Physiological responses to light explain competition and facilitation in a tree diversity experiment

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Abstract

- Ecologists often invoke interspecific facilitation to help explain positive biodiversity-ecosystem function relationships in plant communities, but seldom test how it occurs. One mechanism through which one species may facilitate another is by ameliorating abiotic stress. Physiological experiments show that a chronic excess of light can cause stress that depresses carbon assimilation. If shading by a plant's neighbors reduces light stress enough, it may facilitate that plant's growth. If light is instead most often a limiting factor for photosynthesis, shading may have an adverse, competitive effect.

- In a temperate tree diversity experiment, we measured growth rates and photosynthetic physiology in broadleaf tree species across a gradient of light availability imposed by their neighbors. At the extremes, trees experienced nearly full sun (monoculture), or were shaded by nearby fast-growing conifers (shaded biculture).

- Most species had slower growth rates with larger neighbors, implying a net competitive effect. On the other hand, the two most shade-tolerant species (*Tilia americana* and *Acer negundo*) had

faster growth rates with larger neighbors, implying a net facilitative effect. Compared to the

others, these two species had larger increases in photoinhibition (reduced dark-acclimated F_v/F_m)

across the gradient of increasing light availability, which suggests that they are more vulnerable

to chronic light stress. While most species had lower carbon assimilation rates in the shaded

biculture treatment, *T. americana* had rates up to 25% higher. *T. americana* also dropped its

leaves 3-4 weeks later in shaded bicultures, extending its growing season. We conclude that

although large neighbors can cause light limitation in shade-intolerant species, they can also

increase growth through abiotic stress amelioration in shade-tolerant species.

• Synthesis: Both positive and negative species interactions in our experiment can be explained by

the photosynthetic responses of trees to the light environment created by their neighbors. We

show that physiological measurements can help explain the species interactions that underlie

biodiversity-ecosystem function relationships. The insights that ecologists gain by searching for

physiological mechanisms may help us forecast species interactions under environmental change.

Keywords: Biodiversity and ecosystem function; facilitation; light competition; photoinhibition;

photoprotection; photosynthetic light-response curves; shade tolerance; species interactions; tree diversity

Introduction

The sun's light powers photosynthesis—the chain of reactions through which plants turn carbon dioxide into the living matter they use to grow and reproduce. A common view in plant ecology is that light is often a limiting resource, such that competition for scarce light can govern the fate of plant species (Braun-Blanquet 1932; Canham et al. 1994; Monsi & Saeki 2005; Dybzinski & Tilman 2007; Hautier et al. 2009). But physiologists have also shown that plants often absorb light beyond their capacity to use it for photosynthesis. Beyond an initial linear rise with increasing light, photosynthesis begins to saturate, such that additional light contributes little to the plant's carbon gain. Any factor that limits photosynthesis can exacerbate the excess of light, including photosynthetic downregulation and environmental stresses like water limitation, cold temperatures, or nutrient-poor conditions (Fig. 1; Long et al. 1994; Demmig-Adams & Adams 1992).

Light is thus both an essential resource and a potential stressor. A chronic excess of light can cause plant cells lasting oxidative damage, especially to Photosystem II (PSII). Such damage reduces the efficiency of photosynthesis and can be costly to repair, but plants can avoid it using mechanisms of photoprotection (Murchie & Niyogi 2011). These mechanisms include biochemical pathways that safely dissipate excess light as heat—most notably, the xanthophyll cycle. They also include structural means to avoid absorbing too much light, such as self-shading canopies, reflective leaves, or steep leaf angles (Lovelock & Clough 1992; Streb et al. 1997; Pearcy et al. 2005; Kothari et al. 2018). In general, plants use such photoprotective mechanisms more under environmental conditions that put them at high risk of damage, including high light (Montgomery et al. 2008) and cold or dry climate (Cavender-Bares 2007; Savage et al. 2009; Wujeska et al. 2013; Ramírez-Valiente et al. 2013). Under such conditions, plants need strong photoprotection to prevent even more costly damage.

Depending on how well-protected a plant is, prolonged exposure to excess light can still cause enough damage to reduce carbon assimilation rates (Egerton et al. 2000; Howell et al. 2002; Murchie & Niyogi 2011). Consistent with Long et al. (1994), we use the term 'photoinhibition' to describe a drop in carbon assimilation caused by either photodamage or sustained biochemical photoprotection. Bright

sunlight causes photoinhibition directly by exposing leaves to more radiation, but it also acts indirectly by

altering other microclimatic conditions like temperature, soil moisture, and vapor pressure deficit (VPD;

Björkman & Powles 1984). For example, high VPD can cause stomata to close, which limits the use of

light for photosynthesis; without an energetic sink, the resulting excess of light can then cause

photoinhibition, which limits photosynthetic capacity even further. Alternately, limitation in water or

other resources may arrest tissue expansion and cause carbon sink limitation (Tardieu et al. 2014); in

response, plants may downregulate photosynthesis and initiate sustained photoprotection (Adams III et al.

2013). Such scenarios may be worsened by belowground competition for resources. Because such stresses

vary along environmental gradients, photoinhibition and photoprotective strategies could affect the

distribution and persistence of plant populations (Külheim et al. 2002).

Given that excess light can cause stress, it may be that one plant species can facilitate another

through shading. Whether a plant receives net costs or benefits from its neighbors' shade depends on the

balance of two effects:

1.   Reduction in light available at any instant, which can increase the frequency and severity of

102   photosynthetic light limitation; and,

2.   Avoidance of chronic light stress, which can increase photosynthesis at any given light level.

Monteith et al. (1977) defined light-use efficiency (LUE) as net photosynthesis divided by the amount of

photosynthetically active radiation (PAR) absorbed. Chronic photoinhibitory stress lowers LUE, such that

plants do less with the amount of light they absorb. Shade helps when, across time, the gain of avoiding

chronic stress is greater than the loss caused by potential light limitation. Changes in light availability

alter the horizontal position along a photosynthetic light-response curve at each moment; changes in LUE

alter the shape of the curve such that plants assimilate more or less carbon at a given position along the

curve (Fig. 1). Both effects must be considered together to test whether shading helps or hurts.

The species interactions that result from shading could help to explain the results of experiments

that manipulate biodiversity, including tree diversity. Experimental and observational research alike has

shown that biodiversity often promotes ecosystem functions such as productivity (Tilman et al. 2014). In

forests, this trend has been shown in both experimental (Potvin & Gotelli 2008; Tobner et al. 2016;

Grossman et al. 2017; Williams et al. 2017; Huang et al. 2018; Zemp et al. 2019) and observational

(Gamfeldt et al. 2013; Liang et al. 2016; Oehri et al. 2017) data, across temperate (Tobner et al. 2016;

Grossman et al. 2017; Williams et al. 2017), subtropical (Huang et al. 2018), and tropical (Potvin &

Gotelli 2008; Zemp et al. 2019) biomes. These positive relationships are robust to variation in climate,

successional stage, and other factors (reviewed in Grossman et al. 2018; Ammer 2018).

One of the potential drivers of positive biodiversity-productivity relationships is interspecific

facilitation (Wright et al. 2017; Barry et al. 2018). In biodiversity experiments, patterns of productivity

within species or functional groups can sometimes serve as evidence for facilitation (Mulder et al. 2001;

Fichtner et al. 2017), but they cannot tell us the physiological mechanisms that cause one species to

facilitate another. Ecologists may need to understand these mechanisms to forecast species interactions

and productivity under environmental change (Wright et al. 2017; Barry et al. 2018).

Here, we tested whether the shade created by tree crowns can facilitate the carbon assimilation

and growth of neighboring trees in the Forests and Biodiversity (FAB) experiment at Cedar Creek

Ecosystem Science Reserve (East Bethel, MN). We asked: Do some species aid others by shielding them

from stress caused by intense sunlight? In particular, we sought to test whether slow-growing, shade-

tolerant broadleaf trees might be facilitated by adjacent fast-growing conifers (Table 1). We addressed

this question by measuring the physiology and woody growth of broadleaf species across plots that vary

in species composition. If certain species do benefit from the shade of their neighbors, we would expect

individuals with larger neighbors to have higher growth and carbon assimilation rates. We also posed the

hypothesis that trees exposed to high light would use structural and biochemical photoprotective

mechanisms more than trees shaded by their neighbors. If true, this finding would imply that trees

growing in high light must invest in photoprotection to avoid damage, reinforcing the claim that light

stress influences the interactions within these communities.

Methods

Experimental design

142     The Forests and Biodiversity (FAB) experiment began in 2013 and comprises 142 4×4 m plots,

each planted with 64 trees in a 0.5×0.5 m grid. The experiment has three blocks, each with 49 plots

arranged in a 7×7 square. Neighboring plots within a block share a boundary. Each plot is planted with

one- or two-year-old seedlings of 1, 2, 5, or 12 species from a pool of 12 species, including four

evergreen conifers with needle-like leaves and eight deciduous angiosperms with broad leaves (Table 1).

Except for five-species plots, each distinct species composition has a replicate in each block. All species

in a plot have nearly equal frequency, and the placement of species within a plot is random. The

experimental design is described in greater detail in Grossman et al. (2017). The mean relative growth

rate in monoculture during the first three years varied dramatically among species, being highest in

conifers and *Betula papyrifera* and lowest in *Acer negundo* (Table 1).

Tree growth

154     We surveyed tree growth in late fall of each year. For all living trees, we measured diameter and

155   height to the tallest leader. We measured basal diameter (5 cm from the ground) for trees less than 1.37 m

156   tall, and diameter at breast height for trees more than 1.37 m tall. For a subset of individuals, we

157   measured both diameters. To predict basal diameter when it was not measured, we used linear

158   relationships predicting basal diameter from the height of the same individual in the same year and the

159   basal diameter in the previous year.

