF, we would expect older saplings (e.g., 5 - 10 cm DBH) and fewer identifiable younger cohorts (i.e., since STS commenced 60 years ago) on the lower quality site due to slower growth rates. Assessing associations of these factors with established sugar maple age structure is key in assessing efficacy of STS management across the range of deer browsing intensities, site qualities/forest community compositions that support NHF.

Failure to recruit young age cohorts may portend a future decline in sugar maple canopy stem quality and volume. It has already been reported that poletimber and sapling classes are understocked in Michigan NHF (Walters et al., 2020a; Walters et al., 2022); if existing poletimber and sapling classes are nearly as old as the canopy sawtimber cohort due to limited partial harvest recruitment (i.e., 100-year-old poletimber), then productivity could decline if STS harvesting continues given the preponderance of residual trees with a long history of suppression. In the longer term, with continued STS partial harvests and no recruitment from smaller classes, stocking of the overstory will diminish and ultimately managers will run out of canopy trees to harvest. Exacerbating pole, sapling and ultimately saw timber stocking shortfalls and declining productivity, sugar maple wood quality is likely to decline for sawtimber increasingly recruited from aging, suppressed poletimber and sapling classes (Donoso et al., 2000; Baral et al., 2016; Dey et al., 2017). This would have serious implications for wood

product productivity as well as ecosystem functioning and stability. Since stands are estimated to be approaching or surpassing 100 years since stand establishment, and therefore economic maturity for the original age cohort (Dey et al., 2017), the question of success of STS in recruiting younger age cohorts is particularly timely.

To address this considerable knowledge gap in our understanding of long-term STS efficacy over the past ~ 60 years of management, we analyzed sugar maple age structure for 51 NHF distributed over a broad area varying in site characteristics, deer use history, and other factors. Given knowledge of management history of forests and deer and current structure of NHF, we predict that single tree selection has failed to recruit new, vigorous age cohorts over the past 60+ years of management. Specifically, we examine the following hypotheses:

- 1) Age is poorly predicted by typical proxies (e.g., size measurements)
  - a. Age and diameter will be weakly or non-linearly related, with a wide range of tree diameters having similar ages.
  - b. Age will be weakly related to crown class (i.e., suppressed to dominant), live crown ratio, and tree height.
- 2) Stands will have a single dominant canopy cohort ~ 100 120 years old coinciding with the intense exploitative harvests that occurred in the early 20<sup>th</sup> Century. Except for a possible 45 50-year-old class corresponding to low deer populations in the 1970s, there will be little evidence of sapling cohorts originating since partial harvesting began in the 1960's (e.g., < 60 years old)</p>
- 3) Observed age structure will differ from theoretical expectations, as evidenced by few trees younger than the age of stand initiation (e.g., 100 years).

 Older saplings will associate with areas of historic high deer density and/or lower site quality.

We test these hypotheses by examining sugar maple age-size structure for 51 STS managed NHF stands using aged basal discs from 1499 recently harvested sugar maple trees > 5 cm DBH.

# 3.3 Methods

# 3.3.1 Sites and study area

To characterize sugar maple age structure, we aged basal disc samples collected from 51 managed NHF stands distributed throughout northern Michigan. Stands were generally on upland mesic to wet mesic sites, with fertile medium-textured upland soils (Dickmann & Leefers, 2016) and dominated by sugar maple on most stands (for total basal area stems > 5 cm, sugar maple most abundant on 48 sites; red maple (Acer rubrum) 2; and balsam fir 1). Stands were considered ready for partial harvest by single-tree selection criteria ( $\sim > 23 \text{ m}^2 \text{ ha}^{-1} \text{ BA}$  and well stocked in sawtimber classes (Arbogast, 1957). These stands were part of a larger project of 141 stands assessing regeneration outcomes to silvicultural treatments, including four harvesting methods (Walters et al., 2020b) implemented Fall 2017 through Winter 2017/18. This study includes stands harvested by two of the four harvest systems that provide concentrated areas of stumps for sampling: large group selection/patch-cut (numerous dispersed 0.1 - 0.4 ha harvest openings) and seed tree (15 - 20 tree ha<sup>-1</sup> retained; Walters et al. 2020b). From a pool of 71 seed tree and large group selection/patch-cut harvested stands, we stratified sampling by three geographic areas (Western Upper Peninsula, Eastern Upper Peninsula, Northern Lower Peninsula, n=18 per region) and three ordinal site qualities (high, medium, and low, categorized by a habitat classification system based on understory flora (Burger & Kotar, 2003)) to select the 54 stands for this study. Three stands were dropped due to harvesting or logistical challenges,

resulting in a total of 51 stands. Due to dropped stands and variable representation of stands by site qualities across regions, representation of stands by strata were imbalanced (Appendix D, Table D.1).

## *3.3.2 Field methods*

Prior to harvest, we established plots in summer 2017 to characterize stand structure and composition; we also randomly selected sugar maple trees for age analysis to measure preharvest individual tree characteristics, such as crown class (i.e., relative competitive status). From a permanent grid of 25 survey points, we randomly selected plots with minimal slope (< 10 degrees) and adequate sugar maple tree sample size (> 10 sugar maple trees, > 10 cm DBH, within a 25 m radius plot), randomly resampling if plots failed these criteria. For large group selection/patch-cut harvest stands, random sampling was restricted to the two largest harvest opening sizes (centered on 10 of the 25 survey points). To characterize diameter distribution and composition of canopy stems directly competing with our sampled trees, we recorded DBH, species, and alive/dead status for all trees > 10 cm DBH within a 25-m radius (0.196 ha) plot (referred to as plot-level data). To characterize the broader diameter distribution and composition of the stand, we utilized a larger dataset of stem DBH, species, and alive/dead status (25 plots dispersed throughout the 12.14 ha stand, each 113 m<sup>2</sup>, totaling 0.283 ha per stand; hereafter referred to as stand-level data).

Since we wanted to sample age equally amongst size classes, we used stratified sampling within three DBH sampling bins of equal width, ranging from 10 cm DBH to the plot-specific second-largest sugar maple DBH, to randomly select 21 sugar maple stems. We also randomly selected up to 7 sugar maple stems 5 - 10 cm DBH to ensure saplings were well represented in the data. We maximized spatial dispersion of sampled trees by selecting the closest stem in each

of the four bins at 7 spatially dispersed sub-sampling positions within 20 m of plot center. We attempted to restrict tree selection to within 20 m of plot center to minimize competitive influence of trees just beyond the 25 m radius plot which are not quantified in our diameter distribution data. For each selected stem, we recorded DBH, crown class (dominant, co-dominant, intermediate, overtopped, adapted from Nyland, 2002), and live crown ratio (Schomaker et al., 2007), plus height for intermediate and overtopped trees, and labeled each with an aluminum tag near the base. For the three bins > 10 cm DBH, if fewer than 7 trees were located, samples were re-allocated to other size classes to attempt to always tag 21 trees.

In summer 2018, following harvest, we cut cross-sectional discs from tagged stumps. Many tags placed in summer 2017, particularly for smaller stems, were missing following harvest, most likely due to logging equipment impacts. Therefore, we additionally sampled up to 12 untagged stumps, as needed, equally allocated among the three size bins > 10 cm DBH as previously described, excluding stumps with obvious missing piths due to decay. Up to 5 untagged saplings 5 - 10 cm DBH were also sampled, often by searching a larger area including areas outside, but close, to plots if necessary. We generally collected entire basal discs, but in some cases were only able to recover a portion (but still including pith and most recent rings) due to partially shattered stumps, stumps cut low to the ground, and large stumps. To capture general variance in sampling height, we recorded an estimated stand-level average height for stumps. We collected a total of 1521 sugar maple basal discs.

#### *3.3.3 Lab methods*

We dried and sanded basal discs with progressively finer sandpaper (up to 400 grit) to reveal annual rings. We counted rings on two radii per sample, with the first radius typically the longest radius on the sample, which had most easily identifiable (widest) rings and was least

likely to have missing rings (Lorimer et al., 1999). If the ages of the two radii differed by more than 5 years, we counted a third radius to help ensure we identified the maximum number of rings present. We excluded samples with more than 2 cm radius of pith missing (n = 22, mostly due to rot), as pith ring widths vary widely and may not be estimated reliably (personal observation). Our final sample pool was 1499 discs, an average of 29 samples per stand (range 19 - 38). We were unable to locate any saplings (5 – 10 cm DBH) in three stands, including one stand in which we were unable to locate any stems < 20 cm.

## 3.3.4 Data analysis

## 3.3.4.1 Estimating age and diameter at breast height

Sugar maple is more likely to have missing versus false rings (Lorimer et al., 1999), including for defoliation events, which trigger reduced growth rates as opposed to false rings (Bevilacqua et al., 2021). Therefore, we estimated tree age as the maximum number of rings counted from a single radius. For basal discs missing less than 2 cm radius to the pith (n = 63), we estimated the number of missing rings by estimating distance to the missing center and dividing that distance by the average ring width of the inner-most 20 rings (Duncan, 1989). To ensure that plot-level stump height was not influencing our age, we modeled age as a function of stump height; there was no significant effect (p = 0.61), so we took no further action to correct for variation in stump height.

DBH is a universally used field measurement by ecologists and foresters, and which we used to characterize pre-harvest structure data; we therefore wanted to use DBH rather than disc sample diameter for modeling and age-diameter analyses. Since about half (739 of the 1499) of our basal discs were from untagged trees and lacked pre-harvest DBH measurements, we assessed whether basal disc diameter accurately estimated DBH. We measured up to 6 radii

inside the bark, evenly spaced at approximately  $60^{\circ}$  intervals, for all basal discs; fewer radii were measured for incomplete or damaged basal discs. For basal discs from tagged trees, we modeled DBH as a function of average basal disc diameter using the lm function in R (n = 760). Since average basal disc diameter was a strong estimator of DBH (R<sup>2</sup>= 0.85, y<sub>DBH</sub> = 1.35 + 0.86 \* x<sub>diam</sub>, where y<sub>DBH</sub> is sample DBH in centimeters and x<sub>diam</sub> is basal disc diameter in centimeters), we calculated estimated DBH based on basal disc diameter for all samples in place of incomplete field DBH measurements, and heretofore refer to this as sample diameter or DBH interchangeably.

## 3.3.4.2 Age associations with tree size characteristics (Hypothesis 1)

To quantify stand-level age-diameter relationships, we compared a linear regression model (fit with lm function in R) and a non-linear Chapman Richards growth function model (Richards, 1959; D. G. Chapman, 1961). To fit the Chapman Richards model, we used nlsLM from the minpack.lm package (Elzhov et al., 2016), which incorporates the Levenberg-Marquardt fitting algorithm (Levenberg, 1944; Marquardt, 1963), and generated model starting values with nls\_table from the forestmangr package (Braga et al., 2021). We used Akaike's Information Criterion (AIC) (Sakamoto et al., 1986) to compare models. We additionally considered quadratic, exponential, and segmented linear regression, but these models provided biologically implausible fits (quadratic or exponential), were highly influenced by outlier or leverage points (segmented linear regression), or fit poorly (exponential). We visually assessed preliminary models for outliers and high leverage values and removed five samples for final model fitting. To characterize variation explained by chosen models for each stand, we calculated Efron's pseudo R<sup>2</sup> values using the accuracy function from the rcompanion package (Mangiafico, 2021). In addition to individual stand modeling, we pooled samples from all stands

to develop an age-diameter relationship for the entire study region, using the same methodology as for individual stands.

For all other tree-specific factors assumed predictive of age, we ran three separate mixed effects models of age using a normal distribution and including stand as a random effect. We used lmer from the lme4 package (Bates et al., 2015). Predictors of the three models were: canopy (4 levels); live crown ratio (expressed as a percentage) and DBH, with an interaction effect to account for size-related trends in live crown ratio; and height. The first two models, respectively, were conducted for the full set of tagged and recovered samples (n = 760), while the third model included only overtopped and intermediate tagged and recovered stems with available height data (n = 300). We calculated marginal  $R^2$  values (e.g., the variance explained only by the fixed effects) for all models using r.squaredGLMM from the MuMIn package (Barton, 2020).

To summarize age structure in a format directly applicable to management, we summarized median stand age of sampled discs in each of Arbogast's timber size classes (subset of Arbogast saplings which we sampled, 5 - 11.4 cm DBH, plus poletimber (11.4 - 24.13 cm DBH) and sawtimber (> 24.13 cm DBH)) for each stand.

# 3.3.4.3 Age cohort analysis (Hypothesis 2)

To characterize age cohort density, we generated expected stand-level density distributions (stems ha<sup>-1</sup>) for sugar maple age classes. We used stand-level, rather than plot-level, data to characterize expected age density distribution due to greater area surveyed in stand-level data and to make the results applicable at a stand-scale. We assigned each basal disc a representative stem density based on stand-level diameter distribution data, using 5 cm diameter bin widths to crosswalk basal disc samples with sugar maple stand diameter distribution. For
example, in a given stand, if the stand-level density of sugar maple trees 10 - 15 cm DBH was 20 stems ha<sup>-1</sup>, and we sampled 2 basal discs 10 - 15 cm DBH, each basal disc is assigned a representative density of 10 stems ha<sup>-1</sup>. Given sparsity and skew of large diameter trees, we defined diameter groups in 5 cm intervals from 5 cm DBH to the diameter of the second largest sugar maple basal disc sample, rounded down to the nearest 5, and made the final diameter bin open-ended to include all large stems (for both basal discs and plot-level diameter structure). We occasionally combined adjacent diameter bins when we lacked basal discs in a given diameter class. Expected age densities were then aggregated by age bins. This method may slightly underestimate variability of ages present in the stand since it relies entirely on the age of collected discs, particularly in cases where diameter classes were under-sampled (e.g., smaller stems more likely to be fractured by logging equipment).

To identify potential age cohorts, we characterized peaks in the age-class density distribution of our samples. We generated five histogram iterations by shifting bin centers by 1-year increments using 5-year age bin widths, since histograms are sensitive to both bin width and position (Pond & Froese, 2015). Repeated for each histogram iteration, an age-class bin had to surpass a threshold density to count as a peak. Our threshold density, T, was defined as  $T = \frac{D}{n}$ , where D is the total sugar maple stem density and n is the number of age bins (from minimum to maximum age bin, including unoccupied age bins). We developed T via an iterative process, testing additional thresholds of 10% of total sugar maple basal area and average bin density excluding empty bins. T successfully characterized peaks in our data without being overly sensitive, and peaks were generally robust to varying threshold criteria.

To determine the width of an age cohort, we considered adjacent bins greater than T, plus any adjacent bins greater than one half the value of the threshold  $(\frac{1}{2}T)$ , as a single age cohort.

Including lower density bins  $(\frac{1}{2}T)$  to define cohort width accounts for the anticipated diffuse nature of age cohorts due to competition during stand establishment and due to variability in aging (e.g., missing rings), sample size, and stump height. Histogram iterations were generally robust in identifying similar peaks and age cohorts. Therefore, for each individual age cohort, we defined its placement and width by selecting the histogram iteration which captured the peak in the least number of age bins, randomly selecting when iterations tied for narrowest width; in doing so, we maximized the number of age cohorts which could be defined with our criteria.

After age cohorts were defined for each stand, we visually identified one age cohort which captured most co/dominant stems (typically > 20 cm DBH) which we hereafter refer to as dominant canopy age. We calculated Moran's I to assess landscape spatial autocorrelation in dominant canopy age among stands (Moran, 1950).

# 3.3.4.4 Comparison with theoretical age structure (Hypothesis 3)

To develop expected stocking by age class (as described in 2.4.4), we combined density stocking guidelines outlined in Arbogast (1957), a widely employed stocking guidance manual, with the sugar maple age-diameter relationship quantified in Tubbs (1977a), one of the seminal works establishing age-diameter assumptions for sugar maple in STS managed NHF. Using the equation in Tubbs (1977a), we estimated age for stems ranging from 5 - 61 cm DBH (DBH range of Arbogast guide which we sampled) at 0.25 cm intervals, equivalent to sampling ten stems within each Arbogast 1-inch size class. We assigned each stem an expected density using Arbogast's estimated stem density per hectare for 1-inch size classes, assigning each stem as 1/10 of the total stem density of that class; note, we did not fully sample the smallest Arbogast size class (1.5 - 2.5 inches DBH), thus we are excluding stems < 5 cm from our Arbogast theoretical expectations. We then aggregated stems ha<sup>-1</sup> for age bins to match our data

categorizations. Although Arbogast's guidelines are intended for all stems and not exclusively sugar maple, past management emphasis for sugar maple and dominance of sugar maple in current stands, plus the fact that Arbogast's guidelines are considered minimum residual stem densities targets following harvest (and we are using pre-harvest data), makes this a reasonable approximation of stocking densities and an accurate description of expected age distribution shape.

## 3.3.4.5 Stand and landscape factors associated stand age structure (Hypothesis 4)

We wanted to explore how stem age for key diameter classes varied by stand and landscape-level predictors. To maximize comparability with other studies and minimize sampling bias, we extracted model predicted age values for stems which are 10 cm, 20 cm, and 30 cm DBH (referred to as AGE10, AGE20, and AGE30), per stand. These three DBHs represent key diameter classes: large, recruited saplings (AGE10), poletimber (AGE20), and sawtimber (AGE 30), and were generally within the range of diameters sampled per stand.

