

1 Evaluating forest subcanopy response to moderate severity disturbance and  
2 contribution to ecosystem-level productivity and resilience

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4 Robert T. Fahey<sup>1\*</sup>, Ellen J. Stuart-Haëntjens<sup>2</sup>, Christopher M. Gough<sup>2</sup>, Aubrie De La Cruz<sup>3</sup>,  
5 Elizabeth Stockton<sup>3</sup>, Christoph S. Vogel<sup>3</sup>, and Peter S. Curtis<sup>4</sup>

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7 <sup>1</sup>University of Connecticut, Department of Natural Resources and the Environment & Center for  
8 Environmental Sciences and Engineering

9 <sup>2</sup>Virginia Commonwealth University, Department of Biology

10 <sup>3</sup>University of Michigan Biological Station

11 <sup>4</sup>Ohio State University, Department of Evolution, Ecology and Organismal Biology

12

13 \*Corresponding author

14 1376 Storrs Rd. Unit 4087

15 Storrs, CT 06269-4087

16 robert.fahey@uconn.edu

17 (860) 486-4169

18

19 **Abstract**

20 North American temperate forests have functioned as a terrestrial carbon (C) sink for more  
21 than a century, but the future of this sink is highly uncertain as disturbance frequency increases  
22 and regrown forests approach maturity. The subcanopy is integral to the functional recovery of  
23 forests, supporting short-term resilience of primary production and longer-term shifts in tree  
24 species composition and diversity. However, the factors that contribute to variation in forest  
25 subcanopy response to disturbance are not well understood. In this study, we investigated  
26 subcanopy shifts in aboveground wood net primary productivity (ANPP<sub>w</sub>) and composition  
27 following experimental moderate severity disturbance emulating natural canopy mortality from  
28 age-related senescence. We assessed the importance of variation in disturbance severity, site  
29 fertility, and community composition on subcanopy disturbance response and contribution to  
30 total (canopy and subcanopy) ANPP<sub>w</sub> response. We also assessed the effect of the moderate  
31 severity disturbance on species composition and diversity, and competitive patterns within the  
32 subcanopy layer. Subcanopy aboveground biomass and ANPP<sub>w</sub> increased substantially relative  
33 to pre-disturbance levels by a factor of 1.4 and 22.7, respectively. The subcanopy (stems < 8cm  
34 DBH) made up a large component of overall (canopy plus subcanopy) post disturbance ANPP<sub>w</sub>  
35 (16.2%) and disturbance response (post-disturbance ANPP<sub>w</sub> /pre-disturbance ANPP<sub>w</sub>; 54.1%).  
36 Subcanopy ANPP<sub>w</sub>, subcanopy post-disturbance ANPP<sub>w</sub> response, and subcanopy contribution  
37 to total post-disturbance ANPP<sub>w</sub> response were all most strongly predicted by subcanopy  
38 community composition in combination with canopy composition and site fertility. Variation in  
39 disturbance severity was not a strong predictor of subcanopy ANPP<sub>w</sub> response to disturbance.  
40 Subcanopy compositional trends and growth patterns both indicate likely increased heterogeneity  
41 in canopy composition (greater  $\beta$  diversity) and a potential shift toward greater dominance by

42 mid-tolerant *Quercus rubra* (northern red oak). Our results illustrate the importance of the  
43 subcanopy in the response of forest productivity to moderate severity disturbance and illustrate  
44 that composition of the subcanopy layer exerts a strong influence on the growth response both of  
45 the subcanopy and the forest as a whole. Our findings highlight the unique role of moderate  
46 severity disturbance, relative to more severe disturbances, in promoting biological and structural  
47 heterogeneity in forest ecosystems and favoring underrepresented mid-tolerant species.