160     We estimated woody stem volume as $V = \pi r^2 h$, where h is height and r is the basal radius. This

equation assumes each tree stem can be approximated as a cylinder, an assumption that has been used and

justified in other tree diversity studies in the absence of system-specific allometric equations (Tobner et

al. 2016; Williams et al. 2017). We then calculated the yearly relative growth rate (RGR) between 2016

and 2018 as $RGR = (\ln V_{2018} - \ln V_{2016})/2$ (Hoffman & Poorter 2002). We chose 2016 as a starting

point because no replanting of dead trees occurred after that point. Biovolume loss could cause RGR to be

negative (Mahmoud & Grime 1974), which often occurred when stems or branches broke, or occasionally

when plants lost all aboveground tissue and resprouted the following year.

For each tree, we also calculated the average stem volume of its eight neighbors (in cardinal and

intercardinal directions) in 2018 as a proxy for the intensity of aboveground interactions. For each

species, we tested whether individual growth increased, decreased, or remained constant with increasing

neighbor size using ordinary least squares (OLS) regression. We also tested whether any species showed

non-monotonic responses to neighbor size using the two-lines test (Simonsohn 2018). Neighbors were

sometimes missing because of mortality or because the focal individual was on the experiment's edge. In

such cases, we assigned the neighbor a volume of zero. We performed all analyses in *R v. 3.6.1* (R Core

Team 2019).

Photosynthetic physiology

We measured chlorophyll fluorescence parameters and leaf reflectance spectra among all eight

broadleaf species, and photosynthetic light-response curves among four focal species (*A. rubrum*, *B.*

papyrifera, *Q. ellipsoidalis*, and *T. americana*). For physiological measurements, we focused on plots

belonging to one of three treatments: (1) relatively open monocultures of broadleaf species

('monoculture'); (2) bicultures comprising one broadleaf and one conifer species ('shaded biculture'); and

(3) twelve-species plots ('twelve-species') (see Fig. S1 for sample images). For each species in each

treatment, we measured physiological parameters on six individuals—two in each of three plots. Neither

A. negundo nor *Q. macrocarpa* was planted with a conifer in any biculture. We used *B. papyrifera* as a

shaded biculture partner for *Q. macrocarpa* because it is a fast-growing species that creates shade. We

omitted the shaded biculture treatment for *A. negundo*. When possible, we chose the same trees within a

plot for each kind of physiological measurement so that we could compare these aspects of physiology

within individuals.

Photosynthetic light-response curves

We measured photosynthetic light-response curves from the four focal species during July 2018

using an LI-6400 gas exchange system (LI-COR BioSciences, Lincoln, NE, USA) with a 6400-40 Leaf

Chamber Fluorometer head. From each tree ($n = 72$ total), we selected a fully expanded upper leaf with

no visible sign of disease or herbivory. We noted the angle of each leaf relative to horizontal (to the

nearest 15˚) before beginning measurements. Each curve had nine steps in descending order of brightness:

2000, 1500, 1000, 500, 200, 100, 50, 20, and 0 µmol m^{-2} s $^{-1}$. We also measured the relative electron

transport rate (ETR) using chlorophyll fluorescence at each light level. We maintained favorable

conditions inside the chamber during each measurement. Following each curve, we removed the leaf to

measure leaf mass per area (LMA) using a balance and a flatbed scanner. Further details about light-

response curve procedures and ETR calculations can be found in the Supplemental Materials (Appendix

1). Finally, we estimated parameters like the light-saturated carbon assimilation rate (A_{sat}) by fitting a

non-rectangular hyperbolic model (Johnson & Thornley 1984) to each light-response curve using R code

written by Nick Tomeo, available at: https://github.com/Tomeopaste/AQ_curves.

We aimed to determine how realized carbon assimilation rates vary with the photosynthetic

photon flux density (PPFD) at the top of the canopy, a measure of light availability in the absence of

shading. While all trees received the same range of light intensities inside the instrument chamber, some

trees received much less light than others *in situ* because their neighbors shaded them more. At any

moment, trees that are more shaded are farther to the left on their light-response curve than trees that are

not. For example, for a given tree crown, the topmost leaves may only receive 50% of the light reaching

the top of the canopy. For this tree crown, the realized assimilation rate we estimate at a top of the canopy

PPFD of 2000 µmol photons m^{-2} s $^{-1}$ would be the assimilation rate at a leaf-level PPFD of 1000 µmol m^{-2}

s $^{-1}$. To visualize the impacts of shading by neighbors, we followed the approach of Howell et al. (2002):

we estimated how much light each leaf receives as a constant fraction of light at the top of the canopy,

and rescaled each leaf's fitted light-response curve along the *x*-axis by this fraction's reciprocal. We

estimated the constant rescaling factor for each individual as $RLA \times cos(\theta)$, where *RLA* is relative light

availability (see *Light availability*) and θ is the leaf angle from horizontal. (Taking the cosine of θ

approximately corrects for the reduction in horizontally projected leaf area in steeply inclined leaves

when the sun is directly overhead, as simplified from Ehleringer & Werk [1986].) We used this procedure

on each fitted curve in order to compare across treatments in a way that accounts for the varying fractions

of light that individuals receive due to differences in relative height and vertical light transmission.

Finally, we used the rescaled light-response curves to estimate carbon assimilation by top leaves

throughout July. We used an hourly averaged time series of solar radiation from Carlos Avery Wildlife

Management Area in Columbus, MN, 13.6 km away from the study site. Following Udo and Aro (1999),

we assumed that each Watt m^{-2} of solar radiation has about 2.08 μmol photons m^{-2} s^{-1} of PAR. By using

these data as inputs to the rescaled light-response curves, we estimated carbon assimilation over the

month of July. This procedure assumes that the photosynthetic response to light remains constant and

unaffected by factors like leaf temperature and stomatal closure. While this assumption is never quite

true, we consider it a useful way of estimating the consequences of realistic fluctuations in light.

We express assimilation per unit of leaf dry mass because we aim to determine the plants' return

on the carbon invested in leaf construction. As expected, leaves growing in low light tended to have lower

LMA in most species (Fig. S2; Poorter et al. 2019; Williams et al. 2020). Expressing data on a mass basis

accounts for these large differences in construction cost, since less carbon is spent in making a leaf with

low LMA. We also express stomatal conductance (g_s) and electron transport rate (ETR) on a mass basis to

compare them with assimilation rates.

Instantaneous chlorophyll fluorescence, pigments, and spectral reflectance

In all eight broadleaf species, we measured chlorophyll fluorescence parameters using an FMS2

pulse-modulated fluorometer (Hansatech Instruments Ltd., Norfolk, UK) over two days in late July ($n =$

138 total). We measured dark- and light-acclimated parameters at the same spot on the same leaves. We

attached opaque clips to leaves in the evening before measuring dark-acclimated F_v/F_m within two hours

after sunrise. This parameter describes the maximum quantum yield of PSII and is a general index of

photoinhibition (Murchie & Lawson 2013). We then took light-acclimated measurements between 12:00

and 14:00 each day. The protocol involved the following steps: actinic light at 1000 µmol m^{-2} s^{-1} for 15

seconds, a saturating pulse, and two seconds of far-red light. We exposed all leaves to the same actinic

light because we aimed to assess photoprotective capacity under comparable light conditions, even though

each tree had a different light environment.

From these data, we estimated qN, a parameter that indicates how much a plant relies on

photoprotective dissipation (non-photochemical quenching) under the imposed actinic light. This

parameter is correlated with the de-epoxidation state of xanthophyll cycle pigments (Watling et al. 1997;

Cavender-Bares & Bazzaz 2004). Following Kramer et al. (2004), we also calculated the quantum yields

ϕ_{PSII}, ϕ_{NPQ}, and ϕ_{NO}, which sum to 1. These three parameters are the proportions of light energy dissipated

by photosynthetic chemistry, non-photochemical quenching, and non-regulated dissipation. The last is a

crucial parameter because it represents light that a plant cannot dissipate safely; this portion of the

absorbed light may contribute to photodamage. The Supplemental Materials contain formulas for all

parameters and justifications for our choices in data analysis (Appendix 1).

On separate top leaves of the same trees, we measured reflectance spectra (350-2500 nm) using a

PSR+ 3500 field spectroradiometer (Spectral Evolution, Lawrence, MA, USA). We used these spectra to

calculate the photochemical reflectance index (PRI), calculated as PRI = $(R_{531} - R_{570}) / (R_{531} + R_{570})$,

where R_n is the reflectance at a wavelength of n nm. PRI shows a negative correlation with carotenoid :

chlorophyll ratios and, over shorter time scales, also tracks the xanthophyll de-epoxidation state (Gamon

et al. 1992; Wong & Gamon 2015; Gitelson et al. 2017). We estimated species-specific linear responses

of PRI and chlorophyll fluorescence parameters to relative light availability (see *Light availability*).

To understand how photoinhibition and photoprotective strategies vary with adaptations to shade,

we used shade tolerance values that Niinemets & Valladares (2006) compiled on a five-point scale.

Higher values along this scale indicate that plants are able to grow in lower light conditions.

Phenology

We monitored the timing of leaf abscission in monoculture and shaded biculture plots for *T.*

americana to assess whether shade could delay senescence and abscission, extending the period for

carbon gain (Cavender-Bares et al. 2000). In early August, we surveyed 120 *T. americana* trees (60 in

monoculture, 60 in shaded biculture) to determine the proportion of leaves that had fallen based on how

many of the top five axillary buds lacked a leaf. From each of these trees, we marked a single leaf and

returned every 1-2 weeks to monitor its senescence. We assessed each leaf on a binary basis (senesced or

not) based on whether it had at least 50% remaining green leaf area. We chose this criterion as a simple

proxy for the continuous scale used by Cavender-Bares et al. (2000). Based on the initial survey and the

repeated surveys of marked leaves, we tracked the proportion of leaves that had senesced over time. To

ensure that leaves were photosynthetically active until near the point of senescence, we collected a one-

time measurement of dark-acclimated F_v/F_m in mid-September among remaining leaves.

We also performed a one-time measurement of 60 (30 per treatment) *A. rubrum* plants on October

1, using the same protocol we used to do our initial early August survey of *T. americana*. We aimed for

this survey to help test whether our results hold across species.