We modeled AGE10, AGE20, and AGE30 as a function of two long-term climate variables (PRISM 30-year normal estimates (1981 – 2010) of annual mean temperature and January precipitation (snow); PRISM Climate Group, Oregon State University, http://prism.oregonstate.edu, created 23 September 2021) and two current stand metrics: plotlevel tree density (total basal area of stems > 10cm) and site quality (ordinal variable, ranked 1 – 4 with 1 lowest and 4 highest, Burger and Kotar 2004; Table 3.2). Since the Great Lakes generate long-term (e.g., centuries) predictable regions of deeper snow (i.e., lake effect snow) than surrounding areas and because winter temperature and snow are well-established deer population drivers (Moen, 1976; Holter et al., 1975; Shi et al., 2006), we use normalized climatic measures as a proxy for historic deer population distribution. Current stand-level characteristics

reflect the culmination of management and other disturbances/effects and growth over the last several decades. Site quality indicates relative differences in growing conditions over much longer time scales, driven largely by soil textural differences among post-glacial landforms (Zak et al., 1989; Baribault & Kobe, 2011).

Variable	Mean (Range) / n
Annual mean temperature	5.6 (4.6 – 7.6) °C
January precipitation	51.8 (29.4 – 74.5) mm
Total basal area of stems > 10 cm	26.8 (14.8 – 54.0) m <sup>2</sup> ha <sup>-1</sup>
Site quality 1	n = 15
Site quality 2	n = 9
Site quality 3	n = 17
Site quality 4	n = 10

Table 3.2. For predictors of AGE10, AGE20, and AGE30, mean and range of for continuous variables and number for categorical site quality. Site quality 1 is lowest and site quality 4 is highest.

We calculated Moran's I (Moran, 1950) to assess for spatial autocorrelation; only AGE10 was spatially autocorrelated (observed = 0.092, p-value = 0.029). We implemented regression modeling using a normal distribution with the spLM function from the spBayes package, which uses a Bayesian framework and can incorporate a spatial term if needed (Finley et al., 2015; Finley & Banerjee, 2020). We standardized our continuous predictor variables and ran models for 25,000 iterations. To check model convergence, we assessed chains for evidence of mixing. We calculated Moran's I on residuals for the AGE10 model; the model was able to account for all spatial autocorrelation, so we did not include a spatial term in our AGE10 model.

# 3.4 Results

#### 3.4.1 Data overview

Local growing conditions within plots varied in basal area, sugar maple dominance, and percentage of standing dead trees (Table 3.2). Sugar maple density ranged from 102 to 652 stems  $ha^{-1}$  with relative abundance ranging from 16 – 100% (average 72%; Table 3.3). We tallied 20 total tree species across the 51 plots/stands (Appendix D, Table D.2). Based on stand-level data, diameter distribution was generally understocked (Arbogast 1957) in stems 5 – 10 cm DBH and stocked to overstocked in stems 10 - 45 cm DBH (Figure 3.1).



Figure 3.1. Stand-level diameter distribution of sugar maple, conifer, undesirable, and other desirable species in 5 cm diameter size classes (size classes are labeled with median diameter value) from 51 northern hardwood forests stands sampled in northern Michigan, USA. Undesirable includes American beech and ironwood, while other desirable refers to all species other than sugar maple and undesirable species (sensu Walters et al., 2022 but excluding ash as undesirable species). Black asterisks represent theoretical diameter class stem densities at full stocking, for all species > 5 cm DBH, according to Arbogast (1957).

Variable	Mean (range)
Total stem density	420 (163 - 917) stems ha <sup>-1</sup>
Total BA	27 (15 – 54) m <sup>2</sup> ha <sup>-1</sup>
Sugar maple stem density	295 (102 – 652) stems $ha^{-1}$
Sugar maple BA	$18 (3 - 54) \text{ m}^2 \text{ ha}^{-1}$
Sugar maple relative abundance, by stem count	72 (16-100) %
Sugar maple relative abundance, by BA	68 (14 – 100) %

Table 3.3. Plot-level diameter composition of all living stems > 10 cm DBH for 51 plots, each representative of stand, in which age discs were sampled.

Ages ranged from 23 to 255 years, with a median and mean of 96 years. The youngest stem sampled in each stand averaged 52 years (range 23 to 118 years). The smallest stem sampled in each stand averaged 7.4 cm DBH (median 6.5, range from 5.5 - 28.4; saplings were present in our stand structure data but not located for age sampling in three stands, including one stand in which no stems < 20 cm DBH were sampled).

## 3.4.2 Associations between age and size characteristics (Hypothesis 1)

For data pooled across all stands, the Chapman Richards function was selected over a linear model (AIC 13461 vs. 13503, respectively) for fitting age vs. diameter, with slope decreasing as diameter increased (e.g., a wide range of larger diameter stems with similar ages) (Figure 3.2). For individual stands, AIC scores supported Chapman Richards model fits for 36 stands and linear model fits for 15 stands (Figure 3.3; Appendix D, Table D.3). For the 36 stands modeled by the Chapman-Richards equation, age plateaued at 111 years, on average (range from 93 - 157 years) (Appendix D, Table D.4). For the 15 stands modeled by a linear equation, two slopes were not significantly different from zero (p > 0.05), and significant slopes averaged 2.1 (range from 0.4 to 4.3); all but one intercept were significantly different from zero, and significant intercepts averaged 58 years (range from 31 – 118 years) (Appendix D, Table D.5).

Pseudo r-squared values varied widely for all models, ranging from 0.00 to 0.92 with a median of 0.63 and mean 0.58 (Appendix D, Table D.3).



Figure 3.2. Age versus diameter (cm) for 1499 sugar maple stems  $\geq 5$  cm in managed northern hardwood forests (black points). The Chapman Richards model fit to our data is shown in orange. The other lines represent relationships between age and diameter for sugar maple stems from other studies (for the range of diameter values in each study). Details on studies can be found in Table 3.1.



Figure 3.3. Model-predicted values of sugar maple age by diameter (for the stand-specific range of diameters) for 51 managed northern hardwood forests. Black lines are linear models and red lines are Chapman-Richards models.

For all data (i.e., stands pooled), age varied among crown classes, with means of 122, 105, 94, and 80 years for stems in dominant, codominant, intermediate, and overtopped (i.e., suppressed) positions respectively (marginal  $R^2 = 0.22$ ; Appendix D, Table D.6). There was considerable variation in age within each crown class except the overtopped class (Appendix D, Figure D.1). Tree diameter and live crown ratio interacted to predict stem age. For stems with smaller diameters, age tended to decrease with increasing crown percentage, and for stems with larger diameters, age tended to increase with increasing live crown ratio; however, the model (age as a function of diameter, live crown ratio, and their interaction) was relatively weak (marginal  $R^2 = 0.32$ ; Appendix D, Table D.7). Age also positively associated with height (estimated model, A ~ 55.7 + 2.2\*H<sub>m</sub>, where A is age, H<sub>m</sub> is height (m)), although weakly (margin  $R^2 = 0.22$ ; Appendix D, Table D.8).

Among stands, median age of sampled stems in Arbogast's three size classes (saplings, poletimber, and sawtimber) varied (Figure 3.4). The average median value of sapling discs was 69 years, with most stand median values between 50 - 100 years but eight stands (of 49 with sapling age samples) had median sapling ages < 50 years. Stand-level median age of poletimber



appears bimodal, with peaks around ~85 and ~100 years (overall mean 90 years). Sawtimber had a median age of 102 years (mean 106) and ranged from 72 to 166 years.

Figure 3.4. Histogram (5-year bins) and boxplot of median age for sampled discs in Arbogast's sapling (5 - 11.4 cm DBH), poletimber (11.4 - 24.1 cm DBH), and sawtimber classes (> 24.1 cm DBH), by stand, for 51 STS managed northern hardwood forests. Dashed lines emphasize 50, 100, and 150 years on the x axis. Boxplots show first, second (median), and third quartiles, plus whiskers which extend up to approximately three standard deviations from the mean (1.5 times the inter-quartile range); points represent outliers. Note, we did not sample the full width of Arbogast's sapling class (stems 3.8 - 11.4 cm DBH).

## 3.4.3 Age cohort results (Hypothesis 2)

Age distributions by stand were variable but showed some general trends (Figure 3.5).

Most (59%) stands had three or fewer age cohorts (8 stands had one cohort; 14 had two; 8 had

three). The median age of the dominant canopy cohort averaged 101 years, ranging from 53 to

163 (Figure 3.5C); the width of the dominant canopy cohort (e.g., number of age bins it spanned)

ranged from 5 to 35 years (median 15 years, or three 5-year age bins). There was no statistically

detectable spatial autocorrelation in the dominant canopy age (Moran's I, p value = 0.15, Figure 3.5C). For 12 of 14 stands with two cohorts, there was a cohort ~ 20 - 30 years younger than the main canopy cohort, with all but one of these younger cohorts > 50 years old. Cohorts younger than the canopy were typically low-density, in opposition of J-shaped size-age distribution expectations which with greater stem density in small/younger cohorts. Among all stands, few cohorts (n = 17 of 170, or 10% of all identified cohorts) had a median age < 50 years old. Sixteen stands had at least one cohort older than the dominant canopy class, including five stands with age cohorts above 200 years located in the Upper Peninsula (4 of 5 along the northern half).



Figure 3.5. Histogram-identified sugar maple age cohorts by site (x-axis) for 51 managed northern hardwood stands (3.5A). Point size represents total stems per hectare measured within that cohort, and black points represent the age cohort capturing the dominant canopy. Stands are organized by number of identified cohorts within each stand and sorted (low to high) by median age of the youngest age cohort. Number of identified age cohorts (3.5B) and median age of the canopy cohort (3.5C) are represented spatially, plus a histogram of canopy ages (3.5C).

## 3.4.4 Age structure deviation from expectations (Hypothesis 3)

For all stands pooled, stand-level sugar maple stem density by age class appears normally distributed, with a modal age class of 90 - 100 years old (columns, Figure 3.6A). Compared to theoretical age distribution for successful STS management (black dots, Figure 3.6A), age classes < 60 were understocked, age classes and age classes 80 - 120 years old were stocked to overstocked, and age classes > 120 years were understocked (Figure 3.6A). Stand-level analyses

indicated greater variability in stocking by age class (Figure 3.6B). The 70 - 90 and 90 - 110year age bins were most often stocked, but 25 and 22% of stands, respectively, were not stocked in these age classes, though they typically had a surplus of stems in an adjacent age class (Figure 3.6B). No stands were stocked in the 30 - 50-year age bin (Figure 3.6B).



Figure 3.6. Bar plot of average age density distribution (20-year age bins, labeled with median value) of sugar maple stems > 5 cm DBH, pooled across 51 managed NHF stands; black points represent age density of a fully stocked northern hardwood stand, following harvest, if entirely stocked with sugar maple (Figure 3.6A). Anticipated age distribution was determined by applying the age-diameter relationship from Tubbs (1977a) to the size class densities outlined by Arbogast (1957). Bar plot representing the percentage of stands individually considered stocked (e.g., exceed Arbogast's extrapolated stocking value) for each age class expected under the anticipated age distribution (Figure 3.6B).

# 3.4.5 Factors associated with stand age structure (Hypothesis 4)

Based on stand level age-diameter models, predicted ages of stems at 10, 20, and 30 cm DBH (AGE10, AGE20, and AGE30) averaged 72, 94, and 106 years, respectively, with ranges of ~50+ years (Figure 3.7). AGE10 stem age was greater in areas with higher mean temperatures and lower January precipitation, i.e., regions with milder winters, where we expect long-term deer browsing pressure to be higher (Table 3.4). AGE20 stem age also increased with lower January precipitation, and while AGE30 was unrelated to climate variables, stems on medium quality sites were predicted to be older than stems on low quality sites (Table 3.3).



Figure 3.7. Kernel density estimates of model-estimated values of stems 10 cm (purple solid line), 20 cm (green dashed line), and 30 cm (yellow dotted line) DBH are presented (n = 48 stands for AGE10, n = 50 for AGE20, and n = 51 for AGE30); these are interpretable as a smoothed version of a histogram, where the area under the curve of each line sums to one. We did not extract model estimates beyond the range of data, therefore estimates were not available for three stands lacking saplings, one of which additionally had no stems below 20 cm DBH.

Size	Variable	E 50%	E 2.5%	E 97.5%
	Intercept	71.5	64.9	77.7
	Mean annual temperature	6.4	2.6	10.4
	January precipitation	-5.5	-9.4	-1.4
Size AGE10 AGE20 AGE30	Plot basal area	0.7	-3.1	4.6
	Site quality 2	0.8	-9.3	11
	Site quality 3	1.6	-7.2	11.2
	Site quality 4	0.1	-10	10.4
	<b>T</b> <sup>2</sup>	145	97.3	232
AGE20	Intercept	93.5	87.4	99.5
	Mean annual temperature	3.1	-0.5	6.9
	January precipitation	-3.9	-7.7	-0.3
	Plot basal area	0.9	-2.8	4.3
	Site quality 2	5.3	-4.6	15.1
	Site quality 3	-0.3	-9.1	8.4
	Site quality 4	-1.8	-11.3	8.3
	$T^2$	131	87.7	207
	Intercept	102.7	95.5	109.7
	Mean annual temperature	2.2	-1.8	6.4
	January precipitation	-0.1	-4.1	4.1
ACE20	Plot basal area	-1.2	-5.1	2.6
AGESU	Site quality 2	14.9	3.6	26.3
	Site quality 3	2.8	-7.2	13.2
	Site quality 4	-2.4	-14	9
	$\mathbf{T}^2$	169.2	114.8	264.4

Table 3.4. Model parameter estimates (E 50%) and 95% credible intervals (E 2.5%, E 97.5%) for factors predicting age of sugar maple stems at 10, 20, and 30 cm DBH. Significant predictors (p < 0.05) are bolded and italicized. Site quality 1 (lowest site quality) is included in the intercept, and site quality 4 is the highest site quality.  $T^2$  represents the variance (equal to the squared standard deviation).

# 3.5 Discussion

#### 3.5.1 Overview

Age-diameter relationships from 51 STS managed NHF in northern Michigan support our overall hypothesis that STS has largely failed to recruit new regeneration cohorts (> 5 cm DBH) over the past ~ 60 years of management. Successful uneven-aged management is expected to generate linear tree diameter vs. age relationships (such as in Tubbs, 1977a), but our results suggest stems of similar age can occupy a wide range of pole and sawtimber diameter classes. Sapling classes are generally understocked by STS management standards (Arbogast 1957) and old (5 – 10 cm DBH stems average 66 years old). Overall, few sugar maple trees recruited (grown into > 5 cm DBH classes) since STS became widespread, indicating extensive regeneration and recruitment failure under STS management for sugar maple.

Our results support Hypothesis 1, that age is poorly estimated by diameter and growth form characteristics (crown class, height of understory trees, and live crown ratio); small diameter stems with small crowns in subordinate crown classes were as, or nearly as old, as dominant, large diameter, and large crowned canopy stems. Diameter – fundamentally the most common age proxy – often displayed a non-linear relationship with age, and for the 15 stands best fit by a linear model, slopes were often very small (and two were not significantly different from zero; Appendix D, Table D.2). Variation among relationships indicates that age-diameter relationships need to be quantified at sub-regional (if not stand-level) scales if they are to be used as an accurate, useful tool for estimating age from diameter.

Our second hypothesis that stands are dominated by a canopy cohort of sugar maple stems, approximately 100 - 120 years old, was supported by the data (Figure 3.5A). Interestingly, our results suggest sapling recruitment of current dominant canopy trees (under

mostly open, highly disturbed conditions) was typically gradual (e.g., average width 15 years). Extended canopy establishment may be explained by competition from both non-tree vegetation (e.g., *Rubus* shrubs, Hughes & Fahey, 1991; Walters et al., 2016) and faster growing early successional trees (aspen, birch; Marquis, 1967) delaying/extending establishment and growth of sugar maple. A mixture of stump sprouting and seed-origin establishment, which vary in vertical growth rate and establishment time (Whitney, 1987), could also contribute to this phenomenon.

Canopy age was not spatially autocorrelated across our study region, which suggests local factors drive canopy establishment rather than regionally varying factors. Despite lack of statistical evidence, visual inspection of mapped canopy age indicates most stands with older canopy ages (~ 125 years old) are located along the Great Lakes and near areas with long settlement history (e.g., Traverse City, Houghton), while younger canopies tend to be found in more remote locations (Figure 3.5C). This may suggest that older forests are associated with areas with greater accessibility to mills early in the exploitation era and/or areas where earlier conversion to agriculture followed by abandonment occurred (Dickmann & Leefers, 2016).

The characteristics of additional age cohorts for stands with more than one age cohort further support Hypothesis 2; there was little evidence of young age cohorts establishing since partial harvesting began under STS in the 1960s. Although some stands had 4+ age cohorts, most were older (often older than main canopy cohort). The common pattern of a cohort 20 - 30 years younger than the main canopy cohort defies easy explanation. This secondary cohort may have filled the growing space initially filled by a short-lived early successional species that subsequently died (e.g., pin cherry; Whitney, 1987). They may also reflect the first harvest of a stand that generated a new age cohort; however, they mostly precede, in time, anecdotal evidence of partial harvests (1960s), nor is there any conceivable justification for partially

harvesting a 30-year-old stand. Additional stand history investigation, possibly combining detailed records, if they existed, with age distribution data, may yield greater insight.