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49 Keywords: intermediate disturbance; northern hardwood; ecological silviculture; *Quercus rubra*,  
50 *Acer rubrum*; Great Lakes region

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52 Highlights:

- 53 • Subcanopy contributed substantially to forest growth response to moderate disturbance
- 54 • Variable subcanopy response was mediated by composition not disturbance severity
- 55 • Moderate severity disturbance promoted growth of mid-tolerant northern red oak

## 56 **1. Introduction**

57 For more than a century, secondary forests of eastern North America have represented an  
58 important global carbon (C) sink during their regrowth following widespread harvesting and  
59 agricultural clearing more than a century ago (Houghton *et al.*, 1999; Caspersen *et al.*, 2000;  
60 Birdsey *et al.*, 2006). However, the sustainability of this C sink is in doubt because of changes in  
61 disturbance frequency and intensity, advancing forest age, and land management and global  
62 wood product markets (Birdsey *et al.*, 2006). Agricultural abandonment and resulting  
63 afforestation has slowed considerably, such that the area of forest cover is no longer increasing in  
64 eastern North America (Drummond and Loveland, 2010). In addition, decreased utilization of  
65 northern forests for wood products has led increasing proportions of these forests to attain mid-  
66 late-successional status (Birdsey *et al.*, 2006; Gough *et al.*, 2016). The transition of forests from  
67 early to middle and late-successional communities will have implications for both short- and  
68 long-term C dynamics (Gough *et al.*, 2013) as well as longer-term composition of the regional  
69 forest and resilience to stressors such as global climate change and exotic pest introductions  
70 (Duveneck *et al.*, 2014b). However, the effect that ongoing changes in land management,  
71 disturbance regimes, and successional transitions will have on C sequestration and community  
72 composition in eastern North American forests remains uncertain (Caspersen *et al.*, 2000; Hurtt  
73 *et al.*, 2002; Duveneck *et al.*, 2014a; Gough *et al.*, 2016).

74 An important feature of the regional transition from early to late-successional canopy  
75 dominance will be widespread senescence of canopy dominant early-successional trees and  
76 subcanopy response to this disturbance (Birdsey *et al.*, 2006; Gough *et al.*, 2010). This transition  
77 will likely affect C dynamics in these forests (Bond-Lamberty *et al.*, 2004; Gough *et al.*, 2007;  
78 Amiro *et al.*, 2010). Forest productivity has been shown to be highly resistant to this type of

79 moderate (i.e., non-stand replacing) disturbance, and thus this transition may have limited near-  
80 term effects on C dynamics (Gough *et al.*, 2013). Subcanopy tree response may be an important  
81 factor in forest primary production resilience, helping maintain productivity through the  
82 transition from an early to mid-late successional stage (Stuart-Haëntjens *et al.*, 2015). However,  
83 subcanopy responsiveness to disturbance varies greatly with disturbance severity and among  
84 forest types and the factors associated with this variability are not always clear (Frelich, 2002).  
85 Pre-existing differences in subcanopy community composition and structure may be an  
86 important factor underlying variability in the resistance of ecosystem productivity to disturbance  
87 (Frelich, 2002; Papaik and Canham, 2006).

88 Subcanopy community composition and the growth response of its constituents to  
89 successional transition will also affect the composition and structure of future forest stands (e.g.,  
90 Henry and Swan, 1974; Frelich, 2002; Halpin and Lorimer, 2016a). Future species composition  
91 will greatly affect the resilience of the region's forest to threats such as climate change and  
92 introduced pests and pathogens (Duveneck *et al.*, 2014b). For example, in many forests in  
93 northeastern North America the dominance of maples (*Acer spp.*) is increasing greatly relative to  
94 historical conditions due to loss of fire and resulting "mesophication" in the landscape, and much  
95 of this shift has come at the expense of oak (*Quercus*) species (Nowacki and Abrams, 2008).  
96 Reduced diversity and extreme dominance of forest stands and landscapes by maples could  
97 reduce the resilience of these ecosystems to climate change (Stoy *et al.*, 2008) and could make  
98 the landscape especially vulnerable to the exotic pest Asian Longhorned Beetle (Haack *et al.*,  
99 2010). In addition, there has also been a trend toward reduced tree functional diversity in many  
100 forested landscapes, which is due in part to the shift to maple dominance (Rhemtulla *et al.*,  
101 2009). Coniferous species, which often have important functional roles in northern mixed

102 hardwood forests (e.g., wildlife habitat and unique structural elements; DeGraaf *et al.*, 1993;  
103 Fahey and Lorimer, 2014b), have declined greatly in importance relative to historic conditions in  
104 some areas (Schulte *et al.*, 2007; Fahey and Lorimer, 2014a). Also, many shade intolerant/mid-  
105 tolerant species are underrepresented in modern northern hardwood forests, in part due to  
106 reliance on single-tree selection management which favors shade tolerant, late-successional  
107 species (Neuendorff *et al.*, 2007). This trend toward reduced diversity of species and functional  
108 groups may become even more pronounced in the future if the landscape transitions to increased  
109 late-successional community dominance (Duveneck *et al.*, 2014a).