Water potential

Water deficits—even moderate ones—can arrest tissue growth (Tardieu et al. 2014) and reduce

carbon gain by causing stomata to close (Brodribb et al. 2003). To assess how our treatments affected

water status, we measured leaf water potential using a Scholander pressure bomb. We measured pre-dawn

water potential (Ψ_{PD}) in all eight broadleaf species ($n = 96$) and midday water potential (Ψ_{MD}; between

12:00 and 14:00) in only the four focal species ($n = 48$). In these measurements, we included monoculture

and shaded biculture treatments, except in *A. negundo*, where we used the twelve-species treatment in

place of the absent shaded biculture treatment. We removed leaves and immediately placed them in

impermeable plastic bags with wet paper towels. We carried leaves in coolers to be measured indoors

within 90 minutes.

Because we only intended to test whether there was a general tendency across species for water

potential to vary with light, we included species as a random effect and light availability as a fixed effect

in regression models for both Ψ_{PD} and Ψ_{MD}. We calculated marginal R^2 of the fixed effect using the

method of Nakagawa & Schielzeth (2013) as implemented in the *R* package *MuMIn v. 1.43.17* (Bartoń

2020) and *p*-values using Satterthwaite's degrees of freedom method as implemented in the *R* package

lmerTest v. 3.1.2 (Kuznetsova et al. 2017), noting that such values should be interpreted conservatively.

Leaf angles

Leaf angle is a key control on light exposure because more vertical leaves intercept less light

when the sun is directly overhead, and may allow light to penetrate deeper into the canopy (Posada et al.

2009). As a result, steep leaf angles can be a key structural strategy for avoiding light, temperature, and

water stress (Valladares & Pugnaire 1999; Werner et al. 1999; Pastenes et al. 2004; Valiente-Banuet et al.

2010). We predicted that trees in monoculture would have steeper leaf angles than those in other

treatments. From each of three species (*A. rubrum*, *Q. ellipsoidalis*, and *T. americana*), we randomly

selected five individuals from each of six plots—three from monoculture treatments, three from shaded

bicultures (*n* = 15 per species × treatment). On these individuals, we measured the leaf angle from

horizontal of the top five leaves in the crown to the nearest 15˚.

Because angles of leaves within individuals may be non-independent, we tested whether

treatments differ using a mixed-effects model with treatment as a fixed effect and individual tree as a

random effect. Deviations from horizontal in either direction can reduce light interception, so we used the

absolute value of leaf angle as the dependent variable. We again calculated *p*-values using Satterthwaite's

degrees of freedom method.

Environmental data

Light availability

On a cloudy day with diffuse light conditions, we measured the available light above each tree

selected for this study ($n = 138$) using an AccuPAR LP-80 ceptometer (METER Group, Pullman, WA,

USA). We took the mean of two to four PAR measurements in the open and the mean of two

measurements directly above the topmost leaf of each tree. By calculating a ratio of these values, we

could calculate the percent of light transmitted to the top of each tree. We called this value *relative light*

availability (RLA). This value usually correlates well with the percentage of light a tree can access over

much longer time-scales (Parent & Messier 1996).

Soil water content

We collected a one-time measurement of soil volumetric water content in mid-August using a

FieldScout Time-Domain Reflectometer 350 (Spectrum Technologies, Inc., Aurora, IL, USA). This single

measurement followed six days from the last rain. At midday, we collected and averaged three

measurements in each of 32 plots, spanning monocultures of the four core broadleaf species, shaded

bicultures, and twelve-species plots.

Air temperature

For nine days during early September, we set out four Thermochron iButtons (model DS1921G;

Maxim Integrated Products, Sunnyvale, CA, USA) in shaded plots and two in twelve-species plots,

setting them to log air temperature hourly. Each iButton was suspended on mesh inside a 1 m-tall PVC

pipe that was capped with an elbow to shield against solar irradiance. We compared these data to air

temperature measured hourly under open, sunny conditions at the Cedar Creek LTER weather station

about 0.77 km away. We assumed that data from this station would be representative of air temperature in

non-shaded areas. We ignored nighttime and early morning (20:00-08:00 h) readings because iButtons

may have been influenced by evaporative cooling of condensed water droplets.

Results

Tree size and growth

Species varied in their relationship between neighbor stem volume and focal individual relative

growth rate (RGR; ANCOVA, neighbor volume × species interaction; $p < 10^{-15}$; $F_{11,8118} = 19.850$). For

individual species, relationships between RGR and neighbor volume were noisy but often highly

significant (Fig. 2). In most species, individuals with larger neighbors had lower RGR, including in *A.*

rubrum ($R^2 = 0.033$, $p < 10^{-5}$, $t(622) = -4.922$), *Q. alba* ($R^2 = 0.038$, $p < 10^{-6}$, $t(609) = -5.031$), *Q.*

ellipsoidalis ($R^2 = 0.030$, $p < 10^{-5}$, $t(672) = -4.647$), *Q. macrocarpa* ($R^2 = 0.027$, $p < 10^{-5}$, $t(751) = -4.658$),

and *Q. rubra* ($R^2 = 0.056$, $p < 10^{-9}$, $t(672) = -6.411$). But two species grew faster with larger neighbors: *A.*

negundo ($R^2 = 0.018$, $p < 0.005$, $t(384) = 2.857$) and *T. americana* ($R^2 = 0.019$, $p < 10^{-4}$, $t(757) = 3.997$).

B. papyrifera showed no significant relationship with neighbor size ($R^2 = 0.004$; $p = 0.079$, $t(585) = 1.76$).

No species had a non-monotonic response. These results are qualitatively unchanged by removing trees

on a block edge, incorporating plot as a random effect, log-transforming neighbor volume, or including

focal plant volume in 2013 or 2016 as a covariate (Supplemental Materials, Appendix 2).

Although these individual-level effects are noisy, they reveal strong tendencies in average stem

growth across the full range of neighbor size. Imagine two *T. americana* trees that begin at the same size:

One in monoculture, where the average neighbor in 2018 is 573 cm^3, and one growing in a shaded

biculture with *Pinus strobus*, where the average neighbor is 2175 cm^3. The individual-level regression

predicts that after two years, the average *T. americana* in the shaded biculture would be 39.7% larger

because of the difference in mean RGR. Or consider *Q. rubra*, which had the strongest negative response

to neighbor size: We predict that an average tree in the shaded biculture (mean neighbor size 1633 cm^3)

would be 39.1% smaller than in monoculture (mean neighbor size 257 cm^3). These figures assume that

individuals across treatments start at the same size at the initial point for RGR estimates, so they may

underrepresent the true differences in size that have developed and may continue to develop with time.

The absolute growth rate gives some insight into these differences. For example, the median *T. americana*

in monoculture grew 103 cm^3 (25–75th percentile: 21–255 cm^3) between 2017 and 2018, compared to 588

cm^3 (151–1017 cm^3) in shaded biculture. In comparison, the median *Q. rubra* grew 32 cm^3 (25–75th

percentile: 3.9–124 cm^3) in monoculture, compared to only 1.2 cm^3 (-42–80 cm^3) in shaded biculture.

We can also aggregate these observations to the plot scale—for example, by considering the RGR

of summed stem volume of each species in each plot (Fig. 2). (This approach is distinct from an average

of individual RGRs, which assigns equal weight to small and large individuals.) Although aggregation

leaves less power to detect significant relationships, it also reduces noise, such that the relationships we

find explain much more of the variation in growth across plots. As average neighbor size increases, RGR

declines in *Q. alba* ($R^2 = 0.311$, $p < 0.001$, $t(28) = -3.755$) and *Q. macrocarpa* ($R^2 = 0.174$, $p < 0.01$, $t(31)$

$= -2.779$), but increases in *T. americana* ($R^2 = 0.282$, $p < 0.001$, $t(32) = 3.735$).

Trees that died between 2016 and 2018 dropped out of our individual-level analyses of RGR

because their stem volume was 0 in 2018, making it impossible to calculate RGR. To see whether

mortality could alter our conclusions, we also performed statistical analyses on untransformed growth in

stem volume from 2017 to 2018, in which we alternately (1) treated mortality as a decline in volume to 0,

or (2) removed trees that died from analyses. The rate of mortality was low between 2016 and 2018

(~7.5%), and accounting for mortality makes no qualitative difference when considering absolute growth

as the response variable (Fig. S3; Supplemental Materials, Appendix 2).

Photosynthetic physiology

Photosynthetic light-response curves

As a function of the chamber light level, mass-based assimilation rates were always higher in

shaded biculture and twelve-species treatments than in monoculture in three out of the four focal species

(*Q. ellipsoidalis*, *A. rubrum*, and *T. americana*; Fig. 3, left). In the pioneer species *B. papyrifera*,

assimilation rates in shaded biculture were lower than in the other two treatments. Area-based

assimilation rates showed similar trends, except that in *A. rubrum*, they were lower in monoculture than

the other treatments (Fig. S4); *A. rubrum* had particularly large treatment differences in LMA (Fig. S2).

The rate of photosynthesis is often limited either by ETR or by the RuBP carboxylation rate,

which in turn may be limited by stomatal diffusion of CO_2 (Farquhar et al. 1980). In our data, both ETR

and stomatal conductance (g_s) increased with light availability (Fig. S5). (ETR tended to decline at very

high light levels, especially in shaded treatments, perhaps because of acute photoinhibition.) Within each

species, the rank-order of treatments in carbon assimilation rates, ETR, and g_s were broadly congruent.

One exception is that *T. americana* in shaded biculture had much higher g_s than in twelve-species or

monoculture plots, despite having slightly lower assimilation rates than in twelve-species plots.