We were further surprised to identify several stands with sugar maple age cohorts > 200 years old, generally in the northern Upper Peninsula. That those stands were not entirely clearcut during the era of exploitive intense harvests (late 1800s - 1920s) is not surprising given that exploitative harvests of that era may have often resulted in non-commercial (small, poor form) stems being left behind. Alternatively, some stands likely had old-growth structure with no harvesting up to the time that STS was implemented in the 1960s; three of our stands which had 5+ age cohorts of similar size including age cohorts > 200 years are potential candidates for this category (Figure 3.5A).

The widespread lack of sugar maple cohorts < 50 years old coincides temporally with increasing deer densities beginning in the 1970's, related to mill expansion and the creation of abundant browse via widespread harvesting of aspen (MDNR 2016). However, the lack of cohorts 50 - 60 years old defies expectations, as establishment of these stems in the 1960's would have coincided with relatively low deer populations (MDNR, 2016) and the beginning of STS partial harvesting (Bernie Hubbard, MDNR retired, personal communication). The lack of saplings (e.g., > 5 cm DBH) aged 50 - 60 years suggests STS widely failed to create well stocked sapling recruit classes throughout northern Michigan despite lower deer populations. While deer browsing likely still contributed to low sapling recruitment other factors may have also contributed, including small STS gaps creating insufficient light environments for sugar maple growth and survival via maintenance of high basal area/low light (Henry et al., 2021). In addition, STS commenced when stands were young (40 - 50 years) with likely small diameters, such that sugar maple seed (the source of future sapling recruits) could have been limiting in

many stands (Henry et al., 2021). Overall, our results point to a surplus of aging sugar maple canopies and a lack of younger recruited age cohorts, indicative of persistent recruitment failure.

Lastly, we found support for our hypothesis that high deer use would be associated with older saplings but not for high site quality associating with younger saplings (Hypothesis 4). AGE10 was spatially autocorrelated, with regions of younger sapling ages along the northern half of the Upper Peninsula where snow is deep in the winter (Appendix D, Figure D.2). Related, AGE10 was older in regions with warmer winters and less snow, with both factors known to be associated, in parts of the study region, with higher winter deer populations in the region (Holter et al., 1975; Moen, 1976; Shi et al., 2006). Given the general lack of AGE vs. site quality relationships, we found no support for our hypothesis that presumably higher growth rates on higher quality sites would lead to younger stems at given diameter (e.g., lower AGE10). This could be due to lack of a site quality vs. growth relationship or other factors (deer browsing, stand density) may have overwhelmed or confounded site effects. As an example of potentially confounding effects, basal area and shading of subordinate tree strata could be greater on higher quality sites and diminish growth. Regardless of factors driving pattern, overall, the preponderance of old, low density sapling classes in managed NHF is concerning and emphasizes the importance of understanding age as well as diameter distributions for stocking if employing STS.

## 3.5.2 Comparison to other studies

Our size-age relationships differ from previous studies of sugar maple in managed forests (Table 3.1) and appear more similar to those for old-growth/unmanaged forests (Tubbs, 1977a; Canham, 1985). Compared to previous studies conducted in the Great Lakes region, differences in our results might be explained by our study encompassing a broader region (as compared to

Harmala 2021) with more variation in deer use (as compared to studies in areas of low deer populations Dey *et al.* 2017 and Harmala 2021). Restricting our sampling to one or two stands, particularly in a region of lower deer browsing, may have resulted in similar conclusions; for example, the map of AGE10, which was spatially autocorrelated, indicates a cluster of stands along the central northern portion of the Upper Peninsula which have stems 10 cm DBH averaging 50 - 60 years old (Appendix D, Figure D.2), which is more comparable to past studies (Table 3.1).

Our summary also indicates macro-regional variation in sugar maple age-diameter relationships, with stems older at given size in Michigan (Tubbs, 1977a) than in New England, both for old-growth (e.g., Leak 1985 in New York) and managed stands (e.g., Kenefic and Nyland 1999 in New York). This may be explained by greater summer precipitation in the Northeast ( $\sim 76 - 89$  cm annual precipitation in northern Michigan vs. 102 - 152 cm in New England, PRISM 30-year normals) or by regional variation in stand composition and history, including management. However, contrary to patterns for larger trees, Kobe (1996) reports similar growth rates for sugar maple saplings in Michigan vs. Connecticut, USA. An important caveat to our comparison of studies is that sampling height can influence age estimations, since it can take sugar maple a median of 2 - 10 years to progress from 30 cm to breast height (Harmala, 2021). Odom and Ford (2021) and Kenefic and Nyland (1999) sampled at breast height; however, adding ten years to their estimates of sugar maple age (across all diameters) would still indicate stems in New England are considerably younger than in Michigan. Overall, except for saplings, evidence suggests sugar maple grows more quickly in New England than the Great Lakes region.

## 3.5.3 Management implications

Individually and in combination, stand diameter and age distributions are concerning for management by STS (Figures 3.1, 3.6). The deficit of young stems in sapling and poletimber diameter classes suggests persistent recruitment failure, which has several ecological implications. Older stems tend to acquire greater damage and release with less vigor (e.g., slower basal area increment) in response to harvest (Baral et al., 2016), such that stands dominated by old stems could have lower growth rates than stands dominated by young stems. Lack of vigorous young stems may also mean that sugar maple is less likely to capture gaps following canopy disturbance or harvest. If this provides opportunity for species diverse stems to recruit, this outcome could be beneficial for stand resilience; however, it could also lead to gap capture by highly abundant, non-viable (American beech due to beech bark disease) or non-commercial (ironwood) species (Elenitsky et al., 2020; Walters et al., 2022), which would lead to canopy failure (e.g., loss of continuous canopy).

From a timber supply standpoint, our results point to pressing challenges in both near (e.g., one to two decades) and intermediate (e.g., four to five decades) time frames. Poletimber stems averaging ~ ten years younger than sawtimber indicates an imminent impact of unbalanced age structure. Given that sugar maple stems tend to decline in quality after 100 years due to decreasing proportions of high-grade white wood lumber (Dey et al., 2017), trees harvested as larger sawlogs (e.g., 40 cm DBH +) in the future will have lower rates of return than those harvested at 40 cm DBH now (i.e., at approximately 100 years old). Stands with limited age distribution may benefit from more intense harvests to release poletimber to the sawtimber class and to increase light availability in the understory to promote sugar maple recruitment.

From a slightly longer-term view, stand-level median sapling ages are currently 66 years old; these stems would be expected to reach economic maturity in 40 years. Although possible, this scenario would require full release of stems, and their quality may be compromised due to extended suppression. Furthermore, saplings are currently understocked (Figure 3.1), so their release would likely result in perpetual understocking of larger size classes. Stands with most saplings dominated by old, low-vigor saplings may benefit from eliminating or reducing current sapling stock and attempting to promote high-light growth environments to release younger and presumably more vigorous advanced regeneration. From a regional (Michigan) perspective, alternative silvicultural regimes may need to be implemented imminently across broad areas to diversify sugar maple age structure (Appendix D, Figure D.3); this may be necessary to minimize disruption to economic supply of sugar maple timber products, as many of these stands will be approaching economic maturity simultaneously.

# 3.5.4 Caveats and future directions

An important caveat for our method of identifying age cohorts was that it was biased to minimize age cohort width and maximize number of age cohorts to rigorously test our hypothesis that most stands would have few age cohorts. Some identified cohorts close in age may truly represent the same recruitment event and therefore age cohort, which would further support our notion that recruitment or establishment events are likely to be diffuse in age range. Furthermore, some potential low-density, old-age (e.g., > 200 years) age cohorts are missed by our methodology by employing a single threshold across all diameter classes, although stands with older trees overall tended to have more empty diameter bins (0 stems ha<sup>-1</sup>) and therefore a lower threshold relative to total sugar maple stem density.

Future analyses, both with our samples and in other stands, can contribute significantly to our understanding of sugar maple age structure in STS managed NHF. Additional analyses with more refined measures (e.g., soil measurements for site quality) may shed additional insights on the influence of deer browsing and nutrient and water availability in determining sugar maple age structure. Further analysis with a subset of this data to characterize ring width over time would enable a look at past suppression and recruitment dynamics. This could yield insight into how recruitment patterns have changed over time, such as average age and diameter of stems at canopy recruitment. Additionally, more intensive, directed sampling of stems 5 - 10 cm as well as 0-5 cm could yield more refined insight into the age – diameter relationships in smaller diameter classes. We assumed sampled saplings represented all saplings within the 5 - 10 cm class, but additional sampling intensity may reveal greater age diversity; given the highly suppressed nature of saplings, 5 cm DBH vs. 10 cm DBH stems may vary widely. Additional visual characterizations of these small stems may also reveal traits that may serve as a better proxy for age than diameter, height, or live crown ratio, such as crown shape (Bartholomew, 2018).

#### **CHAPTER 4**

# PATTERNS AMONG SPECIES FROM SEEDLING TO CANOPY PORTEND FUTURE COMPOSITIONAL SHIFTS AND LOW RESILIENCE IN MANAGED NORTHERN HARDWOOD FORESTS

## 4.1 Abstract

Declining biodiversity is pervasive globally, challenging sustainability and resilience of ecosystems facing multiple environmental stresses. Ecologically and economically valuable northern hardwood forests of the Great Lakes region have experienced declines in canopy tree diversity since European colonization. Current tree regeneration trends (i.e., restricted to shade tolerant, deer browsing resilient/resistant species) in managed stands suggest low tree-diversity patterns among mature/canopy trees will continue and may intensify in the future; however, studies characterizing current diversity and projecting further trends at regional ( $10^6$  ha) scales are lacking. For 141 managed northern hardwood forests across northern Michigan, USA, we analyzed patterns and possible drivers of species diversity and relative abundance by species from seedlings to canopy trees. Overall, the estimated asymptotic stand-level richness for seedlings (< 137 cm tall) and saplings (> 137 cm tall and <10 cm diameter) averages 6 and 7 species, respectively, and with relatively low evenness (effective number of species, Hill  $q_{1}$ , < 3 species). Stands with lower species diversity of canopy/mature stems (> 10 cm DBH) tend to have lower seedling and sapling diversity. Models of individual species abundance indicate that high total canopy/mature basal area (i.e., shade) and low conspecific mature stem density (lack of seed sources) consistently contributed to low seedling/sapling density. Among the 17 most common species, seven have lower relative importance (RI) as seedlings than mature trees (e.g., Tsuga canadensis, Betula alleghaniensis, Populus grandidentata, Tilia americana) suggesting

seedling establishment (germination substrate and/or early survival) limitations; two (*Acer* saccharum, Quercus rubra) had lowest RI as saplings suggesting seedling to sapling recruitment bottlenecks (e.g., deer browsing); six (e.g., *Fagus grandifolia, Ostrya virginiana, Pinus strobus*) had highest RI as saplings indicating low seedling to sapling recruitment bottlenecks; two (*Prunus spp., Acer rubrum*) had highest RI as seedlings with moderate declines in sapling and canopy/mature classes indicating modest sapling and canopy/mature recruitment bottlenecks; and *Fraxinus americana* had highest RI as seedlings and lowest for canopy/mature trees, which is consistent with its recent widespread canopy mortality from emerald ash borer. Observed regeneration diversity patterns suggest future shifts in mature canopy stratum composition and highlight a need for changes in management aimed at increasing tree species diversity and desirable species mixes.

## 4.2 Introduction

Biodiversity contributes to ecosystem resilience, or the capacity of a system to recover following disturbance. Resilience via diversity will be necessary for ecosystem functioning in the face of increasingly frequent and diverse human-mediated disturbances including forest fire activity (Flannigan et al., 2000), invasive pests and pathogens (Dukes et al., 2009; Ramsfield et al., 2016), and frequency and severity of drought (Allen et al., 2010), plus interactions among factors (Dale et al., 2000) and with climate change (Dale et al., 2001). Unfortunately, for North American temperate forest ecosystems, accelerating multiple threats to ecosystem functioning comes at a time of declining tree species regeneration diversity (Ramirez et al., 2018; Miller & McGill, 2019; Vickers et al., 2007); these patterns that may portend future stasis or further declines in canopy biodiversity and resilience. Characterizing tree regeneration species diversity broadly,

geographically, will be important for forecasting future forest resilience globally under increasingly volatile environments.

Covering over 50 million hectares of northeastern United States (Oswalt et al., 2014), northern hardwood forests (NHF) are an important forest ecosystem currently dominated by ecologically and economically valuable sugar maple (*Acer saccharum* Marsh) (Linehan & Jacobson, 2005; Duval et al., 2014). Compared to pre-European colonization, current canopy tree species richness and structural complexity are lower overall (Schulte et al., 2007) and may be lower in managed than unmanaged stands (Crow et al., 2002; Neuendorff et al., 2007; Powers & Nagel, 2009). Low diversity and density in the regeneration layer of key species, such as sugar maple, are particularly concerning (Leak, 2006; Neuendorff et al., 2007; Powers and Nagel, 2009; Matonis et al., 2011). However, few studies have analyzed stand-level tree regeneration diversity at regional (10<sup>6</sup> ha) scales, which may capture driving factors that operate at that scale. Addressing challenges to securing diverse tree regeneration is becoming increasingly important, given predicted declines in ecosystem functioning in the Great Lakes region from climate change (Rogers et al., 2017) and continued invasive pest and pathogen establishment and spread (Liebhold et al., 2013; D. Chapman et al., 2017; Panzavolta et al., 2021).

Although diversity alone can contribute to resilience in the face of multiple potential disturbances, species composition also matters, and forests may be undergoing compositional shifts that change their capability to provide multiple ecosystem goods and services that are resilient to diverse disturbances (Walters et al., 2022). Many studies indicating declining tree regeneration density for key species also show, or suggest, different composition for regeneration than canopy trees (e.g., Neuendorff et al., 2007; Matonis et al., 2011; Elenitsky et al., 2020; Hupperts et al., 2020), indicating possible compositional shifts are underway. However, there are

few explicit examinations of composition in seedling, sapling and canopy classes and factors potentially underlying differences across size class. Quantifying current northern hardwood tree regeneration composition and diversity and associated driving factors is key for forecasting potential future canopy composition and diversity and for development and strategic implementation of adaptive forest management.

Several potential driving factors likely influence tree regeneration composition and diversity in managed northern hardwood forests. Abiotic drivers of nutrient and water availability, which vary spatially, influence tree species distribution, assemblages, and therefore diversity. For the Great Lakes region, post-glacial landforms, which vary in nutrient availability and soil water holding capacity (site quality), help drive forest type distribution, from xeric jack pine-black oak dominated coarse textured outwash soils, to mesic northern hardwood dominated finer textured moraine soils (Zak et al., 1986, 1989; Baribault et al., 2010). Interacting with site quality is a Great Lakes-climate driven (i.e., lake-effect) three-fold gradient of annual snowfall, further influencing forest distribution and species distributions, particularly for sugar maple (Henne et al., 2007). Over the narrower range of higher site qualities supporting northern hardwood forests, tree regeneration composition varies (Elenitsky et al., 2020), but the potential for site to influence future composition of all strata could be overridden by other factors, such as deer browsing or management (Matonis et al., 2011; Bannon et al., 2015). Compositional data supporting habitat classification development for the region (i.e., frequency of occurrence of saplings, Burger and Kotar 2003), suggest that lower site quality sites supporting NH tend to have more species; however, explicit examinations of tree diversity in relation to site quality are lacking.

Interacting with the abiotic landscape, biotic drivers contribute to patterns of tree diversity in NHF of the Great Lakes region. Invasive pests and pathogen epidemics have decreased the likelihood of canopy ascendance/maturity for American beech (Fagus grandifolia), white ash (Fraxinus americana) and American elm (Ulmus americana) (Parker & Leopold, 1983; Forrester et al., 2003; Nuckolls et al., 2009; Klooster et al., 2014). Impacting a wider range of tree species, invasive non-native earthworms alter nutrient cycling, retard decomposition, and reduce understory plant abundance (Bohlen et al., 2004; Holdsworth et al., 2007, 2008; Resner et al., 2015), potentially influencing tree regeneration composition (Hale et al., 2006). Abundant white-tailed deer (Odocoileus virginianus) populations, which have increased drastically since the early 1900s (MDNR, 2016), can shift species composition through selective browsing and depress overall regeneration density (Côté et al., 2004; Bradshaw & Waller, 2016), including canopy dominant sugar maple (Curtis and Rushmore, 1958; Horsley et al., 2003; Kain et al., 2011; Matonis et al. 2011; White, 2012, Henry et al. 2021). Deer browsing interacts with stand level variables to influence tree regeneration density and composition, including site quality (Randall & Walters, 2011) and light availability/gap size (Sage et al., 2003; Walters et al., 2016). Since deer browsing is vertically limited to a maximum of  $\sim 2$  m (Walters et al., 2020a), regeneration which has vertically escaped the deer browsing zone may be particularly indicative of future canopy structure and composition. Capturing variability in biotic drivers, such as invasive pests and pathogens or local deer populations, is key to fully characterizing trends in tree regeneration diversity and distribution, which is necessary for identifying particularly low-diversity regions in need of alternative management intervention.