110       The transition from early to mid-late succession can happen rapidly and produce a somewhat  
111 abrupt moderate severity canopy disturbance, especially if mediated by environmental factors  
112 (e.g., drought) or pest/pathogen outbreaks (Allen *et al.*, 2010; Flower *et al.*, 2013). Such  
113 disturbances can promote shade intolerant and mid-tolerant species and maintain or increase  
114 overall species diversity (Frelich and Reich, 1999), but may also lead to accelerated succession  
115 (Allen *et al.*, 2012). Due to potential positive impacts on species diversity, size and age structure,  
116 productivity, and structural complexity, management approaches modeled after moderate  
117 severity disturbance have been advocated for use in ecological forest management (Hanson and  
118 Lorimer, 2007). Assessing the impact of moderate severity disturbances, whether related to  
119 natural canopy senescence or management, on ecosystem functioning and species composition  
120 will be important to predicting future conditions in forested landscapes (Halpin and Lorimer,  
121 2016b). However, although many studies have addressed the response of the subcanopy layer to  
122 gap creation (e.g., Canham, 1988; Lorimer and Frelich, 1989; Runkle, 1990), few have assessed  
123 the response of the subcanopy to moderate severity disturbance or its contribution to overall  
124 ecosystem resilience (Dyer *et al.*, 2010; Stuart-Haëntjens *et al.*, 2015).

125 The primary objective of this research was to better understand how the composition and  
126 structure of the forest subcanopy affect its response to canopy senescence and its role in overall  
127 forest resilience to moderate disturbance. We address this topic using an experimental forest  
128 canopy senescence experiment and focus on the following specific research objectives: 1)  
129 Evaluate how moderate severity disturbance affected characteristics of the subcanopy including  
130 biomass, stem density, species composition, and compositional trajectories, 2) Assess the effect  
131 of community composition and structure on individual species' growth responses following  
132 moderate severity disturbance, 3) Assess the effect of community composition and structure on  
133 subcanopy productivity following moderate severity disturbance, and 4) Evaluate how variation  
134 in subcanopy community composition and structure affect its relative contribution to overall  
135 forest productivity response to moderate severity disturbance.

136

## 137 **2. Methods**

### 138 *2.1 Study Site*

139 The study was conducted at the University of Michigan Biological Station (UMBS) in  
140 northern Michigan, USA (45° 35N 84° 43W). The location has a mean annual temperature of  
141 5.5° C and a mean annual precipitation of 817 mm. The forests of the area are primarily ~100  
142 year old aspen-dominated stands resulting from intense clear-cut harvesting and wildfires in the  
143 early 20<sup>th</sup> century (Gough *et al.*, 2007). Short-lived early successional species such as bigtooth  
144 aspen (*Populus grandidentata*), trembling aspen (*Populus tremuloides*), and paper birch (*Betula*  
145 *papyrifera*) have dominated the upper canopy, but are now beginning to reach the limit of their  
146 natural life-spans (Gough *et al.*, 2013). Many stands in the region are undergoing gradual  
147 transition to dominance by mid- or late-successional species such as northern red oak (*Quercus*

148 *rubra*), eastern white pine (*Pinus strobus*), red maple (*Acer rubrum*), sugar maple (*Acer*  
149 *saccharum*), and American beech (*Fagus grandifolia*). At UMBS these species dominate the  
150 subcanopy layer, defined in this forest as stems < 8cm in diameter at breast height (DBH, 1.37  
151 m), based on relationships between DBH and canopy position (Gough *et al.*, 2008). However,  
152 other species that are primarily confined to the subcanopy layer are also present, such as  
153 serviceberry (*Amelanchier arborea*), striped maple (*Acer pensylvanicum*), and ironwood (*Ostrya*  
154 *virginiana*).