Compared to the unscaled chamber light-response curves, the picture that emerged from rescaled

light-response curves is more complex (Fig. 3, right). In *B. papyrifera*, the assimilation rate was still

lowest across light levels in the shaded biculture treatment. In late-successional *T. americana*, the

assimilation rate was higher in shaded biculture and twelve-species treatments by up to 25% compared to

monoculture across most light levels. In the other two species, the assimilation rate was highest in

monoculture at low light, but twelve-species and (in *Q. ellipsoidalis*) shaded biculture plots intersected

and surpassed monocultures when enough light was available. *A. rubrum* and *Q. ellipsoidalis* grew less

and had lower relative light availability (RLA) in shaded bicultures than *T. americana* (Fig. S2), so their

assimilation rates dropped considerably more in the rescaled light-response curves.

Using a time series of solar radiation, we estimated total assimilation rates in July 2018 based on

each rescaled light-response curve (Fig. S6). We found that trees would have lower total assimilation in

shaded biculture than in monoculture in *A. rubrum* (ANOVA; $p = 0.009$; $F_{2,14} = 6.759$) and *B. papyrifera*

($p = 0.002$; $F_{2,15} = 10.27$). In both species, Tukey's HSD test showed that trees in twelve-species plots did

not significantly differ from those in monoculture. There were no significant differences at all among

treatments in *Q. ellipsoidalis* ($p = 0.294$; $F_{2,15} = 1.330$) or *T. americana* ($p = 0.359$; $F_{2,15} = 1.099$).

Instantaneous chlorophyll fluorescence, pigments, and spectral reflectance

Among all eight species but early-successional *B. papyrifera*, dark-acclimated F_v/F_m declined as

RLA increased (Fig. 4). Among the four focal species, dark-acclimated F_v/F_m was positively correlated

with mass-based A_{sat}, the light-saturated assimilation rate (OLS regression; $R^2 = 0.493$; $p < 10^{-11}$; $t(69) =$

8.304; Fig. 5); slopes were not significantly different among species (ANCOVA, species × dark-

acclimated F_v/F_m interaction; $p = 0.658$; $F_{3,63} = 0.539$). The same overall relationship holds when A_{sat} is

expressed on an area basis ($R^2 = 0.447$; $p < 10^{-9}$; $t(69) = 7.588$).

In most species, qN rose with RLA, which shows that trees that grow in high light have greater

capacity to use non-photochemical quenching to dissipate excess light (Fig. 4). Two species were clear

exceptions: In *T. americana* and *A. negundo*, qN was nearly constant across light environments.

Extracting the species-specific slopes of dark-acclimated F_v/F_m against RLA, we found that the species

with the greatest decline in F_v/F_m are the most shade-tolerant (Fig. 5; $R^2 = 0.601$; $p = 0.025$; $t(5) = -3.167$).

Likewise, qN increases less with RLA among shade-tolerant species ($R^2 = 0.619$; $p = 0.022$; $t(5) = -$

3.276).

To illustrate why shade-tolerant species show such large increases in photoinhibition with light,

consider ϕ_{PSII}, ϕ_{NPQ}, and ϕ_{NO}—the quantum yields of PSII photochemistry, non-photochemical quenching,

and non-regulated dissipation (Fig. S6). We set aside *B. papyrifera*, which grows fast enough that it is

seldom truly shaded. In most other species, ϕ_{NO} decreases with light in the growth environment (Fig. 4).

Trees growing in the open can either use light at 1000 μmol photons m^{-2} s^{-1} for photochemistry or

dissipate much of the remainder through non-photochemical quenching, keeping ϕ_{NO} close to the oft-cited

theoretical minimum of ~0.17 (Tietz et al. 2017). But in *T. americana* and *A. negundo*, ϕ_{NO} is nearly equal

across the gradient of light availability.

PRI declined with increasing light availability in all species, which supports the hypothesis that

trees in monoculture allocate more to carotenoids (Fig. 4). Although this relationship had similar slopes

across all species, *B. papyrifera* had higher PRI at all light levels within its narrow domain.

Phenology

In *T. americana*, leaf abscission in monoculture began in early August, and more than half of all

leaves had senesced by September 3 (Fig. 6). In shaded bicultures, more than 90% of leaves remained by

September 3, and no survey found greater than 50% senescence until October 1. We assigned the

senescence date of each leaf as the date of the first survey by which it had senesced; using this response

variable, leaves in monoculture senesced 22 days earlier than those in shaded cultures (t-test; $p < 10^{-11}$;

$t(116) = 7.822$). These results suggest that leaves in shaded bicultures had a longer period for carbon gain.

Here, we assume that leaves in shaded biculture did not downregulate photosynthesis any longer before

abscission than those in monoculture. This assumption was supported by mid-September measurements

of dark-acclimated F_v/F_m. At that point, dark-acclimated F_v/F_m averaged 0.745 among remaining leaves in

shaded biculture, but only 0.642 in monoculture (t-test; $p = 0.061$; $t(33) = 1.936$).

Our one-time measurement of *A. rubrum* leaves in early October confirmed that this pattern is not

limited to *T. americana*. *A. rubrum* trees in shaded biculture plots retained about 56% of their leaves,

while those in monoculture retained only 10% (t-test; $p < 10^{-4}$; $t(44) = -4.486$).

Water potential

We removed two data points, one in each treatment, with Ψ_{PD} more negative than -0.8 MPa on the

grounds that they may have been subject to measurement error. Across treatments, the mean Ψ_{PD} and Ψ_{MD}

were -0.15 and -1.29 MPa. Ψ_{PD} increased with light availability ($p < 10^{-5}$; $R_m^2 = 0.272$; $t(70.0) = -5.152$),

making it less negative in monocultures than shaded bicultures by an average of 0.07 MPa (Fig. S8). This

result holds even including the outliers ($p = 0.014$; $R_m^2 = 0.085$; $t(54.3) = -2.555$). By midday, the pattern

reversed; Ψ_{MD} declined with light availability ($p < 10^{-4}$; $R_m^2 = 0.346$; $t(36.6) = 4.450$), making it more

negative in monocultures than shaded bicultures by an average of 0.33 MPa.

Leaf angles

Among each of the three species whose leaf angles we measured, leaves were nearer to horizontal

in the shaded biculture treatment (Fig. 7). The size of this fixed effect varied among species: 11.0° in *Q.*

ellipsoidalis ($p = 0.005$; $t(34.2) = -3.01$), 21.2° in *A. rubrum* ($p < 10^{-9}$; $t(108.4) = -6.807$), and 14.3° in *T.*

americana ($p = 0.003$; $t(124.5) = -3.040$). Because *T. americana* leaves usually drooped downward in

monoculture, leaf angles became less negative in the shade. In *Q. ellipsoidalis*, whose leaves were often

inclined upward in monoculture, leaf angles became less positive in the shade. In *A. rubrum*, leaf angles

were inclined both up and down in monoculture, but varied less in either direction in the shade.

Environmental factors

In our survey, soil moisture varied among treatments (ANOVA; $p < 0.01$; $F_{2,29} = 6.115$); it was

lowest (5.7%) in shaded biculture plots and highest (7.2%) in broadleaf monoculture plots. We found that

daytime air temperature in the open was higher than in shaded bicultures or twelve-species plots at 62.5%

of hourly log-points during the nine-day logging period. The air temperature in monocultures averaged

1.03° C higher over this period.

Discussion

We investigated how species interactions in a tree diversity experiment might emerge from the

physiological responses of trees to the light environment created by their neighbors. We found that two

species (*A. negundo*, *T. americana*) grew faster on average with larger neighbors, while five species (*A.*

rubrum, *Q. alba*, *Q. ellipsoidalis*, *Q. macrocarpa*, and *Q. rubra*) grew slower. The two species with

positive responses to neighbor size were shade-tolerant and susceptible to photoinhibition in high light,

which could reduce carbon assimilation in monoculture. The five species with negative responses were

shade-intolerant, resistant to photoinhibition, and tended to have lower carbon assimilation in shaded

bicultures. The divergent responses of these two groups of species appear to be driven in part by their

tolerance to excess light. Here, we interpret these patterns further.

Competition and shade-intolerant species

Individuals of all four *Quercus* species and *A. rubrum* grew less when surrounded by larger

neighbors. This growth pattern implies that neighborhood interactions were dominated by competition,

potentially both above- and belowground. (The fact that soil moisture was lowest in our dense-growing

shaded biculture plots suggests an important role for belowground competition.) The pioneer species *B.*

papyrifera, whose leaves had lower carbon assimilation rates in the shaded biculture, had a neutral

response to neighbor size, perhaps because its rapid growth allowed it to overtop its neighbors.

The physiological data can help us make sense of these growth patterns. We measured light-

response curves of three species in this group. In the untransformed chamber light-response curves, *B.*

papyrifera had lower assimilation rates in shaded biculture than in monoculture or twelve-species plots,

while *A. rubrum* and *Q. ellipsoidalis* had lower assimilation rates in monoculture than in the other two

treatments. In all species except *B. papyrifera*, dark-acclimated F_v/F_m declined as light availability rose,

confirming that high light exposure causes photoinhibition. Because dark-acclimated F_v/F_m correlates

with A_{sat}, this trend implies that *A. rubrum* and *Q. ellipsoidalis* had lower A_{sat} in monoculture due to

photoinhibition. Since photoinhibition occurs through damage or downregulation of PSII, it is usually

thought to depress photosynthesis by lowering ETR. But both ETR and stomatal conductance mirrored

assimilation across treatments in chamber light-response curves (Fig. S5), so both may have some role in

explaining why most species had lower assimilation in monoculture at a given chamber light level.

Rescaling the light-response curves allowed us to visualize the costs of shading—the fact that

some trees receive less light than others do, and assimilate less carbon as a result. *B. papyrifera* was little

affected by rescaling. Compared to monoculture, *A. rubrum* and *Q. ellipsoidalis* in shaded biculture had

lower carbon assimilation under low light at the top of the canopy, but higher carbon assimilation under

high light. For these two species, shading often appears severe enough to cause light limitation of growth

and photosynthesis, especially under otherwise low light.