In addition to biotic and abiotic drivers, past forest management may have played both direct and indirect roles in patterns of tree regeneration diversity and density. The single-tree

selection (STS) silvicultural system was introduced in the late 1950's (Arbogast 1957) and has since dominated northern hardwood forest management (Kern et al., 2014). Under this system, periodic partial harvests every 10-20 years remove select trees, resulting in small, dispersed harvest gaps. These gaps theoretically promote the regeneration of economically desirable shadetolerant species, such as sugar maple. Under this system, tree diversity may decline via both intentional removal of competing stems of less economically valuable species and decreased regeneration and canopy recruitment of less shade tolerant species in small STS gaps (Niese & Strong, 1992; Crow et al., 2002). Local (stand) scale, species-specific declines or elimination of mature/canopy tree seed sources can negatively impact regeneration rates for many NHF species (Willis et al, 2016), particularly those with shorter distance dispersal potential (McEuen & Curran, 2004). Small seeded species, such as yellow birch and eastern hemlock, may face additional seedling substrate establishment constraints via low coarse wood coverage resulting from periodic harvests capturing potential large tree mortality and from winter logging and use of lower impact harvesting equipment resulting in minimal exposure of mineral soil substrate via disruption of the forest floor (Marx & Walters, 2008; Bolton & D'Amato, 2011; Willis et al., 2016). Given widespread implementation of STS management in NHF, characterizing tree diversity regeneration outcomes for these forests is key for future management success as well as landscape-level forest health and functioning.

Despite potential failure of STS to promote diverse forests, management can play a role in improving future diversity. Climate change forest management will involve efforts to adapt, mitigate, or migrate forests (Aitken et al., 2008), and managers will need to be increasingly flexible, employing a variety of strategies to adapt to future conditions (Millar et al., 2007). For northern hardwood forests, alternative, more intense management harvesting regimes are being

explored (Hupperts et al., 2020; Walters et al., 2020; Rogers et al., 2021). These include testing the notion that larger canopy harvest gaps promote more species diverse tree regeneration by promoting admixtures of shade-intolerant tree species (e.g., Niese and Strong, 1992; Walters et al., 2016). However, some studies suggest more intense harvests may not increase regeneration diversity in the long-term (Kern et al., 2013; Knapp et al., 2021; but see Niese and Strong, 1992). Accurately characterizing current patterns and associated driving factors is necessary to guide regional implementation of future management of NHF, whether by continued use of STS or by alternative silvicultural strategies.

In this study we address knowledge gaps in understanding the patterns and potential causes of current and potential future diversity and composition of NHF. To do this we quantified patterns of individual species regeneration abundance and diversity, plus associations with likely stand and landscape drivers. We also characterized variation in species diversity and abundance as a function of size class and interpret these patterns in terms of potential drivers of pattern and implications for compositional shifts in the future. We address these goals by analyzing a dataset of 141 selection-managed northern hardwood forests, distributed throughout northern Michigan, which capture a variety of potential landscape and stand level drivers. Given current management regimes and other abiotic and biotic factors, we predict:

- Tree diversity will be low particularly on high quality sites and in all size strata, but especially low in seedling and sapling layers due to filtering effects of substrate, resource limitation, and selective deer browsing pressure.
- As driven by limiting factors listed in (1) and others, we predict compositions in seedling, sapling, and canopy strata will differ, portending possible future shifts in canopy composition.

- 2a) Canopy composition (conspecific basal area of stems > 10 cm DBH) will associate with regeneration density and diversity because of seed source limitation, particularly for large-seeded species
- 2b) Small seeded species will face seedling establishment limitations, negatively associating with hardwood litter coverage, and having greater relative representation in the overstory than in the understory
- 2c) Shade intolerant and mid-tolerant species will associate negatively with increasing basal area and will have greater representation in the canopy/mature strata versus the understory
- 2d) Highly palatable tree species, like sugar maple, will negatively associate with deer use for the sapling class and will have greater representation in the canopy or seedling layers versus the sapling layer

We interpret these results in the context of implications for future tree diversity patterns and with explicit consideration to management implications.

## 4.3 Methods

#### 4.3.1 Study area and data collection

Our data are from 141 managed northern hardwood forest stands distributed throughout northern Michigan which are part of a larger experiment analyzing harvest methods and regeneration outcomes (Walters et al. 2020b). Here, we focus on pre-harvest regeneration. These stands were considered ready or near ready for partial harvest by single-tree selection criteria (i.e.,  $> 23 \text{ m}^2 \text{ ha}^{-1}$  BA and well stocked in sawtimber classes, Arbogast, 1957) at the time of sampling and were likely subject to 3-4 partial/selections harvests prior to our sampling them. Additional vegetation sampling details can be found in Henry et al. 2021); we recount pertinent details here.

Most stands were state owned (n=119) and managed by the Michigan Department of Natural Resources (MDNR), with the remaining 22 stands owned and managed by private forest industry companies (Hancock Timber Resource Group, The Rohatyn Group). Stands were generally on upland sites, mesic to wet mesic, with fertile medium-textured upland soils (Dickmann & Leefers, 2016) and dominated by sugar maple (vast majority), red maple or basswood (see Table 4.1 for scientific names and authority). We established a systematic grid of 25 survey points within a 12-ha square within each stand and collected species-level data in the following plots at each survey point. We counted all tree seedlings shorter than 50 cm and ocularly estimated percent cover of shrubs and groundcover composition (e.g., hardwood litter, bare mineral soil, downed woody material) within 2 quadrats, each 1 m<sup>2</sup> (50 m<sup>2</sup> total per stand). Within a 2 m radius circular plot (314 m<sup>2</sup> total per stand) we tallied tree regeneration for stems 50 to 137 cm tall and tallied and measured DBH for stems > 137 cm tall and < 5 cm DBH. Within a 6 m radius circular plot (2,827 m<sup>2</sup> total per stand), we tallied and measured DBH for all stems > 5 cm DBH. Diameters were converted to basal areas (stem cross sectional areas at breast height), summed, and expressed on a m<sup>2</sup> per ha basis, with values (BA) representing the density of trees > 10 cm DBH (i.e., superior to our largest sapling class). Using herbaceous plant assemblages and key indicator species, we assigned each stand a dominant habitat type, which is an index of site quality (Burger & Kotar, 2003).

We assessed winter deer use with fecal pellet transect surveys in spring of 2017 at a subset of 50 sites, surveying 628 m of transect (6 m wide) subdivided into 103 segments at each site. We modeled estimated winter deer use (percent transect segments occupied by deer pellets)

for all stands in 2017 with a spatial model in a Bayesian framework; the model also incorporated well-established climate and landscape factors known to influence local deer populations, plus deer pellet transect surveys conducted in 2019 at 139 sites following experimental harvest (Appendix A).

Sciontific nomo		CODE	Common	Mean stems ha <sup>-1</sup> (Max stems ha <sup>-1</sup> )   n			
Scientific name		CODE	name	Seedlings	Saplings	Canopy	n
Abies balsamea [L] Mill.	**	BF	Balsam fir	817 (11,295)   47	154 (936)   64	41 (279)   62	77
Acer rubrum L.	**	RM	Red maple	17,823 (139,733)   118	641 (5,673)   71	86 (460)   84	119
Acer saccharum Marsh.	**	SM	Sugar maple	71,797 (448,752)   140	1,217 (6,943)   140	282 (531)   139	141
Betula alleghaniensis Britton	**	YB	Yellow birch	609 (3,664)   47	81 (725)   39	15 (113)   66	85
<i>Betula papyrifera</i> Marsh.	*	PB	Paper birch	350 (1,055)   9	10 (32)   7	10 (28)   20	27
Fagus grandifolia Ehrh.	**	AB	American beech	4,090 (49,171)   89	1,671 (5,903)   96	41 (195)   87	98
Fraxinus americana L.	**	WA	White ash	7,885 (55,453)   71	451 (5,188)   43	12 (67)   31	72
Fraxinus nigra Marsh.		BA	Black ash	64 (64)   1	39 (39)   1	4 (4)   5	5
<i>Fraxinus pennsylvanica</i> Marsh.		GA	Green ash	1,483 (5,600)   5	35 (64)   3	9 (18)   5	8
Ostrya virginiana Mill.	**	IW	Ironwood	2,744 (14,566)   108	955 (4,545)   111	17 (74)   77	116
Picea spp.	*	SP	Spruce species	244 (2,159)   22	25 (223)   35	9 (39)   32	3

Table 4.1. Summary statistics for tree species regeneration at the stand level when present in a given size class, including mean, maximum (stems ha<sup>-1</sup>), and number of stand occurrences (n, out of 141) by size class, plus overall number of site occurrences n (across all size classes). Summary statistics were calculated for the subset of stands which that species was present on, in a given size class. Seedlings are 0 - 137 cm tall, small saplings are > 137 cm tall and < 10 cm DBH, and canopy trees are > 10 cm DBH. Species labeled with \*\* were modeled for individual abundance and relative abundance by size class; species labeled with \* were modeled only for relative abundance by size class. Species without any \* were included only for general species diversity calculations and stand summaries.

Scientific nome		CODE	Common name	Mean stems ha <sup>-1</sup> (Max stems ha <sup>-1</sup> )   n			
Scientific frame		CODE		Seedlings	Saplings	Canopy	n
Pinus resinosa Ait.		RP	Red pine	NA (0)   0	4 (4)   1	51 (85)   2	53
Pinus strobus L.	*	WP	White pine	279 (2,059)   22	77 (700)   17	9 (21)   14	30
Populus balsamifera L.		BP	Balsam poplar	NA (0)   0	187 (187)   1	NA (0)   0	1
Populus grandidentata Michx.	*	BTA	Big tooth aspen	459 (1,495)   5	73 (127)   7	12 (67)   13	20
Populus tremuloides Michx.	*	QA	Quaking aspen	342 (833)   9	98 (421)   11	11 (57)   23	31
Prunus pennsylvanica L.f.	*	PC	Pin cherry	497 (1,632)   6	70 (396)   10	6 (11)   3	16
Prunus spp.	**	СН	Cherry species	3,283 (31,632)   112	126 (863)   73	32 (191)   59	11 8
Quercus rubra L.	**	RO	Red oak	1,166 (3,382)   35	138 (605)   13	16 (64)   26	37
Sorbus americana Marsh.		MA	Mountain ash	505 (1,600)   6	4 (4)   2	11 (11)   1	7
Thuja occidentalis L.		EWC	Eastern white cedar	351 (805)   4	115 (226)   2	18 (53)   8	10
Tilia americana L.	**	BW	Basswood	884 (3,832)   65	89 (732)   31	49 (279)   78	87
<i>Tsuga canadensis</i> [L.] Carr.	*	EH	Eastern hemlock	253 (864)   13	44 (138)   13	16 (88)   29	35
Ulmus americana L.	**	AE	American elm	441 (2,400)   31	47 (389)   20	7 (21)   20	47
Ulmus rubra Muhl.		SE	Slippery elm	216 (432)   4	33 (35)   3	4 (4)   3	4

Figure 4.1[cont'd]

#### 4.3.2 Species diversity analyses

All analyses and modeling were conducted in R (R Core Team, 2018).

To characterize species diversity, we chose to calculate Hill numbers, a unified set of diversity measures parameterized by a diversity order q, which correspond to common diversity metrics (species richness, q=0; exponential of Shannon value, q=1; the inverse of Simpson diversity, q=3). All Hill numbers are interpretable as number of species; for example, q = 0 is the number of species present, and q = 1 is the effective number of common species, or the number of species, equally abundant, which would result in the same Shannon value (Chao et al., 2014). Given low species richness of our stands, we chose to focus on Hill numbers q = 0 (species richness, SR) and the q = 1 (effective number of common species, CS) (Hsieh et al., 2016).

We were limited in having sampled a fixed area for all size classes, although we did sample larger size classes in larger sample plots to attempt to achieve similar sampling intensity by size class. Despite this, there could still be imbalances in sampling intensity (by number of individuals sampled) among size classes and sites due to variable stem density. Because of these considerations, we used the iNEXT package (Hsieh et al., 2016), which conducts rarefaction and extrapolation as outlined in Chao et al. (2014) to estimate asymptotic Hill numbers. Rarefaction accounts for sampling completeness and involves generating a rarefaction curve, which shows number of species as a function of sampling intensity (number of individuals or number of sampling units). iNEXT conducts rarefaction and then extrapolates the curve to estimate the value at which the curve asymptotes, representing the theoretical species richness as sampling effort achieves completeness. We analyzed data as abundance values (total number of individuals sampled, per species) rather than incidence values (total number of plot occurrences per species), given the relatively small number of plots (25). We used the iNEXT function to calculate SR and CS for all three size classes.

We conducted generalized linear models for both diversity metrics, by size class, as a function of landscape and site-level factors, which we standardized for modeling: winter deer use (deer), average stand-level basal area (BA, m<sup>2</sup> ha<sup>-1</sup>), standard deviation of stand-level basal area (BA<sub>SD</sub>), canopy/mature (> 10 cm DBH) asymptotic species richness (SR10) and effective common species (CS10), plus groundcover percentages of hardwood litter (HWL), bare mineral soil (BMS), and coarse woody debris (CWD) and shrub coverage (shrub) for the seedling class models (Table 2.2). While we focus on the two most important groundcover classes (HWL and BMS), there were a variety of other groundcover classes recorded, such that BMS and HWL do not sum to 1. We calculated variance inflation factors (VIF), a measure of multicollinearity among predictors in linear regression models (Marquaridt, 1970), using check\_collinearity in the performance package (Lüdecke; Makowski; Waggoner; Patil, 2020); at a threshold of 5, below which predictors are generally considered to have low collinearity (James et al., 2013), none warranted removal.

In a Bayesian framework, we used a gamma distribution to model estimated asymptotic SR and CS (both of which are continuous positive). We ran models using the rjags package (Plummer, 2018), which interfaces JAGS (Just Another Gibbs Sampler) software (Plummer, 2003) with R. We ran models for 20,000 iterations after a 5,000-step adaptation period with three chains, which, based on our visualizing trace plots and calculating Gelman's Diagonal, was sufficient for model convergence. We used the coda package for model convergence diagnostics (Plummer et al., 2006).
Predictor	Acronym	Mean	Min	Max	
Deer use (% segments occupied)	DEER	19	2	60	
Basal area (m <sup>2</sup> ha <sup>-1</sup> ), stems > 10 cm DBH	BA	28	20	46	
Basal area standard deviation	BA <sub>SD</sub>	14	7	23	
Conspecific density (stems ha <sup>-1</sup> ) of stems > 10 cm DBH †	SeedS	See Ta	See Table 4.1 Canopy		
Asymptotic species richness, stems > 10 cm DBH	SR10	7	1	20	
Hill q=1 (effective common species) for stems > 10 cm DBH	CS10	3	1	7	
Shrubs (% cover) *	SHRUB	2	0	26	
Bare mineral soil (% cover) *	BMS	1	0	56	
Hardwood litter (% cover) *	HWL	91	27	97	
Coarse woody debris (% cover) *	CWD	3	1	7	

Table 4.2. Summary statistics for predictor variables included seedling and sapling models for diversity and individual species distribution. Variables include estimated deer use (DEER), average basal area ( $m^2ha^{-1}$ , BA), standard deviation of basal area (BA<sub>SD</sub>), seed source as estimated by the density of conspecific stems > 10 cm DBH (SeedS), estimated asymptotic species richness of stems > 10 cm DBH, estimated effective number of common species (Hill q=1) of stems > 10 cm DBH, % coverage of shrubs (shrub), % coverage of bare mineral soil (BMS), hardwood litter (HWL), and coarse woody debris (CWD). † SeedS was only included in the individual species models. \* SHRUB, BMS, HWL, and CWD were only used as predictors for the seedling models.

We evaluated associations with site quality through a separate analysis to preserve degrees of freedom. To determine whether ordinal site quality diversity measures were comparable between the three broad geographic regions of our study, we compared generalized linear models of diversity predicted by site quality versus a site quality – region interaction, using gamma distributions. We compared the simple model versus the interaction model using the Akaike information criterion (AIC), which is an estimator of prediction error (Akaike, 1974). The interaction model had a lower AIC for all three size classes and both diversity metrics. Since we were primarily interested in the effect of site quality, we used the emmeans function to estimate marginal means and the pairs function to conduct pairwise comparisons across site quality for each region separately (Lenth, 2020).

#### 4.3.3 Generalized linear modeling for species abundance

We included 11 species present in > 10% of stands (> 14 stands) for each of the three size classes. We included the same predictors as in the species diversity models (Table 4.2), plus an additional predictor to measure density of conspecific stems > 10 cm DBH (SeedS), a proxy for seed source (McEuen & Curran, 2004). We did not include site quality as a predictor because our chosen index of site quality and tree composition are confounded and because tree composition is included in the model in the SeedS predictor (Burger & Kotar, 2003). For each species, we subset the data to stands which that species was present on, in any size class. We compared two candidate models: a negative binomial generalized linear model and a zero inflated negative binomial generalized linear model. We checked for multicollinearity, simulated models in a Bayesian framework, and checked for model convergence using the same procedures as described for species diversity metric models. To select a model, we compared Deviance Information Criteria (DIC) values, which incorporates model fit and complexity (Spiegelhalter et al., 2002). DIC comparison led to model selection of a negative binomial model for seedlings and saplings of AB, CH, IW, RM, and SM, plus BW seedlings. BW saplings, plus AE, BF, RO, WA, and YB were modeled with a zero-inflated negative binomial model following DIC comparison (Appendix E, Table E.1). We reported selected model results at p > 0.10 given relatively low number of stands, coarse aggregation of data, and our intention to highlight trends.