155 The ecosystem-scale study used in this analysis was the UMBS Forest Accelerated  
156 Succession Experiment (FASET), which was designed to examine how the senescence of early  
157 successional tree species (aspen and birch) affects C cycling in a regionally representative forest  
158 ecosystem (Nave *et al.*, 2011). All aspen and birch trees within a 39 ha treatment area (>6700  
159 stems) were mechanically stem girdled in Spring 2008. These species comprised 39% of total  
160 pre-treatment basal area within the experimental treatment area. Peak disturbance (expressed as  
161 loss of leaf area) occurred in 2010, when 97% of all girdled trees within the treatment area had  
162 senesced and leaf area index (LAI) was 44% lower than pre-disturbance values (Gough *et al.*,  
163 2013). This experimental disturbance is similar in severity and extent to moderate natural  
164 disturbances related to wind or pest/pathogen outbreaks (Amiro *et al.*, 2010; Fahey *et al.*, 2015).  
165 The experimental area has canopy density of 700-800 stems per hectare, approximately 25 m<sup>2</sup>/ha  
166 basal area, and leaf area index of ~3.5 (Gough *et al.*, 2013)

167

## 168 2.2 Data collection and preparation

169 We assessed subcanopy response at three spatial scales, with the expectation that growth  
170 response to disturbance varies depending on mechanisms operating at neighborhood to landscape



171 scales: landscape (entire 39 ha manipulated area); plot (0.08 ha plots,  $n = 21$ ); and neighborhood  
172 (5 m radius subplots within a subset of 8 plots). Within the treatment area, 21 circular 0.08 ha  
173 plots (16 m radius) were established at 100 m intervals along seven transects radiating from the  
174 base of a meteorological tower (Gough *et al.*, 2013). All canopy trees ( $\geq 8$  cm DBH) in each plot  
175 were tagged and had DBH measured in 2006, 2010, and 2012, encompassing pre- (2006-2010)  
176 and post-peak (2010-2012) disturbance periods. For the subcanopy ( $< 8$  cm DBH), stems were  
177 tallied by species and diameter classes:  $< 2$  cm, 2-3.9 cm, 4-5.9 cm, and 6-7.9 cm in 2006, 2010,  
178 and 2012. Species and site or region-specific allometric equations were used to estimate  
179 aboveground biomass from DBH (Gough *et al.*, 2008) and biomass was converted to carbon  
180 based on a site specific conversion of 0.48 g of C per g of biomass (Gough *et al.*, 2013). For the  
181 subcanopy, the midpoint DBH for each diameter class (e.g., 3.0 cm for the 2 – 3.9 cm class) was  
182 used to estimate biomass for all stems. Mean annual aboveground wood net primary productivity  
183 ( $ANPP_w$ ) was calculated as the absolute change in total aboveground wood biomass between  
184 census years divided by the number of years in the interval, providing a production value leading  
185 up to (2006-2010) and following (2010-2012) peak disturbance (Stuart-Haëntjens *et al.*, 2015).

186 To assess neighborhood scale responses, four non-overlapping 5 m radius (0.008ha) subplots  
187 were established along each cardinal axis at randomly selected distances (but far enough from  
188 plot center to avoid overlap) within a subset of 8 plots (Table 1) that spanned a range of  
189 disturbance severities (13 to 67% canopy tree basal area loss). For all small subcanopy stems (0-  
190 3.9 cm DBH) in these subplots we measured height growth increment based on visual analysis of  
191 growth whorls and terminal bud scars. For all large subcanopy stems (4-7.9 cm DBH) we  
192 collected increment cores using an increment hammer to assess recent basal area (BA) growth.  
193 Increment cores were mounted on grooved wood blocks and sanded with progressively finer

194 sand paper to distinguish rings. Annual growth increments were measured to 0.001 mm using a  
195 Velmex stage micrometer and Metronics Quick-Check 4100. Current DBH and annual growth  
196 increments were used to calculate tree diameter in each year, which was then converted to basal  
197 area. Annual basal area increment (BAI) was then calculated based on change in basal area  
198 between years, and total BAI for pre- and post-peak disturbance periods was calculated based on  
199 these increments.