Given that light availability fluctuates—diurnally, seasonally, and with cloud cover—the

treatment that performs best may also vary from moment to moment. We used a time-series of solar

radiation to estimate total carbon assimilation throughout July, assuming that each tree's photosynthetic

light-response curve does not vary through the month. We found that both *A. rubrum* and *B. papyrifera*

had lower carbon assimilation in shaded biculture than in the other treatments. PPFD was under 500 µmol

m^{-2} s^{-1} more than 60% of the time during the month; it is under such dim conditions that trees in

monoculture most outperform those in shaded biculture, no matter what benefits shading may confer at

the sunniest of times. Trees in twelve-species plots had similar assimilation rates to those in monoculture

in all species. The intermediate light environment in these twelve-species plots may have allowed trees to

avoid photoinhibition without becoming frequently light-limited (Fig. S2), although we caution that we

only measured top leaves, so we cannot directly compare whole-plant carbon gain.

Trees in FAB and elsewhere often vary in size by orders of magnitude, which foregrounds the

size-asymmetry of light competition. Research on light competition in biodiversity-ecosystem function

research underscores a tension between competitive imbalance caused by size-asymmetry and

competitive relaxation caused by light-use complementarity, in which species may partition their

exploitation of the light environment (Yachi & Loreau 2007; Sapijanskas et al. 2014; Williams et al.

2017). We leave the potential for light-use complementarity unaddressed, but we show that competitive

imbalance can suppress the photosynthesis and growth of species that are poorly adapted to shade. Such

imbalances could promote selection effects (*sensu* Loreau & Hector 2001), in which the most productive

species become even more dominant in mixtures.

Facilitation and shade-tolerant species

In *T. americana* and *A. negundo*, the two most shade-tolerant species, having larger neighbors

increases aboveground growth. We interpret this pattern as a result of facilitation, reinforcing prior results

that showed that shade-tolerant species have more facilitative responses to being shaded (Montgomery et

al. 2010). Because we did not measure root growth, we cannot rule out the chance that this trend is caused

by a shade avoidance response that increases allocation to shoots compared to roots (Schall et al. 2012).

Nevertheless, other studies have shown that shade can increase combined root and shoot biomass (e.g.

Ryser & Eek 2000; Semchenko et al. 2012), and in another tree diversity experiment, both root-to-shoot

ratios and the allometric relationship between basal area and biomass were nearly unaffected by the local

neighborhood (Guillemot et al. 2020). These results reinforce the conclusion that *T. americana* and *A.*

negundo have greater aboveground growth due to facilitation, not just changes in biomass allocation.

Here, too, the physiological data can help us explain the growth patterns. *T. americana* was

among the four focal species, and the only one to have lower carbon assimilation rates in monoculture

than in the other treatments across most light levels in both chamber and rescaled light-response curves.

The steep decline in dark-acclimated F_v/F_m with increasing light suggests that low A_{sat} in monoculture

results from photoinhibition. This pattern aligns with *T. americana*'s role as a late-successional dominant

species throughout much of its range (Braun 1950). Our finding that carbon gain in *T. americana* can

increase under shade is reinforced by Carter and Cavaleri (2018), who found that its assimilation rates

may increase down vertical gradients from the upper canopy to the sub-canopy.

In *T. americana*, we found no significant differences among treatments in estimated assimilation

throughout July. These estimates cannot explain certain features in our stem growth data—including, for

instance, the clearly positive effect of neighbor size on growth in *T. americana*. We next discuss two

kinds of phenomena that could contribute to variation in growth rates, but which analyses based on light-

response curves do not account for: Delayed senescence and microclimatic effects. Both may contribute

to potential positive effects of shading by neighbors.

T. americana in monoculture dropped its leaves nearly a month earlier into the fall compared to

the shaded biculture (Fig. 6). We also found that *A. rubrum* in monoculture had earlier leaf senescence.

Given that leaves remained photosynthetically active until shortly before abscission (as in Mattila et al.

2018), shade could allow these trees a longer period of carbon gain. This pattern admits at least two

explanations: (1) that plants in the sun are earlier to accumulate phenological cues for senescence related

to temperature or photoperiod; or (2) that excess light and other stresses cause damage that accelerates

senescence (Cavender-Bares et al. 2000; Brelsford et al. 2019), especially as the temperature declines,

rendering leaves more susceptible to photodamage (Juvany et al. 2013; Renner & Zohner 2019). Contrary

to the first explanation, air temperatures in full sun remained higher than in shaded plots.

Although we emphasize the role of light stress, other microclimatic factors could contribute to

variation in assimilation rates among treatments. For example, an alternate perspective on photoinhibition

is that it mainly results not from photodamage but from carbon sink limitation, which leads plants to

downregulate light capture and electron transport as a photoprotective response (Adams III et al. 2013).

From this perspective, photoinhibition seldom causes carbon limitation or reduces growth rates in nature;

rather, slow growth rates cause photoinhibition. A potential cause of sink limitation that could affect

monocultures of *T. americana* and *A. negundo* more than shaded bicultures is water limitation, which

could physically constrain the growth of tissues (Tardieu et al. 2014). We found that leaf Ψ_{PD} was slightly

more negative in shaded bicultures, while Ψ_{MD} was more negative in monocultures (Fig. S8). (These

results are consistent with the idea that soil moisture is higher in low biomass plots, as we found, while

daytime VPD is lower in high biomass plots, as Wright et al. (2018) found on hot, dry days at Cedar

Creek.) If Ψ_{MD} responds in other tissues as in leaves, highly negative values could arrest growth in

monoculture, causing carbohydrates to accumulate in leaves under high light.

But from this perspective, it is hard to explain why this sink limitation would only matter in *T.*

americana among the four focal species, since it is not the only one to have more negative Ψ_{MD} in

monoculture. Moreover, trends in ϕ_{NO} suggest that *T. americana* and *A. negundo* in high light are

especially vulnerable to photodamage, causing steep declines in dark-acclimated F_v/F_m with increasing

light. Photoinhibition due to source-sink imbalance is usually accompanied by compensatory upregulation

in protective non-photochemical quenching (Adams III et al. 2013), but these two species fail to increase

qN under high light, which may put them at risk of damage. Sink limitation may contribute to the

photoinhibition we observe in this species, but we believe it is not the dominant explanation.

The microclimate could also alter photosynthesis in ways our light-response measurements failed

to capture because we controlled the chamber microclimate. Our water potential measurements show that

water stress was not very severe in any treatment, but even modest differences could change the timing of

stomatal closure (Brodribb et al. 2003). It also seems plausible that greater light exposure may make leaf

temperature higher in monoculture (Schymanski et al. 2013), which could push leaves above thermal

optima for photosynthesis. Such microclimatic factors could contribute to the physiological benefits of

shading beyond what we could measure in light-response curves.

The finding that shade can increase growth and photosynthesis has precedents. Ball et al. (1991)

and Egerton et al. (2000) showed that shade enhances assimilation and growth of evergreen *Eucalyptus*

seedlings during the winter, when cold may cause photoinhibition. Howell et al. (2002) also found that

among evergreen divaricating shrubs of New Zealand, leafless outer branches reduce photoinhibition

during the winter by shading inner leaves. Species can also facilitate each other from light stress in

communities of phytoplankton (Gerla et al. 2011), which nearly all face some risk of photoinhibition

(Edwards et al. 2015), especially under cold temperatures (Edwards et al. 2016) These studies show that

facilitation through shading can occur in many environments, and perhaps more so in stressful ones.

Mechanisms of photoprotection

Plants' strategies to avoid damage from the stress of excess light can be classified broadly into

biochemical and structural strategies. We show that trees in full sun allocate more to biochemical

photoprotection, as indicated by lower PRI in all species and higher qN in all species but *T. americana*

and *A. negundo*. Our leaf angle survey is also consistent with the idea that plants may steeply incline their

leaves to avoid intercepting excess light—a form of structural photoprotection. Leaves that are steeply

inclined intercept less light per unit area, particularly during midday, when solar radiation is otherwise

most intense. The dramatic trend in leaf angles suggests that gross structural characteristics allow these

species use to regulate their light absorption.

Some plants may lack perfect photoprotective mechanisms in part because they are adapted to

environments where the costs of high photoprotective investment outweigh the benefits. Delayed

relaxation of non-photochemical quenching may hinder photosynthesis even when light declines enough

to relieve the imminent threat of damage (Zhu et al. 2004; Murchie & Niyogi 2011; Kromdijk et al.

2016). Building pigments and proteins for photoprotection uses resources that plants could otherwise

allocate to chlorophyll, RuBisCO, or structural tissue components, although the concentrations of such

photoprotective compounds are often low. Steep leaf angles may also decrease photosynthesis by

reducing light interception even when light is limiting—and all plants experience light-limiting conditions

at least sometimes, as during cloudy days or nighttime. All the same, plants in more open environments

may need photoprotective mechanisms to avoid even greater costs due to photodamage under brighter

conditions. That most species upregulate photoprotection in high light, despite the costs, implies that

these species may otherwise incur a risk of damage from excess light. These findings reinforce that light

stress and its alleviation through shading can contribute to net species interactions.

Competition and facilitation in biodiversity-ecosystem function relationships

Much of the work on facilitation in plant communities is guided by the stress-gradient hypothesis

(SGH), which proposes that facilitative interactions are more common in stressful environments (Bertness

& Callaway 1994; Maestre et al. 2009). We can understand photoinhibition in light of the SGH as

follows: Stressful conditions limit the use of light for photochemistry, turning high light exposure into an

additional stressor. Under such stressful conditions, shade from neighbors can ameliorate stress, resulting

in facilitation. Under benign conditions, where plants can use more of the available light for

photochemistry, high light exposure may not cause damage. Here, shading by neighbors can cause light

limitation, resulting in an adverse, competitive effect on growth. As we show here, whether a species is

stressed by its environment also depends on its own physiological tolerances.