#### 4.3.4 Relative abundance as a function of size class

For each stand, we calculated relative abundance for each present species by size class (stems of a given species divided by total stems). We limited our relative abundance by size class

analyses to the 18 tree species present, in any size class, on at least 10% of stands (> 14 stands). In this case, relative abundance is useful because all three size classes are portrayed on the same scale, independent of stem density, which varies widely between seedlings and canopy/mature trees; it also has the benefit of reflecting potential shifts in composition among classes as it places species-specific density in the context of competitors. For each stand, we extracted the rank order of species abundance, by size class, for each species present. For example, in a hypothetical stand, sugar maple has the highest relative abundance in canopy/mature (referred to as canopy for brevity) stems (0.7), followed by seedlings (0.6) and saplings (0.4), making the ranking "canopy – seedling – sapling"; in the same stand, hemlock had a relative abundance of 0.01 in the canopy, and 0 for seedlings and saplings, making the ranking just "canopy". We then tallied the frequency of each relative abundance ranking among stands, for each species. For example, sugar maple had the ranking "canopy – seedling – sapling" in X stands, "seedling – canopy – sapling" in Y stands, and so on. We only included sites in which a species was present in at least one size class for this analysis. This resulted in n < 141 for all species except sugar maple, which was present on all sites. We used a Kruskal-Wallis rank sum test to determine if relative abundance varied by size class, and, if significant, pairwise Wilcoxon tests for paired (within stand) values with a Bonferroni correction (n=48) to determine individual differences. To summarize differences in relative abundance among size class by species, we labeled seedlingsapling-canopy distributions with four shape categories: left skewed, right skewed, positive parabola, and negative parabola.

# 4.4 Results

#### 4.4.1 Overview of stand characteristics

BA of trees > 10 cm DBH averaged 28 m<sup>2</sup> ha<sup>-1</sup>, and BAsD averaged 14 m<sup>2</sup> ha<sup>-1</sup> (Table 4.2). SHRUB, BMS, and CWD were all low percentages (< 3 %), compared to HWL (91%). DEER was variable, ranging from 2% to 60% pellet occupancy of transect segments. We documented 25 tree species in our survey. Total stem density (stems ha<sup>-1</sup>) in stands ranged from 9,605 – 460,112 (mean 99,033) for seedlings, 357 – 12,771 (mean 3,797) for saplings, and 219 – 778 (mean 449) for canopy/mature stems (>10 cm DBH). Nine species were sampled in any size class on at least half of the sites, including sugar maple which was sampled on all stands (Table 4.1). Summing across all 141 stands, nine species had an average relative abundance of at least 1% in at least one size class (Figure 4.1). Although sugar maple had the highest average relative abundance for seedlings, saplings, and canopy/mature trees, relative abundance of ironwood and American beech were nearly equivalent to sugar maple for the sapling class (Figure 4.1).



Figure 4.1. Landscape-level (i.e., across 141 stands in Michigan) average relative abundance, by size class, for species which achieved at least 1% relative abundance in any size class; species include sugar maple (SM), American beech (AB), ironwood (IW), red maple (RM), balsam fir (BF), white ash (WA), cherry species (CH), basswood (BW), and yellow birch (YB). Scientific nomenclature and all species included in "Other" can be found in Table 4.1. Seedlings are 0 - 137 cm tall, saplings are > 137 cm tall and < 10 cm DBH, and canopy/mature are > 10 cm DBH.

Among stands and size classes the highest estimated asymptotic species richness was 20 and the lowest was 1 (Figure 4.2). Among size classes, average species richness was highest for seedlings (7.6), followed by canopy/mature stems (7.4) and saplings (6.4), though only seedlings and saplings differed from one another (p < 0.05) (Figure 4.2). Compared to species richness, effective common species was generally lower and averaged 3.2 for saplings, 2.8 for canopy/mature trees, and 2.4 for seedlings, with all size categories different from one another (p < 0.05) (Figure 4.2).



Figure 4.2. Stand-level estimated asymptotic species richness (SR) and effective number of common species (CS) for seedlings (0–137 cm tall), saplings (> 137 cm tall and < 10 cm DBH), and canopy/mature trees (abbreviated canopy for brevity, > 10 cm DBH). Asymptotic estimates account for sampling effort and are the estimated true number of species present in the stand. Within each panel, box plots labeled with the same letter are not significantly different.

### 4.4.2 Species diversity measures predicted by stand and landscape factors

There were few significant associations of species richness and diversity with factors we assessed (Table 4.3). Seedling diversity (SR, CS) was generally positively associated with canopy/mature diversity (CS10), and saplings SR was negatively associated with BA. For substrate, seedlings SR declined with BMS, while CS decreased with HWL. Diversity measures were generally flat or declining from low to high site quality for all size classes, with declines only significant for CS for seedling and canopy/mature strata in the Eastern Upper Peninsula (Figure 4.3).

Size	Variable		SR	CS		
class	variable	50%	95% CI	50%	95% CI	
	Intercept	2.3106	(1.3203 - 3.348)	1.856	(0.7659 - 2.986)	
S	Shrub	0.0069	(-0.0086 - 0.0234)	-0.0102	(-0.0274 - 0.0083)	
	Deer	0.0000	(-0.0054 - 0.0055)	-0.0004	(-0.0067 - 0.0062)	
	BA	0.0027	(-0.0121 - 0.0177)	0.0004	(-0.0158 - 0.0159)	
	$BA_{SD}$	0.0089	(-0.0124 - 0.0297)	0.0095	(-0.0145 - 0.0332)	
dling	BMS	-0.0204	(-0.03520.0047)	-0.0079	(-0.0243 - 0.0099)	
See	HWL	-0.0079	(-0.0184 - 0.002)	-0.0121	(-0.02250.0019)	
	CWD	-0.0158	(-0.0672 - 0.0357)	-0.0355	(-0.0971 - 0.0261)	
	SR10	0.0251	(0.0062 - 0.0446)	-0.0077	(-0.0304 - 0.0158)	
	CS10	0.0498	(-0.0077 - 0.1075)	0.0986	(0.0287-0.1693)	
	Beta	9.5024	(7.4123 - 11.9492)		-	
	Shape		-	7.0027	(5.4912 - 8.7879)	
	Intercept	1.8002	(1.3858 - 2.2575)	0.9976	(0.6065 - 1.4053)	
	Deer	-0.0035	(-0.0089 - 0.0019)	-0.0007	(-0.0057 - 0.0044)	
S	BA	-0.0156	(-0.030.0016)	-0.0093	(-0.0224 - 0.0038)	
Sapling	BA <sub>SD</sub>	0.0052	(-0.0171 - 0.0271)	0.0034	(-0.016 - 0.0235)	
	SR10	0.0125	(-0.0076 - 0.033)	0.0082	(-0.0096 - 0.0268)	
	CS10	0.1308	(0.07 - 0.1921)	0.1111	(0.0558 - 0.1669)	
	Beta	8.4001	(6.6 - 10.5038)		-	
	Shape		-	9.9588	(7.7741 - 12.5021)	

Table 4.3. Parameter estimations for models of species richness (SR) and diversity (CS) for seedlings (0 - 137 cm tall), saplings (> 137 cm tall and < 10 cm DBH), and canopy/mature trees (> 10 cm DBH), as predicted by total canopy/mature basal area, standard deviation of canopy/mature basal area, and deer use, plus shrub coverage and percent hardwood litter coverage (HWL), coarse woody debris (CWD), and bare mineral soil (BMS) for the seedling models. Model fit parameters (Intercept, Beta and Shape) are also provided: Beta characterizes dispersion in a negative binomial distribution (SR model only) and shape influences the shape of the gamma distribution (CS model only). Significant associations (95% CI) are bolded.



Figure 4.3. Histograms for asymptotic species richness (SR) and number of common species, CS, by site quality and region. Site quality 1 =lowest quality and 4 = highest (Burger and Kotar 2003). Letters denote significant differences (pairwise comparison of estimated marginal means) among site qualities within individual panels; bars with the same letters are not significantly different.

### 4.4.3 Individual tree species density predicted by stand and landscape factors

In contrast to community-level species richness/diversity metrics, seedling and sapling densities of individual species had numerous significant stand-level associations with key drivers, including Deer, BA, and SeedS (Figure 4.4). Balsam fir seedling densities were positively associated with BA, while saplings of red maple, sugar maple, and cherry species negatively associated with BA. Maple species had higher densities in stands with greater BA<sub>SD</sub>, while balsam fir had negative associations. SeedS had, by far, the most consistent associations among factors potentially driving species specific densities, with positive associations for all species in both seedling and sapling classes. Biotic factors were significant in a few cases, including sugar maple sapling density negatively associations with seedlings (the only size class tested), but they lacked generalization, again perhaps, due to limited variation and low coverage of CWD and BMS substrates.



Figure 4.4. Parameter estimates (median with 0.05 - 0.95 confidence interval) for seedling (0 – 137 cm tall) and sapling (> 137 cm tall and < 10 cm DBH) densities with predictor variables by tree species: average stand canopy/mature basal area (BA), standard deviation of canopy/mature basal area (BASD), deer use (Deer), seed source as estimated by conspecific canopy/mature tree density (SeedS), shrub coverage (Shrub), percent bare mineral soil (BMS), percent hardwood litter (HWL), and percent coarse woody debris (CWD). CWD, HWL, BMS, and Shrub were only included in the seedling model. Median values of significant (p > 0.10) parameters are diamonds, outlined in black. Seedlings and saplings of AB, CH, IW, RM, and SM, plus BW seedlings were modeled with a negative binomial model; all others were modeled with a zero-inflated model.

# 4.4.4 Species relative abundance across size classes

Of the 18 species analyzed, relative abundance varied significantly by size class for 16 species (Figure 4.5). Size class distributions were left-skewed (relative abundance increasing as a function of size) for seven species (e.g., eastern hemlock), a negative parabola shape (saplings having highest relative abundance) for six species (e.g., American beech), positive parabola (saplings least abundant class) for two species (e.g., sugar maple), and right-skewed (relative abundance declining with size class) for three species (e.g., white ash).



Figure 4.5. Relative abundance by species relative to stems of all species of the same size class for seedlings (0–137 cm tall), saplings (> 137 tall and < 10 cm DBH), and canopy/mature (abbreviated canopy in figure, > 10 cm DBH) strata. Within a species plot, different letters indicate different significances in relative abundance between size classes (p-value < 0.05). Size class was not a significant predictor for American elm and white pine, so we did not calculate pairwise comparisons. Histogram color corresponds to their identified shape across size classes (legend on right).

Summaries of relative abundance rankings among size classes for each stand (Table 4.4). are generally consistent with patterns suggested by the average, landscape-level patterns of relative abundance (Figure 4.5); however, stand by stand variation reveals additional details important in terms of species ecologies and management (see Discussion). For example, left skewed species are defined by canopy/mature trees as the dominant relative abundance category (Figure 4.5), and this pattern is generally supported by the frequency of relative abundance patterns at the level of individual stands (for 1<sup>st</sup> and 2<sup>nd</sup> most common patterns, canopy trees ranked first, Table 4.4). However, for individual stands, it is striking that for four of the seven species (yellow birch, paper birch, big tooth aspen and eastern hemlock) the most common stand/site level relative abundance pattern was to have only canopy/mature trees in the sample (i.e., no saplings or seedlings). Similar but oppositely for right-skewed species for two of the three species (white ash, cherry), the most common relative abundance pattern at the stand level was for samples to contain only seedlings. There were also important differences within shape groups. For example, over the landscape, red oak and sugar maple have lower relative abundance of saplings than seedlings and canopy/mature trees (positive parabola), but saplings are still present (relative abundance = 0.32 for SM and 0.01 for RO). Among stands, the most common relative abundance pattern reflects this overall landscape pattern (relative abundance of seedling > canopy > saplings). However, for red oak, the most common pattern was to have only seedlings. Although stand-level patterns (particularly size-class absences for a given species) are influenced by stem density and sampling completeness, they have clear broad implications for species ecologies and management (see Discussion).

Cl	<b>C</b>	pecies n	Seedling	Sapling %	Canopy	1 <sup>st</sup> ranking		2 <sup>nd</sup> ranking	
Snape	Species		%		%	Order	n	Order	n
	BF	77	0.09	0.32	0.58	Can-Sap-Seed	21	Can-Sap	15
Left	YB	85	0.16	0.14	0.69	Can	23	Can-Seed	14
	PB	27	0.19	0.07	0.74	Can	13	Seed	5
	SP	53	0.11	0.4	0.49	Sap 14 Ca		Can	9
Shewed	BTA	20	0.15	0.3	0.55	Can 8 Sap		Sap	4
	BW	87	0.11	0.05	0.84	Can-Seed	29	Can	17
	EH	35	0.09	0.09	0.83	Can	15	Can-Sap-Seed	6
	AB	98	0.01	0.95	0.04	Sap-Can-Seed	60	Sap-Seed-Can	20
	IW	116	0.05	0.86	0.09	Sap-Seed-Can	33	Sap-Seed	31
Negative parabola	WP	30	0.3	0.37	0.33	Seed	7	Sap-Seed	5
	QA	31	0.06	0.29	0.65	Can	14	Sap	5
	PC	16	0.31	0.5	0.19	Sap	7	Seed	5
	AE	47	0.51	0.23	0.26	Seed	17	Can	6
Positive	SM	141	0.46	0.11	0.43	Seed-Can-Sap	51	Can-Seed-Sap	43
parabola	RO	37	0.41	0.14	0.46	Seed	10	Can-Seed	9
Right skewed	RM	119	0.8	0.08	0.13	Seed-Can-Sap	35	Seed	31
	WA	72	0.78	0.18	0.04	Seed	19	Seed-Sap	15
	CH	118	0.55	0.25	0.2	Seed	31	Sap-Seed	17

Table 4.4. Summaries of stand-specific relative abundance rankings by size class for 18 species surveyed in > 10 % of our 141 stands. Scientific name and authorship can be looked up by species code in Table 4.1. Species are grouped by the shape of their average relative abundance size class distribution, and n is the number of sites that a species was present on, in any size class. Seedling, sapling, and canopy % indicates the percentage of stands in which that size class had the highest relative abundance (seedling 0 - 137 cm tall, saplings > 137 cm and < 10 cm DBH, and canopy/mature, abbreviated as canopy, > 10 cm DBH). 1<sup>st</sup> and 2<sup>nd</sup> ranking refer to two most common rankings of relative abundances by size class (n is number of stands). The nomenclature presents the size classes in decreasing rank order (i.e., the first size category listed had the highest relative abundance), with size classes which had zero stems omitted. For example, 21 out of 77 stands in which balsam fir (BF) was present had the relative abundance ranking "Can-Sap-Seed" (canopy is highest relative abundance); as a left-skewed species, it has greatest relative abundance in the canopy, on average.

# 4.5 Discussion

#### 4.5.1 Overview

Our results suggest individual stands contain few of the total tree species present in northern hardwood forests for seedling, sapling, and canopy/mature size classes and underscore current regeneration concerns. Despite sampling 25 tree species across the landscape, most had limited geographic distribution and low abundance; fewer than ten species were sampled in most stands or achieved at least 1% average relative abundance in any size class averaged across sites (landscape scale). At the stand level, all three size classes averaged fewer than three effective common species (CS), and although SR was similar between seedlings and canopy/mature stems, natural species filtering of sub-canopy and disease-impacted species means canopy/mature species richness is likely to decline as these seedling cohorts recruit to the canopy. Diversity patterns furthermore appear to be self-perpetuating, as stands with lower canopy/mature diversity (CS of stems > 10 cm DBH, CS10) tended to have lower seedling and sapling diversity (Table 4.3). These results have concerning implications for future stand resilience of managed northern hardwood forests in Michigan.

Our diversity metrics are similar, if not slightly lower, than past studies of northern hardwood forests. However, contextualizing our results in relation to past literature is hampered by inconsistencies with scales of analysis and diversity metric calculations and limited studies analyzing diversity in managed northern hardwood forests at comparable scales. Average standlevel tree species richness of our stands, for all size classes, was lower than values reported for managed and old growth stands in Crow et al. (2002), though Crow covered a smaller geographic range than analyzed here and reported values for aggregates of three stands, which may inflate species diversity metrics, if stand histories, site qualities, or other factors differ. Our

results suggest CS was generally lower than shown in Danyagri et al. (2019), although that comparison relies on modeled estimates rather than direct comparison of range and mean. Sugar maple in our stands dominated the seedling class similarly to the managed northern hardwoods characterized in Neuendorff et al. (2007) and Crow et al. (2002) but is less dominant in the sapling size class than described in these studies, both of which were conducted in a limited geographic region in the western Upper Peninsula of Michigan which has been shown to support higher densities of sugar maple (Henry et al., 2021). Rather, our saplings patterns are characterized by roughly even representation of sugar maple, ironwood, and American beech (Figure 4.1), similar to findings in Angers et al. (2005), near Ottawa, Canada, and Elenitsky et al. (2020), in north-eastern Michigan, both of which are in regions of relatively lower lake-effect snowfall and therefore likely higher deer browsing pressure. Our results add to the pool of studies suggesting that existing tree regeneration diversity is unlikely to yield diverse and resilient canopies across a broad geographic range.