200 For each plot and subplot we quantified disturbance severity as the fraction of total pre-  
201 treatment BA (for stems  $\geq 8$  cm DBH) that senesced due to stem girdling (Stuart-Haëntjens *et al.*,  
202 2015). Due to the patchy distribution of aspen and birch across the site, disturbance severity  
203 varied greatly with the treatment area, with plot-level BA senescence (across all 21 plots)  
204 varying from 9 to 69%. We also used pre-disturbance canopy layer NPP<sub>w</sub> as a proxy for plot-  
205 scale variation in site fertility (hereafter “pre-disturbance canopy productivity”). Importantly,  
206 pre-disturbance canopy productivity and disturbance severity were not correlated ( $r = 0.21$ ,  $p =$   
207  $0.35$ ), indicating the lack of a pre-existing relationship between forest production and disturbance  
208 severity (Stuart-Haëntjens *et al.*, 2015). In addition, total subcanopy biomass prior to disturbance  
209 in 2006 varied greatly among plots (750 to 3267 kg C ha<sup>-1</sup>), but did not differ systematically with  
210 pre-disturbance canopy productivity ( $r = -0.07$ ) or disturbance severity ( $r = -0.30$ ).

211 Hemispherical canopy images and light data were collected in subplots at peak LAI in 2013  
212 to examine how disturbance affected canopy structure and light distribution through the canopy.  
213 Images were taken under diffuse sky conditions in each subplot at ground level, 1 m, and 5 m  
214 using a leveled camera with an 180° fisheye lens facing skyward. We derived estimates of  
215 canopy openness at each level using Gap Light Analyzer (GLA Version 2.0) software with the  
216 automatic optimal threshold algorithm applied (Nobis and Hunziker, 2005). We quantified the

217 fraction of absorbed photosynthetically active radiation (fAPAR) in each subplot as percent of  
218 above canopy PAR available at 5 m, 1m, and ground level. A BF2 sunshine sensor (Delta-T  
219 Devices, Cambridge, UK) on a meteorological tower measured total and diffuse PAR above the  
220 canopy. Separate PAR sensors (SQ-110 Quantum Sensor, Apogee Instruments, Logan, Utah  
221 USA) were stationed at ground level and 5 m above the forest floor in all subplots and connected  
222 to a central datalogger. To minimize the influence of sunflecks, we quantified fAPAR based over  
223 >4 hours when light was >85% diffuse (e.g., Tobin and Reich, 2009) and to reduce the influence  
224 of solar angle, the PAR time-series from each sensor was averaged. Because of poor diffuse  
225 conditions or sensor failure, we report results from 20 subplots.

226

### 227 *2.3 Data analysis*

228 Characteristics of the subcanopy layer (including biomass, stem density, and biomass  
229 increment) were compared between pre-disturbance and post-disturbance periods using two-  
230 tailed t-tests in Sigmaplot v. 13 (SYSTAT Software Inc. 2014). Correlations of these variables  
231 with disturbance severity, pre-disturbance primary production (~site fertility), and community  
232 composition (from ordination, see below) were tested using Pearson Product Moment  
233 Correlation statistics using Sigmaplot. Differences among years in biomass and relative  
234 dominance for subcanopy tree species were tested for using repeated measures Analysis of  
235 Variance (ANOVA) and individual comparisons were made using Holm-Sidak multiple  
236 comparison correction using Sigmaplot. Plot-level biomass and tree species distribution data for  
237 both canopy ( $\geq 8\text{cm DBH}$ ) and subcanopy (0-7.9cm DBH) were used to assess community  
238 composition and species diversity (richness, evenness, Shannon diversity index, and Beta  
239 diversity; McCune and Grace, 2002). Compositional patterns were extracted using nonmetric

240 multidimensional scaling (NMS) ordination on matrices of species biomass by plot for both the  
241 canopy and subcanopy layers. NMS was performed using PC-ORD v.5.31 (McCune and  
242 Mefford, 2006) with the “slow-and-thorough” auto-pilot setting, using 250 runs of real data and  
243 250 Monte Carlo randomizations to assess the robustness of the solution. Axes from the resulting  
244 ordination solutions were used as predictors representing the major compositional gradients in  
245 both layers. Variables relating to subcanopy production and response were overlaid on  
246 ordination graphs to visually assess their relationship with species composition.