The SGH brought attention to facilitation in plant ecology, including in biodiversity-ecosystem

function research. For example, Mulder et al. (2001) showed that in experimental moss communities,

biodiversity increased productivity only under drought, as species interactions became more positive. But

the mechanisms and consequences of facilitation are still overlooked in biodiversity-ecosystem function

research (Wright et al. 2017). A small but growing literature based in biodiversity experiments has

investigated various mechanisms through which plant species can facilitate each other, including

amelioration of water stress (Wright et al. 2015), dilution of natural enemies (Grossman et al. 2019),

nitrogen fixation by legume-rhizobia mutualisms (Temperton et al. 2007), and other forms of nutrient-

related belowground facilitation (Li et al. 2014). We show that shading can reduce stress induced by the

light environment—a specific form of abiotic stress amelioration, in which certain species alter the local

environment such that it is less stressful to other species (Wright et al. 2017). But shading can also cause

light limitation, resulting in a net competitive effect.

During the first three years of the FAB experiment, Grossman et al. (2017) found that diverse

plots overyield—they are more productive than a weighted average of the constituent species in

monoculture. Using the partition of Loreau & Hector (2001), most overyielding in FAB was a result of

complementarity effects, which are often interpreted as arising from synergistic interactions among

species, including facilitation. Overyielding is a hallmark of positive biodiversity-ecosystem function

relationships, and a whole community can only overyield if at least one of its constituent species

overyields. Even though we test how growth changes with neighbor size, not neighbor diversity, we show

that shading allows some species to grow faster in shaded bicultures or twelve-species plots. In the same

experiment, Grossman et al. (2017) found that *T. americana* had the greatest overyielding in multi-species

plots of any broadleaf species, especially when growing with conifers or *B. papyrifera*. Grossman et al.

(2017) further showed that conifer partners in these species combinations also grew faster than they did in

monoculture, resulting in community overyielding.

The two broadleaf species that were facilitated by shade had low tolerance to excess light, and

likely benefited from the milder microclimate their larger neighbors created. The conifer species, which

are fast-growing and less shade-tolerant (except for *P. strobus*), may also have grown faster in bicultures

with broadleaf species because they faced less competition for light than in monoculture. Other

researchers have found that heterogeneity in shade tolerance explains the effect of tree diversity on

productivity, but they attributed this result to higher stand-level light capture enabled by crown

complementarity (Zhang et al. 2012; Toïgo et al. 2017; Searle and Chen 2019). We show that facilitation

of shade-tolerant species by shade-intolerant species is an equally compelling (and perhaps

complementary) explanation.

In the literature on biodiversity and ecosystem function, the attempt to parse out mechanisms—

such as niche partitioning and facilitation—often depends on statistical partitions of biomass or

productivity data (Loreau & Hector 2001; Fox & Kerr 2012) whose interpretation is often debated

684 (Carroll et al. 2011; Pillai & Gouhier 2019). But because productivity has physiological determinants, we

685 can explain its patterns using physiological measurements. In this study, survey-based tree growth data

686 and physiological measurements both show that certain species respond positively and others respond

687 negatively to the size of their neighbors. But only the physiological measurements allow us to explain this

688 divergence—they show that these interactions emerge in part from the photosynthetic responses of trees

689 to the light environment created by their neighbors. The insights gained from such physiological

690 techniques may benefit plant ecology more widely as we seek to explain community patterns in terms of

691 basic aspects of plant function.

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1042 **Tables**

1043

1044 Table 1: Characteristics of species in FAB. Except for *Betula papyrifera*, broadleaf species tended to

1045 grow slower than needleleaf conifers. Shade tolerance is drawn from Niinemets and Valladares (2006),

1046 who evaluated it along a 1-5 scale where higher values indicate an ability to grow under lower light.

1047

Scientific name	Species code	Common name	Family	Leaf form	Mean RGR in monoculture, 2013-6 (/year)	Shade tolerance	Partner in shaded biculture
Acer negundo	ACNE	Box elder	Sapindaceae	Broadleaf	0.039	3.47	None
Acer rubrum	ACRU	Red maple	Sapindaceae	Broadleaf	0.379	3.44	*Pinus banksiana*
Betula papyrifera	BEPA	Paper birch	Betulaceae	Broadleaf	1.272	1.54	*Pinus banksiana*
Juniperus virginiana	JUVI	Eastern redcedar	Cupressaceae	Needleleaf	0.975	1.28	NA
Pinus banksiana	PIBA	Jack pine	Pinaceae	Needleleaf	2.288	1.36	NA
Pinus resinosa	PIRE	Red pine	Pinaceae	Needleleaf	1.619	1.89	NA
Pinus strobus	PIST	White pine	Pinaceae	Needleleaf	1.722	3.21	NA
Quercus alba	QUAL	White oak	Fagaceae	Broadleaf	0.735	2.85	*Pinus strobus*
Quercus ellipsoidalis	QUEL	Northern pin oak	Fagaceae	Broadleaf	0.722	NA	*Pinus strobus*
Quercus macrocarpa	QUMA	Bur oak	Fagaceae	Broadleaf	0.547	2.71	*Betula papyrifera*
Quercus rubra	QURU	Northern red oak	Fagaceae	Broadleaf	0.505	2.75	*Pinus strobus*
Tilia americana	TIAM	American basswood	Tiliaceae	Broadleaf	0.724	3.98	*Pinus strobus*

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 Figure captions

 Fig. 1, top: Imagine the same slow-growing broadleaf species (round) found in two communities: a

 monoculture (left), and a biculture (right) with a faster growing conifer (triangular). In the absence of

 shading by neighbors, all trees would receive the amount of light available at the top of the canopy

 (dashed yellow line). In monoculture, the top leaves of the broadleaf trees still receive close to the full

 amount of available light—the relative light availability (RLA), as we define it, is close to 1. In shaded

 biculture, these trees are shaded by their larger conifer neighbors such at any moment, their top leaves

 receive half the light that reaches the top of the canopy—their RLA is 0.5. These trees may be more likely

 to suffer from light limitation, but they are also at less risk for photoinhibition.

 Bottom left: For a single leaf, carbon assimilation tends to increase along a saturating curve as light

 intensity increases. Consider two hypothetical photosynthetic light-response curves of leaves from shaded

 (blue) and unshaded (red) plants. Given a light-response curve, a reduction in light reduces photosynthesis

 (represented by the small black arrow along the red curve). But if a long-term reduction in light spares the

 plant from photoinhibition, it can change the shape of the curve such that carbon assimilation at any given

 light level is greater (represented by the small black vertical arrow). The balance (represented by the large

 dashed arrow) may either increase or reduce carbon assimilation. In this framework, trees benefit most

 from shading when light is otherwise abundant: the risk of photoinhibition is greater, and the proportional

 cost of a given percentage decline in light is lower, since light-response curves are usually concave-down.

 Bottom right: The original light-response curves, plotted as a function of light reaching the leaf surface,

 offer no simple visual way to compare assimilation rates as a function of light at the top of the canopy.

 When the amount of light at the top of the canopy is 2000 μmol m^{-2} s^{-1}, the top leaves of the shaded plant

 (blue; RLA = 0.5) only receive $2000 \times 0.5 = 1000$ μmol m^{-2} s^{-1} of light (blue dashed line)—or a

 corresponding fraction of any other light level. The unshaded plant (red; RLA = 1) receives the full 2000

 μmol m^{-2} s^{-1} (red dashed line). To make it easy to compare assimilation rates as a function of light at the

1079   top of the canopy, we can rescale the blue light-response curve two-fold (the inverse of its RLA) along

1080   the *x*-axis, such that the two dashed lines coincide.

1081

Fig. 2: Relative growth rate (RGR) of woody stem volume in individuals of each species between fall

2016 and 2018 as a function of the total stem volume of all neighbors. Gray dots represent individuals,

and the OLS regression line is fit through the individual-level data. Large colored dots are aggregated to

the plot scale, as described in the main text, and color-coded by treatment. A shaded biculture is a plot

where the focal broadleaf species grows with either a conifer or *Betula papyrifera* (unless *B. papyrifera* is

the focal broadleaf species). An open biculture is a plot where the focal broadleaf grows with a broadleaf

species other than *B. papyrifera*.

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Fig. 3: Mass-based photosynthetic light-response curves for four broadleaf species. The left panels

display carbon assimilation rates as a function of chamber light intensity; the right panels represent an

estimate of realized assimilation rates as a function of light intensity at the top of the canopy. These rates

are lower because of shading by neighbors (as in Fig. 1, bottom right). Species are arranged from the least

shade-tolerant (*Betula papyrifera*) to the most (*Tilia americana*). Error bars in left panels and gray

ribbons in right panels are $\pm$ 1 SE.

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Fig. 4: Chlorophyll fluorescence parameters and the photochemical reflectance index (PRI) as a function

of relative light availability across species and treatments. Best-fit lines come from species-specific OLS

regressions. Dark-acclimated F_v/F_m is the maximal quantum efficiency of PSII photochemistry and

declines as a plant becomes more photoinhibited. qN describes a plant's capacity to use non-

photochemical quenching to dissipate energy. ϕ_{NO} represents the proportion of light dissipated in an

unregulated way, which may contribute to photodamage.

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1104   Fig. 5: Compared to shade-intolerant species, shade-tolerant species show greater declines in dark-

1105   acclimated F_v/F_m (top) and smaller increases in qN (middle) as light availability increases. A_{sat} derived

1106   from light-response curves correlates positively and strongly with dark-acclimated F_v/F_m; slopes are

1107   statistically indistinguishable among species (bottom). Species codes are found in Table 1.

1108

1109   Fig. 6: In *Tilia americana*, fall leaf abscission occurs later in monoculture than in shaded biculture.

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1111   Fig. 7: The angles of top leaves of three species are more horizontal in shaded biculture plots than in

1112   monoculture.

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1128 **Figures**

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1130 **Fig. 1**

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1159   **Fig. 2**

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1163 **Fig. 4**

Fig. 5

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1185   **Fig. 7**

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1187 **Supplement to: "Physiological responses to light explain competition and facilitation in a tree**

1188 **diversity experiment"**

1189

Appendix 1

Photosynthetic light-response curve methods

We measured photosynthetic light-response curves from the four focal species (*Acer negundo*,

Betula papyrifera, *Quercus ellipsoidalis*, and *Tilia americana*) during July 2018 using an LI-6400 gas

exchange system (LI-COR BioSciences, Lincoln, NE) with a 6400-40 Leaf Chamber Fluorometer head.