We found mixed support for our first hypothesis that tree diversity would be highest for canopy/mature stems and on low quality sites. Seedling and canopy/mature stems tied for highest SR, but saplings had highest CS (Figure 4.2). However, although saplings are generally more equally represented by different species (Figure 4.1), four of the six species of highest relative abundance have low likelihood of attaining our canopy size class (> 10 cm DBH) and are even less likely to reach the upper canopy stratum of northern hardwood forests (i.e., 25 - 35 m tall) due to size-dependent pest and pathogens (i.e., American beech, ash species) or small maximum height (i.e., subcanopy species ironwood and balsam fir). Furthermore, although regeneration SR was comparable to the canopy, the actual pool of species with the potential to reach the canopy is smaller as seedlings must successfully transition through the sapling class to the canopy class,

and transition to sapling class is a bottleneck driven by several factors, including deer browsing (this study, Walters et al 2022) and deep shade/high BA (this study, Henry et al 2021, Walters et al. 2022). Overall, patterns across size classes suggest a future decline in canopy species richness. Patterns of diversity with site quality matched our hypotheses in the EUP, with CS generally declining with increasing site quality, but not in the other regions and only significantly for seedling and canopy/mature CS. Failure to detect patterns of diversity with site quality in other regions may be due to associations between high-quality sites and other factors. For example, deer use was positively associated with site quality in the WUP (Figure 1 in Henry et al., 2021). However, this does little to explain the lack of a negative site quality-sapling diversity relationship in the WUP as both high deer use and high site quality would be expected negatively associated with sapling diversity which we did not see. However, expectations that deer diminish diversity are not supported (this study). Via selective browsing deer likely impact composition more than diversity (this study; Walters et al., 2022).

Our results support our overall second hypothesis, that tree species importance varies by size class (Figure 4.1), and many of our proposed mechanisms consistent with specific size pattern shapes were supported as well (Hypotheses 2a-d). Regarding Hypothesis 2a, we find that local seed source limitations may play a dominant role in driving regeneration composition and density. Longer-distance dispersal of seed from adjacent stands is unlikely, particularly for large-seeded, passively dispersed species such that local/intra-stand seed sources likely dominate seed availability (McEuan & Curran, 2002; Willis et al, 2016), and certainly sprout availability. Across all species and most regeneration size classes, we found positive associations with conspecific canopy/mature density for both large and small seeded species, and overall positive associations of seedling and sapling diversity measures with canopy/mature species diversity.

This finding supports the notion that local canopy seed/sprout sources have a strong bearing on seedling/sapling composition. Fewer significant associations between SR10 and regeneration diversity metrics as compared to CS10 suggests that canopy/mature density, rather than presence alone, is key for driving diverse regeneration. A caveat to our local seed/sprout source limitation interpretation is that other factors could contribute to canopy-sapling-seedling compositional continuity including site factors (e.g., soil water regimes) that favor establishment/growth/survival from seedling to maturity. (Webster et al., 2018).

Our results, particularly our relative abundance analyses, suggest substrate limitation for small-seeded species, supporting our hypothesis 2b. Many small-seeded species, including eastern hemlock, white spruce, and birch species, had left-skewed relative abundances, with lower relative abundance in the understory compared to the overstory, supporting potential substrate limitation. Some of those species, such as eastern hemlock, have well-documented narrow seedling establishment substrate requirements with those substrates currently uncommon in managed forests (summarized in Alverson et al., 2019), which place our results in accordance with existing literature.

However, distribution of relative abundance by size class is likely driven by multiple interacting factors and must be interpreted carefully. In addition to seedling establishment limitations, other species traits/properties may also contribute to left-skewed distributions. For example, for balsam fir and spruce high long-term survival, low growth rates for saplings and low deer browsing pressure may lead to the accumulation relatively high sapling densities. Another example, many of our small-seeded species (which is generally associated with narrow establishment substrate requirements: Marx and Walters, 2008; Bolton and D'Amato, 2011;

Willis et al., 2016) are also highly shade intolerant (e.g., PB, BTA) such that seedling and sapling densities are constrained by both establishment substrate and high low light mortality.

Although substrate-limitation for small seeded species was supported by several relative abundance patterns, we identified few significant associates with species regeneration abundance and forest substrate composition (percent hardwood litter (HWL) or bare mineral soil (BMS)). However, several factors may explain this, particularly overall very low average coverage of BMS and CWD (Table 4.1). Less common species also had fewer non-zero data points, making association more difficult to detect, and our broad height range of seedlings (< 137 cm tall) may include stems that established decades ago (e.g., stems can persist < 1 m for > 30 years; Marks and Gardescu, 1998), which would potentially only weakly (or not at all) associate with current substrate conditions. Lastly, substrate limitation may be difficult to detect at a stand-scale, instead better analyzed at the plot level, as in Willis et al. (2016). Nevertheless, patterns of relative abundance suggest potential for seedling establishment to prove limiting in our stands dominated by hardwood litter coverage.

Consistent with hypothesis 2c, high BA (i.e., low light) may strongly limit the transition from seedling to saplings for mid-shade tolerant species, including cherry, red maple, red oak, and tolerant sugar maple. Moderate to high shade tolerance can facilitate development of multiple, longer-lived seedling cohorts, which can accumulate as dense seedling populations; however, tolerance is not sufficient to support survival and growth into the sapling class, at least not in the low light environments characterizing dominant single tree selection silviculture in Great Lakes NHF. In general, negative associations of density with BA, positive associations with BA<sub>SD</sub> and positive parabolic or right-skewed relative abundance distributions for midtolerant cherries, maples, and red oak all support the notion of low light limitation of seedlingsapling recruitment. In addition to direct low-light/high BA limitations (i.e., mortality via carbon starvation) to seedling-sapling transitions for mid tolerant species, low-light conditions could exacerbate resource shortfalls for tree seedlings by negatively impacting arbuscular mycorrhizal (AM) colonization (Cheng et al., 2005; Koorem et al., 2017; Neuenkamp et al., 2021); red maple, cherry species, and sugar maple are all AM associated species (Bennett et al., 2017).

In addition to high BA (light) limitations affecting relative abundance patterns across size classes, deer browsing may also drive declines in relative abundance from seedling to saplings for maples, cherries, and red oak, partially supporting hypothesis 2d. This pattern is most clearly supported for sugar maple, given significant negative associations between sapling density and deer use. Previous studies suggest deer are a likely contributing factor in regeneration bottlenecks, including for sugar maple (Côté et al., 2004; Leak, 2006; Neuendorff et al., 2007; Powers & Nagel, 2009; Matonis et al., 2011) oak species (Dey, 2014) and species groups including the maples, cherries and red oak examined here (Walters et al. 2022). In contrast to the positive parabola and right skewed forms for maples, cherries and red oak, several species know to be little impacted by deer browsing pressure (ironwood, American beech, white pine) had positive parabolic shape among size classes (i.e., saplings most abundant) which also supports the notion that deer are important drivers of compositional shifts. It is important to note that our species abundance model results regarding deer effects provided only limited support for our interpretation of deer effects based on size class changes in relative abundance. Sugar maple was the only species to negatively associated with DEER, and surprisingly, red oak saplings had a positive association. Our modeling approach may have failed to detect associations, in part, due to low stand-level detection and low abundances of many species (e.g., yellow birch). Densities

are also influenced by other factors previously discussed, such as low light and substrate limitation, which can further mask the effect of deer browsing.

# 4.5.2 Management implications

In the context of forest management, our results support several potential alternative management strategies aimed at overcoming seed, substrate, and growth/survival (e.g., BA, deer browsing) limitations to establish species-diverse seedling and sapling classes. Given the current paucity of many tree species in all strata (and critically in the seed producing mature/canopy class), active seeding or planting of additional tree species may be necessary to promote greater diversity. Tree species selection for planting could be coupled with assisted migration strategies to further prepare forests for a variety of future climate scenarios. Provisions for exposing more mineral soil during harvest operations and leaving more trees to die and become coarse woody debris on the forest floor would help overcome establishment substrate limitations for several species, especially those with small seeds (Marx & Walters, 2008; Willis et al., 2015). Reducing basal area more than is done with current partial harvesting systems in stands would result in higher light levels than currently created following harvest and possibly increase sapling recruitment of shade mid-tolerant and tolerant species. Finally, implementing alternative strategies to deter deer browsing, such as leaving treetops as natural barriers, may increase regeneration density and recruitment of species already present on the landscape. Integrated strategies including several of the management modifications are likely to be most effective in increasing species diversity in NHF in the long term.

Another management implication of our results is that the high variability we found across species in seedling vs sapling abundances (and ultimately to canopy/mature classes) emphasizes the potential risk in assuming seedling composition reflects sapling composition.

This is particularly important for developing meaningful assessments of regeneration success following harvest. Several factors, including shrub competition (Walters et al. 2016), deer browsing (Henry et al. 2021, this paper) and light resource limitation (i.e., canopy BA, Henry et al. 2021, Walters et al. 2022, this paper), may assert strong and variable bottlenecks on seedling to sapling transitions, such that tree seedlings are poor predictors of regeneration success (Walters et al. 2020a). Managers should focus on saplings (0 - 10 cm DBH) for assessment of regeneration success following harvest or other management activity.

With certain key species potentially in decline, relative abundance analyses can be used generally to identify species occupying their relinquished growing space, namely right-skewed species (increasing relative abundance with decreasing size class) and negative parabolic species (with saplings highest relative abundance). Right-skewed species include red maple, which has been increasing in relative abundance in recent decades (Fei & Steiner, 2007). Many of the species which compete with sugar maple for sapling growing space are negative parabolic, including American beech, which creates well-documented dense thickets (Gravel et al., 2011; Collin et al., 2017; Elenitsky et al., 2020), and ironwood, which is a sub-canopy species. Relative abundance analyses prove to be a useful method of quantifying potential species trajectories, when incorporated with literature and knowledge of tree species' life history dynamics.

## 4.5.3 Caveats

Our snapshot-in-time survey is unable to assess trends within size classes over time, and instead relies on assumptions about current patterns and implications for ultimate recruitment; previous studies have suggested comparing size classes from a single survey may not be appropriate for inferring temporal compositional shifts (in the case of tree range shifts, Malis et al., 2016). With this caveat, our results nonetheless suggest that current managed northern

hardwood forests may face long-term challenges with resilience due to low diversity of species without proactive management interventions. Furthermore, our analysis is explicitly on managed systems, where seedling and sapling classes provide the raw material for management options aimed at future compositional goals, which enhances the utility/relevance of such analysis. However, follow-up analyses of species diversity dynamics over time would be useful in confirming the temporal component suggested by our relative abundance analyses, given the long-term maintenance of current compositional drivers over time (e.g., current management regimes, deer use). Nevertheless, our results highlight the importance of intervention to promote diverse, resilient stands which can adapt to changing timber market shifts, pest/pathogen introductions, and climatic changes. Applying ecological knowledge to these stands is vital for continued ecological functioning in managed northern hardwood forests.

#### **CHAPTER 5**

## CONCLUSIONS

The preceding chapters explored the outcomes of 60 + years of single-tree selection management, across a broad geographic gradient, in driving forest regeneration and recruitment outcomes. Chapters 2 and 3 emphasized that regeneration and recruitment have repeatedly failed for the focal species, sugar maple. Not only is sugar maple poorly regenerating, but it is being outcompeted by species less desirable for management which often do not or cannot recruit to the canopy. Overall, stands are characterized by low tree species diversity (Chapter 4). These results have critical implications for northern hardwood forest management.

Analysis of sugar maple regeneration patterns and associations with key drivers confirms that sugar maple regeneration failure is geographically widespread and occurs on a variety of site conditions. While densities of sugar maple seedlings (< 50 cm tall) were generally high (though potentially limited on some sites by low mature sugar maple seed sources and/or competing taller vegetation), sugar maple stems > 50 cm tall and < 5 cm DBH were sparse and often entirely absent. Densities were particularly sparse in portions of the northern Lower Peninsula and the central southern Upper Peninsula; they also associated negatively with increasing winter deer use and understory competition, and positively with seed availability and medium site qualities. Our results echo previous findings of low sugar maple regeneration densities (Leak, 2006; Neuendorff et al., 2007; Powers & Nagel, 2009; Matonis et al., 2011), although differed from (Vickers et al., 2019) which analyzed all northern hardwood forests rather than being focused on selection managed stands. Our results emphasized that current stands are poorly stocked and unlikely to successfully regenerate new cohorts of sugar maple, given current stand dynamics.

Analysis of age structure of 1499 sugar maple basal discs from 51 stands indicated that single-tree selection has chronically failed to recruit new age cohorts of sugar maple in the past 60 + years of management (Chapter 3). As assumed from Michigan forest and logging history, most stand canopies are dominated by stems 90 - 120 years old; however, in contrast with expectations under uneven-aged single-tree selection management, we identified few age cohorts < 70 years old (36 of 160 identified age cohorts), with no stands stocked in stems 30 - 50 years old and only 7 of 51 stands stocked in stems 50 - 70 years old. Saplings average 66 years old, poletimber averages 91 years, and sawtimber averages 106 years. Regions with greater January precipitation (snowfall) and lower annual temperatures, where deer populations have been historically low, generally have younger saplings. These results contradict previous findings of sugar maple age structure in managed northern hardwood forests of the Great Lakes region, which typically identified young (< 60-year-old) saplings and linearly related age and diameter (Tubbs, 1977a; Dey et al., 2017; Harmala, 2021). Our results indicate that sugar maple recruitment has persistently failed in northern hardwood forests managed by single-tree selection, and relying on diameter distribution to assess regeneration and recruitment success may often be misleading; our results have concerning implications for future growth of aging northern hardwood stands.

Despite declines in sugar maple understory dominance, few desirable species appear to have taken its place (Chapter 4). After accounting for sample size, stand-level species richness was slightly lower for saplings (> 137 m tall and < 10 cm DBH) versus seedlings (< 137 m tall), though both classes were comparable to canopy/mature species richness (> 10 cm DBH), and all size classes generally averaged fewer than three effective common species. Stands with lower canopy species diversity tended to have lower regeneration species diversity, suggesting

diversity is self-perpetuating. Our results suggest likely influence of substrate and light limitation, given patterns of relative abundance by size class for individual species and secondarily model results for individual species abundance. At the stand level, sugar maple demonstrates a "bottleneck" of lower sapling diversity compared to seedlings or canopy/mature stems, potentially driven by a legacy of deer browsing pressure and low-light levels prohibiting gap capture. These dynamics are similar to those found in past studies of managed northern hardwood forests (Angers et al., 2005; Danyagri et al., 2019; Elenitsky et al., 2020), and diversity is generally lower than past analyses of old growth northern hardwood forests (Crow et al., 2002). Should regeneration diversity patterns sustain through canopy recruitment, they signal future shifts in mature canopy stratum composition. Our results highlight a need for changes in management aimed at increasing tree species diversity.

Together, these results suggest that continued selection silviculture in many areas would likely result in continued decline in species diversity and potential canopy failure, due to low density of desirable stems for management. Across our analyses, light limitation is a consistent associated factor with undesirable regeneration and recruitment outcomes. This suggests that stands may benefit from more intense harvesting which results in greater light availability. Based on our results, this could benefit not only increased recruitment of shade-intolerant species but the relatively shade-tolerant sugar maple species as well. Deer browsing also proves to be a consistent threat to northern hardwood forest regeneration. Since direct reduction of the deer population is socially unlikely, silvicultural prescriptions intended to reduce the impacts of deer on regeneration, such as treetops acting as natural browsing barriers, may improve regeneration outcomes. Lastly, the low diversity of canopy/mature tree species and associated lack of seed source, appears to limit both large- and small-seeded species. Given lack of local seed source,

future management would benefit from direct seeding or planting to increase future forest diversity; this could also benefit future climate adaptations by selecting species which are projected to be better suited to the future climate, i.e., assisted migration. However, these efforts are likely to be challenged by continued, persistent deer browsing.

Although there are numerous potential directions for related research, two avenues emerged as particularly promising areas for future work. First, given the unprecedented results of our sugar maple age structure analysis, additional research quantifying ring width of the samples would be prudent. This could elucidate changes in sugar maple growth and recruitment over time and could further explore past stand dynamics, such as growth response to disturbance (harvest). The second area of research is review and quantification of species diversity goals for northern hardwood forests. Although studies of northern hardwood forests and discussions of the importance of species diversity proliferate, there are no clear guidelines on what constitutes adequate species diversity to promote forest resilience. Furthermore, existing literature lacks consistency in past reporting of species diversity measures, which hinders efforts to define successful levels of tree diversity. These areas of research would further our understanding of forest stand dynamics in managed northern hardwood forests and guide future management.

In conclusion, these results emphasize that single-tree selection management has not resulted in desired outcomes. Namely, stands have low tree diversity and low densities of desirable sugar maple stems, which have persistently failed to recruit to the canopy. Forest managers in many regions will likely need to consider alternative silvicultural regimes in the immediate future to sustain forest health and functioning. APPENDICES

# APPENDIX A

Details on the deer model

### A.1 Modeling framework

We developed a two-stage modeling approach: Stage 1 was a deer model used to impute pre-harvest deer occurrence information to all stands; Stage 2, was a set of sugar maple regeneration models that were conditioned on Stage 1 (deer model) and a set of other key predictor variables. More specifically, given pre-harvest deer use surveys for a subset of our stands (a key sugar maple regeneration predictor), the Stage 1 stand-level deer use model provided point estimates of deer use, as predicted by climate and landcover variables, postharvest deer use measurements, and a spatial random effect. Stage 2 comprised a multilevel model for plot-level sugar maple regeneration counts at three sizes classes as predicted by standlevel deer use from Stage 1 and stand vegetation structure and site quality measurements. All analyses and modeling were conducted in R (R Core Team, 2018).

With a Bayesian framework, we modeled deer use prior to timber harvest (with deer use defined as the percent of transect segments with deer fecal pellets) as predicted by climate, landcover, post-harvest deer use, and a spatial random effect. Given the purpose of this model was to infer pre-harvest deer use in non-surveyed stands, we aimed to build the best predictive model rather than explain biological processes. Modeling percent of occupied transect segments (i.e., binary responses for each transect segment and rate or percent when considering multiple transect segments within a site) rather than counts or "deer use days," the latter of which is common in the literature (e.g., (Yañez-Arenas et al., 2012), offers several advantages. Using simple pellet occurrence within a transect segment removes the need to count individual pellets or judge what constitutes a pellet group while incorporating uniformity of deposition. Moving to pellet occurrence also mitigated the potential for unusually abundant pellet group counts in a single segment which tended to provide unrealistically high stand-level deer density estimates.