247 To analyze compositional trajectories (Obj. 1), NMS ordination was performed (as above) on  
248 a combined matrix of relative biomass by species in the canopy and subcanopy layers in each  
249 plot. Compositional differences at the time of measurements and potential future community  
250 composition trajectories were illustrated using overlaid transitional vectors (McCune and  
251 Grace, 2002). To assess the effect of disturbance severity and pre-disturbance canopy  
252 productivity on variation in compositional differences between canopy layers, vectors were  
253 translated to the origin and the position of subcanopy points was regressed against these factors.  
254 Differences in subcanopy tree growth (height and BAI) and growth response among species  
255 (Obj. 2) were tested for with ANOVA and Kruskal-Wallis Rank ANOVA (where data were non-  
256 normal) using Sigmaplot.

257 To evaluate factors associated with subcanopy tree growth (Obj. 2), community-level  
258 productivity and resilience (Obj. 3) and the relative contribution of subcanopy to overall  
259 productivity response (Obj. 4) we used multiple linear regression in an information-theoretic  
260 framework (Burnham and Anderson, 2002) using PROC MIXED in SAS v. 9.2 (SAS-Institute-  
261 Inc., 2005). We tested a set of *a priori* models regarding the effect of a suite of stand and site  
262 characteristics (see below) on the following response variables: 1) post-disturbance subcanopy

263 productivity (2010-2012 subcanopy ANPP<sub>w</sub>), 2) absolute difference in pre- vs. post-disturbance  
264 subcanopy productivity (2010-2012 subcanopy ANPP<sub>w</sub> – 2006-2009 subcanopy ANPP<sub>w</sub>), 3)  
265 relative change in pre- vs. post disturbance subcanopy ANPP<sub>w</sub> ((2010-2012 subcanopy ANPP<sub>w</sub> –  
266 2006-2009 subcanopy ANPP<sub>w</sub>)/2006-2009 subcanopy ANPP<sub>w</sub>), 4) proportion of total (canopy +  
267 subcanopy) post-disturbance productivity contributed by subcanopy (2010-2012 subcanopy  
268 ANPP<sub>w</sub>/2010-2012 total ANPP<sub>w</sub>), 5) ratio of subcanopy response to total response ((2010-2012  
269 subcanopy ANPP<sub>w</sub> – 2006-2009 subcanopy ANPP<sub>w</sub>)/(2010-2012 total ANPP<sub>w</sub> – 2006-2009 total  
270 ANPP<sub>w</sub>)), 6) post-disturbance subcanopy growth of large subcanopy stems (4-7.9 cm DBH)  
271 (2010-2013 subcanopy BAI), and 7) post-disturbance growth response of large subcanopy stems  
272 (2010-2013 subcanopy BAI – 2006-2009 subcanopy BAI).

273 Potential predictors of plot-level subcanopy productivity and disturbance response (variables  
274 #1-5 above) included tree species composition (NMS ordination axes, percent late-successional  
275 hardwood or oak dominance) and diversity (Shannon index) in the canopy and subcanopy layers,  
276 as well as disturbance severity (based on percent of canopy basal area girdled) and site fertility  
277 (based on pre-disturbance canopy NPP). Predictors of subplot-level BA growth (variables #6 & 7  
278 above) included these factors and also pre-disturbance subcanopy basal area, canopy openness,  
279 and fAPAR at 5m. For both plot and subplot-level analyses a matrix plot was constructed with  
280 all predictors and response variables to assess linearity of relationships and check for co-linearity  
281 among predictors. For the subplot-level analysis plot identity was included as a random effect in  
282 the model to incorporate the nested design into the analysis. A set of plausible models (Appendix  
283 C) stemming from our understanding of factors likely to be related to subcanopy response  
284 (Gough *et al.*, 2013; Stuart-Haëntjens *et al.*, 2015) were evaluated using the corrected Akaike  
285 Information Criterion (AIC<sub>c</sub>; Burnham and Anderson, 2002). AIC<sub>c</sub> is derived from the maximum