From each tree (n = 72 total), we selected a fully expanded, sun-exposed leaf with no visible disease or

herbivory. We noted the angle of each leaf relative to horizontal to the nearest 15˚ before beginning

measurements. Each curve had nine steps in descending order of brightness: 2000, 1500, 1000, 500, 200,

100, 50, 20, and 0 µmol m^{-2} s $^{-1}$.

During each curve, we controlled environmental conditions to reduce the immediate effects of

environmental conditions other than light on stomatal conductance. We kept relative humidity at a

favorable level, typically between 65 and 75%, and set the block temperature as close as possible to 25

˚C. The reference CO_2 concentration remained at 400±10 ppm. We kept each leaf in the chamber for at

least five minutes before beginning each curve. We checked that outputs were stable, and matched IRGAs

before each measurement.

Since it is hard to control chamber conditions perfectly, any differences that remain among

treatments could contribute to the variation we observe in photosynthetic function. During light-response

curves, we found no difference among treatments in leaf-to-air VPD within the chamber. There were

significant but slight differences in leaf temperature inside the chamber; leaves in monoculture, shaded

biculture, and twelve-species plots had average temperatures of 27.7 ˚C, 26.5 ˚C, and 27.0 ˚C,

respectively. Previous studies on the temperature response of some of the same species in Minnesota have

typically found thermal optima below 26 ˚C (Sendall et al. 2015). But while cooler leaves may be closer

to thermal optima, the small differences we find in leaf temperature cannot explain much of the variation

in assimilation rates.

In all but five light-response curves, we also used a saturating pulse to measure light-acclimated

chlorophyll fluorescence at each light level. These data allowed us to calculate ϕ_{PSII}, a measure of the

efficiency of Photosystem II (PSII) at each actinic (or photosynthetic) light level, as:

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$$\phi_{PSII} = \frac{F'_m - F_s}{F'_m}$$

1218 where F_s is the steady-state fluorescence yield under actinic light, and F_m' is the maximum fluorescence

1219 yield of the actinic light-acclimated sample following a saturating light pulse, which temporarily closes

1220 all PSII reaction centers. From this parameter, we calculated electron transport rate as:

1221
$$ETR = \phi_{PSII} \times PPFD \times 0.42 \times 0.5$$

1222 where PPFD is photosynthetic photon flux density (the light level inside the chamber, in μmol m^{-2} s^{-1}),

1223 0.42 is an estimate of leaf absorptance, and 0.5 is an estimate of the fraction of photons that are captured

1224 by PSII rather than PSI. Electron transport rate is a key parameter because electron transport is one of the

1225 main steps that is capable of limiting photosynthesis (Farquhar et al. 1980), especially when light

1226 availability or PSII efficiency are low.

1227

Chlorophyll fluorescence procedure

As described in the main text, we used dark- and light-acclimated measurements of chlorophyll

fluorescence to calculate several derived parameters, from which we make inferences about

photosynthetic and photoprotective function. Most importantly, we wanted to assess: (1) capacity for non-

photochemical quenching, as measured through the parameters qN or NPQ; and (2) the partitioning of

light energy dissipation among pathways, as assessed using the quantum yields ϕ_{PSII}, ϕ_{NPQ}, and ϕ_{NO}.

These derived chlorophyll fluorescence parameters are calculated from a few basic measurements

taken during the dark- and light-acclimated measurements, including: F_o, the minimum fluorescence yield

of a dark-acclimated sample with all PSII reaction centers open; F_o', the minimum fluorescence yield of a

light-acclimated sample with all PSII reaction centers open (following far-red light exposure); and F_m, the

maximum fluorescence yield of a dark-acclimated sample following a saturating pulse. F_v, the variable

fluorescence yield of a dark-acclimated sample, is calculated as $F_m - F_o$.

The parameters qN and NPQ are meant to indicate the rate constant of the thermal dissipation of

energy from PSII (Bilger & Schreiber 1986; Bilger & Björkman 1990). These parameters increase with

the use of NPQ to dissipate light—in this case, under exposure to 1000 µmol m^{-2} s^{-1}. They are calculated,

respectively, as:

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$$qN = 1 - \frac{F'_m - F'_o}{F_m - F_o}; NPQ = \frac{F_m - F'_m}{F'_m}$$

In our dataset, qN is highly correlated ($R^2 = 0.96$) with the natural logarithm of NPQ; we chose to report

qN to easily visualize variation among leaves with low NPQ and to help satisfy the error term normality

assumption of OLS regression.

The quantum yields ϕ_{PSII}, ϕ_{NPQ}, and ϕ_{NO} represent the proportions of absorbed light energy that are

dissipated through photochemistry, non-photochemical quenching, and non-regulated dissipation. Since

these pathways exhaust the options, their quantum yields sum to 1. We calculated them following the

'lake model' derivation of Kramer et al. (2004), which treats photosynthetic reaction centers as connected

by shared antennae. We achieve qualitatively similar results using the simpler analogues found in

Hendricksen et al. (2005).

The parameter ϕ_{PSII} is calculated as shown above (Supplemental Methods, *Photosynthetic light-*

response curve procedure). In the derivation from Kramer et al. (2004), ϕ_{NPQ} and ϕ_{NO} are calculated using

the parameter qL, an estimate of the fraction of open PSII reaction centers. qL is calculated as:

1257
$$qL = \frac{F'_m - F_s}{F'_m - F'_o}\frac{F'_o}{F_s}$$

1258   Given this parameter, we can calculate ϕ_{NO} as:

1259
$$\phi_{NO} = \frac{1}{NPQ + 1 + qL\left(\frac{F_m}{F_o} - 1\right)}$$

and given the other two quantum yields, we can calculate ϕ_{NPQ} as:

$$\phi_{NPQ} = 1 - \phi_{PSII} - \phi_{NO}$$

Many researchers, including Murchie and Lawson (2013) and Tietz et al. (2017) treat ~0.83 as a

theoretical maximum of ϕ_{PSII} (or dark-acclimated F_v/F_m) after a period of dark-acclimation long enough to

reduce non-photochemical quenching to zero. Similarly, one may treat ~0.83 as a theoretical maximum of

$\phi_{PSII} + \phi_{NPQ}$, deviations from which indicate high ϕ_{NO}, and thus, a declining ability to regulate energy

dissipation. We show this putative maximum in our ternary plot of the quantum yields (Fig. S7).

Appendix 2

Here, we compare statistical models testing the influence of average neighbor volume on relative growth

rate (RGR) of focal individuals (or whole plots). The t-statistics, degrees of freedom (df), and p-values

provided are only for the parameter of interest (β_1, the slope term associated with average neighbor

volume), even for multivariate models. For all fixed-effect models, we report a standard adjusted R^2. For

the mixed-effect model with plot incorporated as a random effect, we calculated marginal and conditional

R^2 using the method of Nakagawa & Schielzeth (2013) as implemented in the R package $MuMIn$ $v.$

$1.43.17$ (Bartoń 2020), and estimated p-values using Satterthwaite's degrees of freedom method, as

implemented in the R package $lmerTest$ $v.3.1.2$ (Kuznetsova et al. 2017).

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In the manuscript, we present summary statistics from the first and simplest model shown here (A) to

predict focal individual RGR. The results, including the slope estimates, are mostly robust to the model

specification (B-F). We highlight in yellow two cases where our inferences about how species respond to

neighbor volume would be different (based on comparing p-values to an α = 0.05 threshold) under an

alternate model specification. For models that include focal individual size in 2013 or 2016, we note that

these values are not usually correlated with RGR between 2016 and 2018 ($R^2 < 0.01$ except for *Acer*

negundo in 2016), so there is no problem of multicollinearity.

1284

The final three models have independent variables other than focal individual RGR. In G, we predict plot-

level RGR from the plot-scale average of each individual's average neighbor biomass. In H and I, we

predict the absolute growth rate of focal individuals. These regressions violate the homoscedasticity and

normality of errors assumptions of OLS regression, but we include them because they allow us to

compare model parameters including (H) and excluding (I) trees that died between 2016 and 2018,

showing very little difference.

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1292

1293   A. $\text{FocalRGR}_{2016-2018} = \alpha + \beta_1 \times \text{NeighborVolume}$

Species	$\beta_1 \times 10^4$ (SE $\times 10^4$)	R^2	T	df	P
Acer negundo	1.544 (0.540)	0.018	2.857	384	0.005
A. rubrum	-1.718 (0.349)	0.033	-4.922	622	1.080×10^{-6}
Betula papyrifera	0.425 (0.241)	0.004	1.760	585	0.079
Quercus alba	-1.902 (0.378)	0.038	-5.031	609	6.429×10^{-7}
Q. ellipsoidalis	-1.498 (0.322)	0.030	-4.647	672	4.049×10^{-6}
Q. macrocarpa	-1.523 (0.327)	0.027	-4.658	751	3.782×10^{-6}
Q. rubra	-2.199 (0.330)	0.056	-6.411	672	2.716×10^{-10}
Tilia americana	1.044 (0.261)	0.019	3.997	757	7.043×10^{-5}

1294

1295   B. $\text{FocalRGR}_{2016-2018} = \alpha + \beta_1 \times \text{NeighborVolume}$ (excluding individuals on a block edge)

Species	$\beta_1 \times 10^4$ (SE $\times 10^4$)	R^2	t	Df	p
Acer negundo	1.620 (0.556)	0.021	2.915	349	0.004
A. rubrum	-1.741 (0.363)	0.034	-4.797	618	2.025×10^{-6}
Betula papyrifera	0.398 (0.247)	0.003	1.610	542	0.108
Quercus alba	-1.932 (0.389)	0.040	-4.964	573	9.135×10^{-7}
Q. ellipsoidalis	-1.426 (0.332)	0.027	-4.293	629	2.040×10^{-5}
Q. macrocarpa	-1.371 (0.332)	0.023	-4.134	684	4.013×10^{-5}
Q. rubra	-2.135 (0.339)	0.058	-6.303	633	5.480×10^{-10}
Tilia americana	0.967 (0.270)	0.016	3.582	710	3.641×10^{-4}