This approach remains directly comparable to previous models based on counts (linear relationship between percent and counts for our data,  $R^2 = 0.90$ ). The bounded response variable expressed as a percent or rate also provides straightforward interpretation. We used a spatially stratified sampling approach (Fig. A.1). Despite a relatively broad range of sample dates, we conducted linear regression to test whether sampling date correlated with total pellet group counts and found no relationship (data not shown).



Figure A.1. Design of the winter pellet survey transects. Points represent 25 sampling locations for vegetation surveys within the 12.14 ha square unit. Each transect (blue line) was 6 m wide, and cumulative transect length was 628 m. For analysis, transects were divided into spatially referenced segments ~ 6 m long. Occurrence of deer pellets were documented for each segment.

The predictors for pre-harvest (2017) deer use include average daily snow depth (01 November 2016 - 30 April 2017, 1 km resolution, (National Operational Hydrologic Remote Sensing Center, 2004); average minimum monthly winter temperature (December 2016 -February 2017, 4 km resolution, 1981-2010, PRISM Climate Group, Oregon State University, http://prism.oregonstate.edu, created 18 October 2019); 2019 (post-harvest) deer use by timber harvest type; and local coverage of deciduous, evergreen, and mixed forests as well as agricultural lands within 25 km<sup>2</sup> of the stand (NLCD, 30m resolution, aggregated to 1 km resolution for faster processing). These are well-documented drivers of deer populations in our study region (snow depth (Moen, 1976; Shi et al., 2006); winter low temperatures (Holter et al., 1975); and proximity to coniferous, deciduous, and agricultural land (Telfer, 1970; Ozoga & Gysel, 1972; Wetzel et al., 1975; Millington et al., 2010; Dawe et al., 2014). Although we acknowledge that local deer use can be concentrated by timber harvest activities (e.g., (Campbell et al., 2004), post-harvest deer use of a stand reflects relative availability of deer occupying the surrounding landscape thereby serving as a useful proxy for browse pressure.

For the complete collection of predictor variables, we calculated variance inflation factors using the vif function in the car package (Fox & Weisberg, 2011) to check for multicollinearity; at a threshold of 2, we removed minimum winter temperature. With the remaining predictors, we modeled pre-harvest deer use (percentage of transect segments occupied by deer pellets at each stand,  $y_{deer,i}$ ) using a polya-gamma binomial regression (Polson et al., 2013) and a spatially varying random intercept (Finley & Banerjee, 2020). The model for the *i*th stand was:

$$y_{deer,i} PG(n_i, \psi_i)$$
$$log(\psi_i) = \mu_{deer} + \beta_{deer,i} X_{deer,i} + w_i$$

where  $n_i$  is the number of trails/transect segments ( $n_i$  is 103 for all stands) and  $\psi_i$  is the standspecific likelihood of any transect segment being occupied (success); the log of  $\psi_i$  is defined by a linear equation including the grand mean  $\mu_{deer}$ , vector of regression coefficients  $\beta_{deer,i}$  and associated predictor design matrix  $X_{deer,i}$ , and the spatially varying random intercept  $w_i$ . Predictive inference was based on 50,000 post burn-in samples from three Markov chain Monte Carlo (MCMC) chains. We assessed model convergence using Gelman Rubin diagnostics in the coda package (Plummer et al., 2006), trace plots, and plots of residuals vs. predicted, all of which indicated model fit. Post burn-in samples were then used to generate predicted  $\psi_i$  for all stands for subsequent use in the Stage 2 vegetation model.

# A.2 Results and conclusions

Across most of the study area, winter deer use during the winter of 2017 was estimated as low to moderate, except for relatively high use in the south-central and eastern parts of the Upper Peninsula (Fig. 2). Our model estimated an effective spatial range of 7.7 km (95% CI 2.8 – 27.5 km), indicating that sites > 7.7 km apart on average were not spatially autocorrelated (Table A.1). None of the weather or landcover variables significantly predicted deer use, and only 2019 deer pellets in single-tree selection harvest were positively associated with 2017 pellet counts (Table A.1). Our deer use predictions (Fig. 2) aligned with other smaller-scale studies of deer use in our study region (Shi et al., 2006; Millington et al., 2010); this, along with reasonable estimates of spatial autocorrelation (~ 7 km) that correspond to deer winter habitat range (Van Deelen et al., 1998), gave us confidence that our model reasonably predicted deer use for sites in our study area.

Predictor	Median (95% CI)			
Intercept	-1.013 (-2.590 - 0.663)			
Average snow depth	-0.015 (-0.038 - 0.008)			
Deciduous forest cover	-0.915 (-2.606-0.739)			
Evergreen forest cover	-1.129 (-6.265 - 3.594)			
Mixed forest cover	4.407 (-1.354 - 10.520)			
Agricultural land cover	3.492 (-4.621 - 12.052)			
2019 pellets (single-tree selection harvest)	0.019 (0.000 - 0.037)			
2019 pellets (seed tree harvest)	0.007 (-0.007 - 0.020)			
2019 pellets (shelterwood harvest)	0.011 (-0.003 - 0.024)			
2019 pellets (group selection harvest)	0.013 (-0.002 - 0.029)			
Sigma squared	0.695 (0.441 - 1.146)			
Effective special range (km)	7.086 (3.789 – 24.984)			

Table A.1. Results from the stage one deer use model are presented with their median value and a 95% Bayesian confidence interval. Sigma squared is the variance parameter for the spatial term, and effective spatial range represents the distance at which spatial autocorrelation becomes insignificant in the model, in kilometers. None of the climatic or landscape predictor variables were significantly different from zero, likely because the spatial term explained most of the variation.

# APPENDIX B

Additional information on site quality categorizations

Site Quality	Abbrev.	NLP		EUP		WUP		TOTAL
Poor - poor/medium	Q1	PArVVb	4	PArVAa, ATFD,	27	pArVAa, AArAst, AArLy, PArVAa[w]	8	39
Medium*	Q2	-	-	AFPo	10	ATM-Sm, ATM- O, ATFAs	12	22
Medium/rich	Q3	AFO	23	AFOAs	8	ATD-Ca, ATD- Hp, AVb	16	47
Rich - very rich	Q4	AFOCa	21	-	-	AOCa	12	33
TOTAL			<b>4</b> 8		45		<i>48</i>	141

Table B.1. Bayesian multilevel model predicting sugar maple regeneration in Michigan included four ordinal categories of site quality (*sensu* (Burger & Kotar, 2003). Kotar site quality grouping, model abbreviation, Kotar types (see Table B.2) included in the grouping, and count of stands sampled by region. The full set of site qualities was only present in one of the regions. NLP = Northern Lower Peninsula, EUP = Eastern Upper Peninsula, WUP = Western Upper Peninsula. \*Medium site quality was included in the intercept.
Habitat type	Region	Primary landforms	Primary soils	Soil moisture regime	Soil nutrient regime
PArVVb (Pinus strobus – Acer rubrum/Vaccinium- Viburnum acerifolium)	NLP	Outwash plains, ground moraines, and beach ridges	Moderately well drained sandy soils	Dry- Mesic	Poor to medium
AFO (Acer saccharum – Fagus grandifolia/Osmorhiza claytoni)	NLP	End moraines, ground moraines, and outwash plains	Well drained sandy soils	Mesic	Medium to rich
AFOCa (Acer saccharum – Fagus grandifolia/Osmorhiza claytoni-Caulophyllum thalictroides)	NLP	End and ground moraines	Well to moderately well drained deep sandy loam till	Mesic	Rich to very rich
PArVVa (Pinus strobus – Acer rubrum/Vaccinium angustifolium – Aralia nudicaulis)	EUP	Deep lacustrine deposits of sand and gravel	Excessively to well drained sandy soils	Dry to dry- mesic	Poor
ATFD (Acer saccharum – Tsuga canadensis – Fagus grandifolia/Dryopteris spinulosa)	EUP	Outwash, lacustrine deposits, glacial till, and end morains	Well to moderately well drained deep sands	Mesic	Poor to medium
AFPO (Acer saccharum – Fagus grandifolia/Polygonatum pubescens)	EUP	Variety of landforms	Well to somewhat excessively drained deep sands and loamy sands; gravelly, cemented, and mottled layers common	Mesic	Medium

Table B.2. Descriptions of site quality types (Burger and Kotar 2003) used to model sugar maple regeneration. NLP = northern Lower Peninsula, WUP = western Upper Peninsula, and EUP = eastern Upper Peninsula.

Habitat type	Region	Primary landforms	Primary soils	Soil moisture regime	Soil nutrient regime
AFOAs (Acer saccharum – Fagus grandifolia/Osmorhiza claytoni – Arisaema atrorubens)	EUP	End moraines and till plains	Moderately well to excessively drained soils; gravelly, cemented, and mottled layers common; thin till over bedrock	Mesic	Medium to rich
pArVAa (Pinus strobus – Acer rubrum/Vaccinium angustifolium – Aralia nudicaulis)	WUP	Glaciofluvial deposits, moraines, lake plains	Excessively well drained soils of lacustrine deposits	Dry to dry- mesic	Poor
AArAst (Acer saccharum – Acer rubrum/Aster macrophyllus)	WUP	Coarse till and shallow till over bedrock	Sandy soils	Dry- mesic	Poor to medium
AArLy (Acer saccharum – Acer rubrum/Lycopodium annotinum)	WUP	Coarse till deposits and thin till over bedrock	Loamy soils	Mesic	Poor to medium
PArVAa[w] (Pinus strobus – Acer rubrum/Vaccinium angustifolium – Aralia nudicaulis [Wisconsin variant])	WUP	Glacial outwash and moraines	Sand to sandy loam	Dry- mesic	Poor to medium
ATM-Sm (Acer saccharum – Tsuga canadensis/Maianthemum canadense – Smilacina racemose variant)	WUP	End moraines and outwash covered moraines	Loamy sand and sandy loam soils	Mesic	Medium
ATM-O (Acer saccharum – Tsuga canadensis/Maianthemum canadense – Osmorhiza claytoni variant)	WUP	Clay and lacustrine deposits	Sandy loam soils	Mesic	Medium

# Table B.2 [cont'd]

Habitat type	Region	Primary landforms	Primary soils	Soil moisture regime	Soil nutrient regime
ATFAs (Acer saccharum – Tusga canadensis – Fagus grandifolia/Arisaema atrorubens)	WUP	Lacustrine deposits of sand and glaciofluvial deposits	Sandy soils with subsurface clayey, gravelly, or cemented layers	Mesic	Medium
ATD-Ca (Acer saccharum – Tsuga canadensis/Dryopteris spinulosa – Caulophyllum thalictroides variant)	WUP	Clay deposits	Loamy cap soils	Mesic to wet- mesic	Rich
ATD-Hp (Acer saccharum – Tsuga canadensis/Dryopteris spinulosa – Hepatica variant)	WUP	Medium textured glacial till	Sandy soils with subsurface clayey, gravelly, or cemented layers	Mesic	Medium to rich
AVb (Acer saccharum/Viburnum acerifolium)	WUP	Medium textured end moraines	Sandy loams	Dry- mesic	Medium to rich
AOCa (Acer saccharum/Osmorhiza claytoni – Caulophyllum thalictroides)	WUP	Moraines and loess deposits	Well drained loamy till and loess	Mesic	Rich to very rich

# Table B.2 [cont'd]

## APPENDIX C

Sugar maple model: parameter and model selection

#### C.1 Parameter selection

We compared linear models of log transformed stand-level average sugar maple seedling counts (+ 0.001 to account for zeros) predicted by different measures of canopy sugar maple abundance. For predictors, we compared average basal area and stem count of sugar maple trees within the 12.6 m<sup>2</sup> plots (25 plots per stand) greater than thresholds of 25, 30, 35, and 50 cm DBH. Basal area of sugar maple canopy trees greater than 25cm DBH had the greatest predictive power and was included in the final model (Table C.1).

Diameter threshold (cm)	Stem count adjusted R <sup>2</sup>	Basal area adjusted R <sup>2</sup>
25	0.08871	0.1171
30	0.09429	0.1128
35	0.1031	0.1068
40	0.06418	0.07021

Table C.1. Adjusted R<sup>2</sup> values for a linear models of stand-level log average sugar maple seedling counts predicted by different measures of canopy sugar maple density. Predictors included stand-level average stem count and basal area of sugar maple trees within 6 m<sup>2</sup> radius plots, at varying size thresholds. The basal area of sugar maple trees > 25 cm had the highest adjusted R<sup>2</sup> model and was used as the predictor of seed availability in model used to predict sugar maple regeneration.

Based on (Elenitsky et al., 2020), we assessed whether densities of sugar maple

regeneration in three size classes were comparable within ordinal site quality categorization, across three geographic regions for which our habitat classification system is parameterized (Burger & Kotar, 2003). We conducted Wilcoxon paired t-tests with a Bonferroni correction (n=24) (Table C.2). Only two of the 24 t-test comparisons significantly differed from zero, meaning that for those two stands, densities were not comparable within the site qualities across the two regions being compared. This suggests that overall regeneration densities are comparable across regions in the same site quality categorization. We also conducted recursive partitioning to indicate the most parsimonious groupings, in this case of site-level sugar maple densities as predicted by region and ordinal site quality (Figs. C.1). If clear trends exist that regeneration is first grouped by region (higher at the top of the tree) and then by site quality, this would indicate that stands are not comparable within the same site quality categorization across regions; we failed to detect clear evidence of this pattern. Collectively, these results indicated that ordinal site quality by region was an appropriate predictor variable in our sugar maple regeneration model.

Size Class	Habitat class	NLP- EUP	EUP- WUP	WUP- NLP
Seedlings	Seedlings Poor to poor/medium		0.0078	0.3879
	Medium	-	1	-
	Medium/rich	1	1	1
	Rich to very rich	-	-	1
Small	Poor to poor/medium	1	.03	1
saplings	Medium	-	1	-
	Medium/rich	1	1	1
	Rich to very rich	-	-	1
Small	Poor to poor/medium	1	1	1
saplings	Medium	-	1	-
	Medium/rich	1	1	1
	Rich to very rich	-	-	1

Table C.2. Results of the pairwise Wilcoxon tests using a Bonferroni correction (n=24). Values indicate probability of the distributions of the pair being significantly different. Values significant at adjusted p-values of 0.05 are highlighted.



#### A.

Figure C.1. Decision tree generated by ANOVA model recursive partitioning for total stand-level counts of sugar maple seedlings (C.1A), small saplings (C.1B), and large saplings (C.1C). Predictive inputs were unique identifiers for site quality (1st, 2nd, 3rd, 4th, with 1st being the highest quality and 4th being the lowest quality) by region (eastern Upper Peninsula - EUP, western Upper Peninsula - WUP, and northern Lower Peninsula - NLP); for example, 1st EUP indicates highest quality sites in the eastern Upper Peninsula. In the bottom boxes, count values indicate the predicted value (total count of sugar maple stems surveyed per stand) while percentage values indicate the percentage of observations (out of 141) in the node.



B.

C.

We used AIC comparison (Akaike, 1974) to assess if interaction terms improved a simple model of regeneration predicted by deer use and site quality (glm.nb from MASS package, (Venables & Ripley, 2002). AIC comparison indicated that an interaction term would benefit the large saplings model ( $\Delta$ AIC -2, no significant interaction terms), but not for small saplings or seedlings, so we chose not to include the interaction term for any model (data not shown).

#### C.2 Model selection

We compared five candidate models for each sugar maple response size class (Table C.3). The models included a null model, a model with only linear terms, a model with a quadratic terms ( $^2$ ) for each continuous predictor, a model with quadratic terms on all continuous predictors which may limit light (all continuous other than deer use and shrub coverage), and a model with quadratic terms based on prior AIC exploration of the data. For the model informed by AIC exploration, we compared the fit of linear vs. quadratic models (which include a linear and second-order polynomial component) for each variable individually with the response, at the stand-level, using glm.nb from the MASS package; we included quadratic terms in the candidate model based on R<sup>2</sup> comparison. For each of the candidate models for each size class, we calculated the DIC value based on 10,000 post burn-in samples from three MCMC chains using the RJAGS package (Plummer, 2018).

Seedlings					
	Candidate Model	Model hypothesis	DIC		
1	SM1 ~ a; a ~ 1	Null model	22228		
2	SM1 ~ a + tot2 + tot3 + sap + wc; a ~ SM10BA + standBA + deer + Nut1 + Nut3 + Nut4	Linear effects only	22063		
3	$SM1 \sim a + tot2 + tot2^{2} + tot3 + tot3^{2} + sap + sap^{2} + wc + wc^{2};$ $a \sim SM10BA + SM10BA^{2} + standBA + standBA^{2} + deer + deer^{2} + Nut1 + Nut3 + Nut4$	Full quadratic effects	22039		
4	$SM1 \sim a + tot2 + tot2^{2} + tot3 + tot3^{2} + sap + sap^{2} + wc;$ a ~ SM10BA + SM10BA^2 + standBA + standBA^2 + deer + Nut1 + Nut3 + Nut4	Quadratic terms for light limitation	22030		
5	$SM1 \sim a + tot2 + tot2^{2} + tot3 + tot3^{2} + sap + sap^{2} + wc + wc^{2};$ a ~ SM10BA + SM10BA^{2} + standBA + deer + deer^{2} + Nut1 + Nut3 + Nut4	Quadratic terms based on prior AIC exploration	22036		

Table C.3. Candidate models for sugar maple seedlings (SM1), small saplings (SM2), and large saplings (SM3) predicted by plot level densities of total small sapling stems (tot2), total large sapling stems (tot3), stems 5-10 cm DBH (sap), and shrub coverage (wc), plus the stand-level measurements of basal area of sugar maple stems > 25 cm DBH (SM10BA), total stand basal area > 10 cm (standBA), deer use (deer), and site quality (Nut1, Nut3, Nut4) which generate a stand-level intercept, a. The DIC values are based on 10,000 post burn-in samples from three MCMC chains using the RJAGS package (Plummer, 2018). For each size class, the model with the lowest DIC value is bolded.