286 log-likelihood estimate and number of parameters in a given model, penalizing models for lack  
287 of fit and the addition of parameters (lower values indicate better models). Models were ranked  
288 by the difference between the  $AIC_c$  value for the model and the lowest value in the full model set  
289 ( $\Delta AIC_c$ ), which included the null model. This method allows for comparison of the strength of  
290 evidence among the models, with increasing  $\Delta AIC_c$  values indicating decreasing probability of  
291 the fitted model being the best model in the set. Models with  $\Delta AIC_c < 2$  are considered to have  
292 substantial support and models above this threshold are generally not interpreted (Burnham and  
293 Anderson, 2002). To approximate the probability of a model being the best in a given set,  $\Delta AIC_c$   
294 values were used to calculate Akaike weights ( $w_i$ ).

295

### 296 **3. Results**

#### 297 *3.1 Landscape-level patterns in subcanopy layer characteristics – Objective 1*

298 There were major shifts in characteristics of the subcanopy layer, but landscape scale  
299 patterns in subcanopy stem density, biomass, and productivity ( $ANPP_w$ ) were not related to  
300 differences among plots in disturbance severity or site fertility (based on pre-disturbance canopy  
301 productivity). Subcanopy stem density five years following disturbance, in 2012, varied greatly  
302 among plots (1175 to 5788 stems  $ha^{-1}$ ) but was not correlated with pre-disturbance canopy  
303 productivity ( $r = 0.03$ ) or disturbance severity ( $r = -0.08$ ). Subcanopy biomass in 2012 was  
304 significantly greater than pre-disturbance levels (2542 vs 1800  $kg\ C\ ha^{-1}$ ;  $t = 2.56$ ,  $p = 0.014$ ;  
305 Table 1), but variation among plots was not explained by plot disturbance severity ( $r = -0.06$ ) or  
306 pre-disturbance canopy productivity ( $r = -0.21$ ). There was a relatively strong correlation  
307 between pre-disturbance and post-disturbance subcanopy biomass at the plot scale ( $r = 0.79$ ).  
308 However, total subcanopy productivity across the study period was not related to pre-disturbance

309 subcanopy biomass ( $r = -0.12$ ), indicating that greater response was not associated with greater  
310 initial biomass. Subcanopy productivity was significantly greater following peak disturbance  
311 ( $341 \text{ kg C ha}^{-1} \text{ yr}^{-1}$  in 2010-2012) relative to the period leading up to peak disturbance ( $15 \text{ kg C}$   
312  $\text{ha}^{-1} \text{ yr}^{-1}$  in 2006-2009;  $t = 7.48$ ,  $p < 0.001$ ; Table 1).

313 Patterns in species composition at the landscape scale (across all plots covering the large  
314 manipulated area), were similar before and after the disturbance with the same five species  
315 making up the majority of subcanopy biomass: beech, sugar maple, red maple, white pine, and  
316 northern red oak (Fig. 1). Although species biomass varied among years (species by year  
317 interaction -  $F_{24,480} = 5.28$ ,  $p < 0.001$ ) for a number of species in the subcanopy (Fig. 1), relative  
318 dominance only changed substantially following the disturbance for northern red oak (species by  
319 year interaction -  $F_{24,480} = 2.00$ ,  $p = 0.004$ ; Appendix A). The major shift was that northern red  
320 oak increased in biomass and dominance (from 9.5 to 17.6% relative dominance) while the other  
321 major species increased in biomass but declined in dominance, especially sugar maple (from  
322 21.1 to 15.1% relative dominance). Northern red oak had the greatest increase in landscape-scale  
323 subcanopy biomass across the study period, but was still only 4<sup>th</sup> most abundant overall in 2012.