1296

1297   C. $\text{FocalRGR}_{2016-2018} = \alpha + \beta_1 \times \log(\text{NeighborVolume})$

Species	$\beta_1 \times 10^2$ (SE $\times 10^2$)	R^2	t	Df	p
Acer negundo	9.731 (1.759)	0.072	5.532	383	5.877×10^{-8}
A. rubrum	-4.498 (1.838)	0.007	-2.447	662	0.0157
Betula papyrifera	2.698 (2.009)	0.001	1.343	585	0.180
Quercus alba	-8.516 (2.044)	0.026	-4.167	609	3.54×10^{-5}
Q. ellipsoidalis	-5.527 (1.809)	0.012	-3.056	672	0.00234
Q. macrocarpa	-5.610 (1.701)	0.013	-3.298	751	0.00102
Q. rubra	-12.507 (2.058)	0.051	-6.079	672	2.030×10^{-9}
Tilia americana	8.844 (2.090)	0.022	4.232	757	2.605×10^{-5}

1298

1299   D. $\text{FocalRGR}_{2016-2018} = \alpha + \beta_1 \times \text{NeighborVolume} + \beta_2 \times \text{Focal\_size}_{2013}$
1300   (*Acer negundo* excluded because not enough individuals survived between 2013 and 2018.)
1301

Species	$\beta_1 \times 10^4$ (SE $\times 10^4$)	R^2	t	df	p
Acer negundo					
A. rubrum	-1.795 (0.399)	0.040	-4.503	457	3.401×10^{-5}
Betula papyrifera	0.390 (0.254)	0.002	1.534	542	0.125
Quercus alba	-2.150 (0.395)	0.063	-5.449	411	5.462×10^{-7}
Q. ellipsoidalis	-1.781 (0.334)	0.056	-5.332	491	2.412×10^{-7}
Q. macrocarpa	-1.458 (0.322)	0.031	-4.536	578	3.584×10^{-5}
Q. rubra	-1.784 (0.378)	0.050	-4.720	497	1.056×10^{-6}
Tilia americana	0.8692 (0.255)	0.014	3.414	702	0.00286

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1303   E. $\text{FocalRGR}_{2016\text{-}2018} = \alpha + \beta_1 \times \text{NeighborVolume} + \beta_2 \times \text{Focal\_size}_{2016}$

1304

Species	$\beta_1 \times 10^4$ (SE $\times 10^4$)	R^2	t	df	p
Acer negundo	1.302 (0.526)	0.078	2.476	383	0.0137
A. rubrum	-1.800 (0.360)	0.034	-5.001	661	7.320×10^{-7}
Betula papyrifera	0.502 (0.248)	0.005	2.022	584	0.0437
Quercus alba	-1.875 (0.382)	0.037	-4.910	608	3.664×10^{-6}
Q. ellipsoidalis	-1.485 (0.325)	0.028	-4.572	671	2.335×10^{-5}
Q. macrocarpa	-1.533 (0.328)	0.026	-4.673	750	2.116×10^{-5}
Q. rubra	-2.088 (0.330)	0.061	-6.325	671	2.929×10^{-10}
Tilia americana	1.078 (0.262)	0.021	4.112	756	1.303×10^{-4}

1305

1306   F. $\text{FocalRGR}_{2016\text{-}2018} = \alpha + \beta_1 \times \text{NeighborVolume} + (1|\text{Plot})$

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Species	$\beta_1 \times 10^4$ (SE $\times 10^4$)	R_m^2	R_c^2	t	df	p
Acer negundo	0.912 (0.543)	0.007	0.376	1.680	377	0.0938
A. rubrum	-1.299 (0.440)	0.020	0.187	-2.953	458	0.00331
Betula papyrifera	0.399 (0.271)	0.005	0.036	1.477	151	0.142
Quercus alba	-1.519 (0.462)	0.025	0.147	-3.286	380	0.00111
Q. ellipsoidalis	-1.598 (0.375)	0.035	0.111	-4.260	296	2.750×10^{-5}
Q. macrocarpa	-1.079 (0.367)	0.013	0.216	-2.939	648	0.00342
Q. rubra	-2.115 (0.370)	0.057	0.087	-5.719	219	3.490×10^{-8}
Tilia americana	0.702 (0.326)	0.009	0.118	2.154	350	0.0319

1308

1309   G. $\text{PlotRGR}_{2016\text{-}2018} = \alpha + \beta_1 \times \text{MeanPlotNeighborVolume}$

Species	$\beta_1 \times 10^4$ (SE $\times 10^4$)	R^2	t	df	p
Acer negundo	-0.634 (1.943)	-0.042	-0.326	21	0.747
A. rubrum	-1.193 (0.891)	0.026	-1.339	29	0.191
Betula papyrifera	0.147 (0.572)	-0.035	0.256	27	0.800
Quercus alba	-3.826 (1.019)	0.311	-3.755	28	8.06×10^{-4}
Q. ellipsoidalis	-1.151 (1.190)	-0.002	-0.967	31	0.341
Q. macrocarpa	-3.052 (1.098)	0.174	-2.779	31	0.00918
Q. rubra	-1.529 (1.115)	0.029	-1.371	28	0.181
Tilia americana	1.578 (0.423)	0.282	3.735	32	7.33×10^{-4}

1310

1311   H. $\text{FocalAbsoluteGrowth} = \alpha + \beta_1 \times \text{NeighborVolume}$

1312

Species	$\beta_1 \times 10^2$ (SE $\times 10^2$)	R^2	t	df	p
Acer negundo	4.413 (0.903)	0.048	4.888	455	1.416×10^{-6}
A. rubrum	-0.907 (0.589)	0.002	-1.540	707	0.124
Betula papyrifera	6.416 (1.804)	0.018	3.556	632	4.043×10^{-4}
Quercus alba	0.116 (1.012)	-0.001	0.114	651	0.909
Q. ellipsoidalis	0.182 (0.704)	-0.001	0.262	715	0.794
Q. macrocarpa	-0.304 (1.049)	-0.001	-0.290	763	0.772
Q. rubra	-3.165 (0.890)	0.016	-3.558	719	3.989×10^{-4}
Tilia americana	13.823 (1.866)	0.064	7.407	788	3.328×10^{-13}

1313
1314

1315   I. FocalAbsoluteGrowth = $\alpha + \beta_1 \times$ NeighborVolume (excluding individuals that died by 2018)

1316

Species	$\beta_1 \times 10^2$ (SE $\times 10^2$)	R^2	t	df	p
Acer negundo	4.902 (1.041)	0.052	4.708	397	3.483×10^{-6}
A. rubrum	-0.459 (0.646)	-0.001	-0.711	662	0.478
Betula papyrifera	7.713 (1.871)	0.027	4.123	585	4.285×10^{-5}
Quercus alba	0.621 (1.111)	-0.001	0.559	612	0.576
Q. ellipsoidalis	0.394 (0.744)	-0.001	0.529	675	0.597
Q. macrocarpa	-0.169 (1.083)	-0.001	-0.156	753	0.876
Q. rubra	-3.007 (0.965)	0.013	-3.118	678	0.00190
Tilia americana	16.145 (1.937)	0.083	8.336	758	3.606×10^{-16}

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1362 **Supplemental figure captions**

1363

1364 Fig. S1: Example images of *Tilia americana* in shaded biculture (top) and monoculture (bottom). Red

1365 scale bars represent about 1 m in the foreground.

1366

1367 Fig. S2: Relative light availability (top) and leaf mass per area (bottom) of each species across treatments.

1368 We show LMA values from all leaves used in carbon assimilation and chlorophyll fluorescence

1369 measurements. Error bars in left panels are $\pm$ 1 SE.

1370

1371 Fig. S3: Growth of woody stem volume in individuals of each species between fall 2017 and 2018 as a

1372 function of the total stem volume of all neighbors. Gray dots represent individuals, while colored dots are

1373 aggregated to the plot scale and color-coded by treatment. Two OLS regression lines are fit for each

1374 species: One (blue) to all data, and one (red) excluding trees that died between 2017 and 2018.

1375

1376 Fig. S4: Responses of area-based carbon assimilation and C_i (internal CO_2 concentration) to light

1377 availability for the four focal broadleaf species. Error bars are $\pm$ 1 SE.

1378

1379 Fig. S5: Responses of ETR and stomatal conductance to light availability for the four focal broadleaf

1380 species. Error bars are $\pm$ 1 SE.

1381

1382 Fig. S6: An estimate of total carbon assimilation across species and treatments throughout the month of

1383 July, calculated by applying a time series of solar radiation from a nearby weather station as inputs to our

1384 rescaled light-response curves. Error bars are $\pm$ 1 SE.

1385

Fig. S7: A ternary plot depicting variation among species and treatments in ϕ_{PSII}, ϕ_{NPQ}, and ϕ_{NO}, as

calculated using the lake model derivations of Kramer et al. (2004). These three quantities describe light

energy dissipation through photochemistry, through non-photochemical quenching, and through non-

regulated dissipation, respectively. A thick dashed line represents 0.83, the theoretical maximum

efficiency of PSII posited in Tietz et al. (2017). Deviations to the lower right of this line represent

increasingly high ϕ_{NO}, which can result in photodamage.

1392

1393 Fig. S8: Water potential before dawn (Ψ_{PD}; top) and at midday (Ψ_{MD}; bottom) as a function of light

1394 availability. Data are from monoculture and shaded biculture treatments in all eight species (Ψ_{PD}) or only

1395 the four focal species (Ψ_{MD}). Best-fit lines come from species-specific OLS regressions. Two outliers in

1396 Ψ_{PD} are not included.

1430 **Supplemental figures**

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1432 **Fig. S1**

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1444   **Fig. S2**
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Fig. S3

 Fig. S4

1470   **Fig. S5**

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1472   **Fig. S6**

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1475   **Fig. S7**

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1477   **Fig. S8**

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