Table C.3 [cont'd]

	Candidate Model	Model hypothesis	DIC
1	SM2 ~ a;	Null model	10620
	a ~ 1		
2	$SM2 \sim a + tot3 + sap + wc;$	Linear effects only	10590
	a ~ SM10BA + standBA + deer + Nut1 + Nut3 + Nut4		
3	$SM2 \sim a + tot3 + tot3^2 + sap + sap^2 + wc + wc^2;$	Full quadratic effects	10436
	$a \sim SM10BA + SM10BA^{2} + standBA + standBA^{2} + deer + deer^{2} + Nut1 + Nut3 + Nut4$		
4	$SM2 \sim a + tot3 + tot3^2 + sap + sap^2 + wc;$	Quadratic terms for light	10175
	a ~ SM10BA + SM10BA^2 + standBA + standBA^2 + deer + Nut1 + Nut3 + Nut4	limitation	
5	$SM1 \sim a + tot3 + sap + wc;$	Quadratic terms based on	10486
	a ~ SM10BA + SM10BA^2 + standBA + deer + Nut1 + Nut3 + Nut4	prior AIC exploration	

C	~~~	1
Smail	sap	ungs

# Large Saplings

	Candidate Model	Model hypothesis	DIC
1	SM3 ~ a; a ~ 1	Null model	8676
2	SM3 ~ a + sap + wc; a ~ SM10BA + standBA + deer + Nut1 + Nut3 + Nut4	Linear effects only	8601
3	$SM3 \sim a + sap + sap^{2} + wc + wc^{2};$ a ~ SM10BA + SM10BA^2 + standBA + standBA^2 + deer + deer^2 + Nut1 + Nut3 + Nut4	Full quadratic effects	8589
4	SM3 ~ a + sap + sap^2 + wc; a ~ SM10BA + SM10BA^2 + standBA + standBA^2 + deer + Nut1 + Nut3 + Nut4	Quadratic terms only on light limiting variables	8575
5	SM1 ~ a + sap + wc + wc^2; a ~ SM10BA + standBA + deer + Nut1 + Nut3 + Nut4	Quadratic terms based on prior AIC exploration	8611

## APPENDIX D

Details on species diversity analysis

Site quality	WUP	EUP	NLP
1	AArAst, AArLy, PArVAa[w] (n=3)	ATFD (n=10)	PArVVb (n=2)
2	ATM, ATM-Sm, ATM-O, ATFAs (n=4)	AFPo (n=5)	-
3	ATD, ATD-Hp, ADT-Ca, ATD-AVb (n=6)	AFOAs (n=3)	AFO (n=8)
4	AOCa (n=4)	-	AFOCa (n=6)

Table D.1. Summary of stand (n = 51) classification under an herbaceous-indicator site-quality system (Burger & Kotar, 2003). Site quality 1 is lowest, 4 is highest. Letters represent key indicator species: A = Acer saccharum, Aa = Aralia nudicaulis, Ar = Acer rubrum, As = Arisaema atrorubens, Ast = Aster marophyllus, Ca = Caulophyllum thalictroides, D = Dryopteris spinulosa, F = Fagus grandifolia, Hp = Hepatica [variant], Ly = Lycopodium annotinum, M = Maianthemum canadense, O = Osmorhiza claytoni [variant], P = Pinus strobus, Po = Polygonatum pubescens, Sm = Smilacina racemose [variant], T = Tsuga canadensis, V = Vaccinium angustifolium, Vb = Viburnum acerifolium, and [w] = Wisconsin variant.

Scientific name	Common name	Plot occurrences
Abies balsamea	Balsam fir	13
Acer rubrum	Red maple	25
Acer saccharum	Sugar maple	51
Betula alleghaniensis	Yellow birch	13
Betula papyrifera	Paper birch	2
Fagus grandifolia	American beech	26
Fraxinus americana	White ash	5
Fraxinus pennsylvanica	Green ash	2
Juniperus virginiana	Eastern red cedar	1
Ostrya virginiana	Ironwood	10
Picea glauca	White spruce	2
Pinus strobus	White pine	1
Populus grandidentata	Bit-tooth aspen	2
Populus tremuloides	Quaking aspen	2
Prunus serotina	Black cherry	10
Prunus spp.	Cherry species	2
Quercus rubra	Northern red oak	4
Tilia americana	Basswood	25
Tsuga canadensis	Eastern hemlock	2
Ulmus americana	American elm	1

Table D.2. Summary of tree species (scientific and common name plus number of plot occurrences (out of 51)) from plot-level data, surveyed in 51 managed northern hardwood forests.

Stand	Linear AIC	CR AIC	Pseudo R <sup>2</sup>	Stand	Linear AIC	CR AIC	Pseudo R <sup>2</sup>
EUP1	196.9	187	0.806	NLP27	320.4	322.7	0.267
EUP2	228.5	226.7	0.417	NLP28	242.3	205.2	0.91
EUP3	236.7	239.8	0.332	NLP29	261.3	259.7	0.541
EUP4	210.4	207.7	0.429	NLP30	157.9	156.3	0.461
EUP5	233.7	232.7	0.482	NLP31	223.7	202.4	0.838
EUP6	267.3	265.8	0.49	NLP32	218.4	215.1	0.384
EUP7	218.6	204	0.645	NLP33	129.4	131.3	0.00101
EUP8	195.9	193.5	0.808	NLP34	243.8	232.6	0.687
EUP9	235.6	233.4	0.647	WUP35	345.5	348.1	0.345
EUP10	219.8	205.4	0.75	WUP36	236.3	235.8	0.631
EUP11	253.3	246.8	0.662	WUP37	327.5	328.5	0.658
EUP12	251.4	255	0.691	WUP38	304.5	306.9	0.624
EUP13	200.9	191.7	0.789	WUP39	247.2	234.9	0.744
EUP14	254	256.6	0.426	WUP40	245.5	235.6	0.642
EUP15	196.9	195.6	0.404	WUP41	195.1	167.7	0.861
EUP16	366.4	369.4	0.512	WUP42	267.6	266.7	0.427
EUP17	242.9	240.5	0.594	WUP43	225.7	209.2	0.698
EUP18	278	278.8	0.824	WUP44	176	154.4	0.881
NLP19	177.8	177.6	0.512	WUP45	271.4	272.8	0.454
NLP20	212.6	213.9	0.4	WUP46	289.2	288.8	0.224
NLP21	215.7	210.8	0.494	WUP47	251.8	222.8	0.785
NLP22	111.2	112.6	0.166	WUP48	246.5	210.8	0.922
NLP23	195.5	181.5	0.824	WUP49	250.3	245	0.671
NLP24	251.6	242.2	0.636	WUP50	229.5	229.7	0.182
NLP25	238.5	233.9	0.312	WUP51	303.4	307.6	0.738
NLP26	233.4	225.7	0.741				
				1			

Table D.3. AIC comparison between linear and Chapman-Richards (CR) models, by stand, plus the pseudo-R2 value of the selected model (lower AIC, selected model AIC is bolded). Stands are identified by region (EUP = eastern Upper Peninsula, WUP = western Upper Peninsula, and NLP = northern Lower Peninsula).

		Asym			b			c	
Stand	Ε	SE	p-val	Ε	SE	p-val	Ε	SE	p-val
EUP1	105.97	5.73	< 0.001	0.1	0.05	0.042	0.96	0.49	0.061
EUP2	104.14	7.33	< 0.001	0.18	0.13	0.182	1.66	1.93	0.400
EUP4	95.77	7.8	< 0.001	0.08	0.08	0.357	0.43	0.39	0.273
EUP5	107.11	10.1	< 0.001	0.09	0.07	0.205	1.12	1.09	0.310
EUP6	112.42	16.92	< 0.001	0.07	0.07	0.330	0.67	0.51	0.200
EUP7	93.3	2.36	< 0.001	0.24	0.09	0.016	1.58	1.12	0.171
EUP8	157.08	56.61	0.012	0.03	0.03	0.376	0.77	0.31	0.023
EUP9	138.74	26.38	< 0.001	0.05	0.05	0.312	0.72	0.41	0.094
EUP10	105.57	4.19	< 0.001	0.16	0.05	0.002	3.2	1.82	0.092
EUP11	128.29	7.36	< 0.001	0.1	0.05	0.042	1.13	0.63	0.086
EUP13	106.76	4.61	< 0.001	0.09	0.04	0.033	0.6	0.25	0.024
EUP15	125.77	4.2	< 0.001	0.08	0.05	0.145	0.53	0.47	0.271
EUP17	140.31	18.31	< 0.001	0.05	0.05	0.304	0.55	0.36	0.134
NLP19	128.67	20.56	< 0.001	0.07	0.07	0.317	1.02	1.02	0.332
NLP21	97.84	3.79	< 0.001	0.14	0.1	0.144	0.64	0.55	0.257
NLP23	110.73	3.77	< 0.001	0.09	0.04	0.044	0.46	0.18	0.018
NLP24	98.09	3.65	< 0.001	0.11	0.05	0.057	0.87	0.55	0.125
NLP25	102.25	3.08	< 0.001	0.24	0.2	0.246	1.63	3.08	0.601
NLP26	116.38	4.79	< 0.001	0.13	0.06	0.028	1.08	0.57	0.068
NLP28	103.44	2.29	< 0.001	0.14	0.02	< 0.001	2.09	0.55	< 0.001
NLP29	102.31	12.97	< 0.001	0.06	0.07	0.387	0.46	0.32	0.164
NLP30	106.31	6.38	< 0.001	0.06	0.07	0.387	0.3	0.28	0.301
NLP31	100.03	2.99	< 0.001	0.08	0.03	0.004	0.67	0.22	0.005
NLP32	101.94	2.9	< 0.001	0.25	0.14	0.097	1.81	2.44	0.464
NLP34	108.04	4.43	< 0.001	0.14	0.05	0.007	1.34	0.69	0.061

Table D.4. Model parameter estimates, including mean (E), standard error (SE), and p-value (p-val), for AIC-selected Chapman Richards models. Model estimates significantly different from zero (p < 0.05) are bolded. The model equation for age of stem *i* as a function of DBH is:  $Age_i \sim Asym * (1 - e^{-b*DBH_i})^c$ , where the variable Asym is interpretable as the age at which model estimates plateau. Stands are identified by region (EUP = eastern Upper Peninsula, WUP = western Upper Peninsula, and NLP = northern Lower Peninsula).

		Asym			b			С	
Stand	Ε	SE	p-val	Ε	SE	p-val	Ε	SE	p-val
WUP36	96.1	26.12	< 0.001	0.04	0.06	0.563	0.42	0.28	0.153
WUP39	96.05	4.44	< 0.001	0.1	0.04	0.027	0.81	0.37	0.036
WUP40	96.67	2.85	< 0.001	0.15	0.07	0.036	1.07	0.75	0.162
WUP41	99.22	1.1	< 0.001	0.32	0.07	< 0.001	4.41	2.74	0.121
WUP42	118.25	20.4	< 0.001	0.1	0.08	0.220	2.33	2.52	0.365
WUP43	117.38	1.57	< 0.001	0.4	0.16	0.017	15.54	24.36	0.529
WUP44	118.19	2.74	< 0.001	0.12	0.03	0.001	1.34	0.48	0.011
WUP46	131.65	20.31	< 0.001	0.09	0.11	0.428	0.9	1.2	0.459
WUP47	107.25	1.45	< 0.001	0.22	0.06	< 0.001	1.35	0.63	0.040
WUP48	107.72	3.44	< 0.001	0.13	0.03	< 0.001	1.93	0.49	< 0.001
WUP49	103.71	6.82	< 0.001	0.09	0.06	0.146	0.6	0.35	0.097

Table D.4 [cont'd]

	]	Intercept				Slop	e
Stand	Ε	SE	p-val		E	SE	p-val
EUP3	36.3	7.97	< 0.001	1	.44	0.4	0.001
EUP12	35.83	7.01	< 0.001	1	.92	0.24	< 0.001
EUP14	33.29	12	0.01	2	.42	0.56	< 0.001
EUP16	24.49	13.76	0.084	2	.91	0.48	< 0.001
EUP18	34.39	6.92	< 0.001	2	.97	0.25	< 0.001
NLP20	80.47	5.17	< 0.001	0	.58	0.14	< 0.001
NLP22	117.6	6.26	< 0.001	0	.31	0.17	0.083
NLP27	66.94	11.63	< 0.001	1	.53	0.45	0.002
NLP33	97.99	4.36	< 0.001	0	.02	0.14	0.891
WUP35	52.03	8.53	< 0.001	1	.63	0.37	< 0.001
WUP37	36.74	12.82	0.007	2	1.3	0.56	< 0.001
WUP38	30.5	10.77	0.008	2	.58	0.37	< 0.001
WUP45	56.18	7.8	< 0.001	1	.41	0.29	< 0.001
WUP50	95.88	3.63	< 0.001	(	).4	0.15	0.015
WUP51	31.2	9.64	0.003	3	.01	0.33	< 0.001

Table D.5. Model parameter estimates, including mean (E), standard error (SE), and p-value (p-val) for AIC-selected linear models. Model estimates significantly different from zero (p < 0.05) are bolded. The model (which is unique for each stand) for age of stem *i* a function of DBH is:  $Age_i \sim Intercept + Slope * (DBH_i)$ . Stands are uniquely numbered and identified by region (EUP = eastern Upper Peninsula, WUP = western Upper Peninsula, and NLP = northern Lower Peninsula).

Variable	Estimate	Std. Error	T value	P value
Intercept	80.2	2.5	31.5	< 2e-16
CPIntermediate	14.0	2.8	5.0	6.7e-07
CPCodominant	25.0	2.4	10.6	< 2e-16
CPDominant	42.7	2.9	14.7	< 2e-16

Table D.6. Model summaries of fixed effects of crown class as a factor predicting sugar maple age (overtopped included in the intercept, intermediate  $CP_{Intermediate}$ , codominant  $CP_{Codominant}$ , and dominant  $CP_{Dominant}$  stems). Significant predictor estimates (p < 0.05) are bolded.

Variable	Estimate	Std. Error	T value	P value
Intercept	74.64	6.00	12.5	< 2e-16
Crown percentage	-0.31	0.13	-2.4	0.0183
DBH <sub>cm</sub>	0.97	0.21	4.6	5.48e-06
Interaction term	0.01	0.005	2.4	0.0190

Table D.7. Model summary of the fixed effect variables crown percentage, DBH (cm), and their interaction as predictors of sugar maple age. Significant predictor estimates (p < 0.05) are bolded.

Variable	Estimate	Std. Error	T value	P value
Intercept	55.7	3.79	14.73	< 2e-16
Height (m)	2.2	0.23	9.54	< 2e-16

Table D.8. Model summary of a fixed effects of sugar maple stem age as a function of height. Significant predictor estimates (p < 0.05) are bolded.



Figure D.1. Individual tree disc ages (n = 1499) by competitive class (O = overtopped, I = intermediate, C = codominant, and D=dominant).



Figure D.2. Map of model-averaged sugar maple stem age at 10 cm DBH for 51 managed northern hardwood forests.



Figure D.3. Maps of average sample age for sugar maple stems in 51 managed northern hardwood forests in three size classes: saplings (5 – 11.4 cm DBH), poletimber (11.4 – 23.9 cm DBH), and sawtimber (> 23.9 cm DBH).

# APPENDIX E

Model selection for stand-level tree species abundance

Species	Size class	NB	ZINB
AB	1	1193.660342	2225.652189
AB	2	1598.378344	2109.714133
AE	1	4.44771E+12	54769793063
AE	2	4.34462E+11	4027.4578
BF	1	17662.55378	5859.864774
BF	2	6956.414702	1540.43038
BW	1	1164.105206	1335.87357
BW	2	8948264249	8203.240323
CH	1	1379.812449	4515.791002
CH	2	1004.47792	2404.543577
IW	1	1305.503868	3883.088101
IW	2	1638.948521	3830.793844
RM	1	1829.801164	7294.77764
RM	2	1152.69937	4105.586483
RO	1	7.17E+25	562.488586
RO	2	2.48E+48	1.12107E+12
SM	1	2457.502827	6065.31646
SM	2	1912.313374	2517.543047
WA	1	43203411.13	3863.874594
WA	2	1.11E+20	6661.727899
YB	1	2455971.93	9078.715043
YB	2	162518909.4	6156.992703

DIC

Table E.1. DIC comparison for negative binomial (NB) and zero-inflated negative binomial (ZINB) models. Lower DIC value is bolded. Species codes can be found in Table 4.1. Size class 1 refers to seedlings (0 – 137 cm tall) and size class 2 refers to saplings (> 137 cm tall and < 10 cm DBH).

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