324 Following the disturbance (in 2012) there was strong plot-level variation in subcanopy  
325 species composition, with groups characterized by northern red oak/serviceberry, sugar  
326 maple/ironwood, beech/striped maple, and pine dominance (Fig. 2a). NMS ordination of  
327 subcanopy composition resulted in a three-dimensional solution which was significant based on  
328 Monte Carlo test (Stress = 9.7,  $p = 0.036$ ). The 3D solution explained 85.7% of the variation in  
329 the original data matrix. Each of the three axes explained a large component of the variation:  
330 Axis 1 = 21.9%, Axis 2 = 40.9%, and Axis 3 = 22.9%. Subcanopy species composition in 2012  
331 was not strongly related to pre-disturbance canopy productivity (Axis 1 -  $r = -0.35$ , Axis 2 -  $r =$

332 0.27, Axis 3 –  $r = -0.15$ ) or disturbance severity (canopy BA loss; Axis 1 –  $r = 0.35$ , Axis 2 –  $r =$   
333  $-0.33$ , Axis 3 –  $r = -0.37$ ).

334 Differences in canopy and subcanopy composition following disturbance indicate a potential  
335 future shift from relatively homogenous canopy to the 3 or 4 much more differentiated species  
336 assemblages mentioned above (based on subcanopy dominance; Fig. 3, Fig. 2a). Species richness  
337 was higher in the subcanopy (mean = 6.6) than the canopy (4.4), but species diversity (Simpson's  
338 diversity index) was very similar between the two layers (0.54 and 0.55 respectively).

339 Composition was more differentiated among plots in the subcanopy than the canopy layer, with  
340 greater beta diversity ( $B_d$ ) evident in the subcanopy (1.6) versus the canopy (1.1). Combined  
341 NMS ordination of canopy and subcanopy composition (relative dominance) resulted in a three-  
342 dimensional solution which was highly significant based on Monte Carlo test (Stress = 10.7,  $p =$   
343 0.004). The 3D solution explained 89.7% of the variation in the original data matrix. The first  
344 two axes explained the majority of the variation – Axis 1 = 44.9% and Axis 2 = 29.2%. The  
345 direction and magnitude of canopy-subcanopy trajectories were somewhat related to disturbance  
346 severity (Axis 1 -  $r = 0.47$ ) and pre-disturbance canopy productivity (Axis 2 -  $r = 0.41$ ; Fig. 3).

347

### 348 *3.2 Species-level subcanopy growth patterns – Objective 2*

349 Height growth response of smaller subcanopy trees (<4cm dbh) varied greatly among species  
350 (based on Kruskal-Wallis rank ANOVA;  $H = 107.30$ ,  $df = 6$ ,  $p < 0.001$ ). Among smaller  
351 subcanopy trees, the species with by far the greatest average individual height growth response  
352 (post vs. pre-disturbance growth increment) was beech, which also ended the study period with  
353 the fastest average growth rate (Fig. 4a;  $H = 184.74$ ;  $df = 6$ ,  $p < 0.001$ ). Although beech was  
354 most responsive, red maple dominated basal area among these small subcanopy trees and also



355 had the greatest overall population-level height increment, despite a relatively minor response in  
356 total height increment following the disturbance (Fig. 4b).

357 Basal area growth response of larger subcanopy trees (4-8cm dbh) also varied greatly among  
358 species (based on Kruskal-Wallis rank ANOVA;  $H = 30.12$ ,  $df = 5$ ,  $p < 0.001$ ). For larger  
359 subcanopy stems northern red oak was dominant in both total population and average individual  
360 stem growth (Fig. 5). Northern red oak also had the largest response to disturbance in terms of  
361 average individual basal area increment (BAI) and especially total BAI across the entire  
362 population (Fig. 5). Prior to peak LAI loss in 2010, northern red oaks were already growing  
363 fastest on a per-BA basis (BA growth relative to total BA) among the canopy tree species in the  
364 system (Fig. 6; behind only ironwood – a subcanopy-only species which had very low total BA;  
365 ANOVA  $F_{7,137} = 2.34$ ,  $p = 0.03$ ). In the period following peak disturbance (2010-2013), northern  
366 red oak BAI per unit BA was much greater than that of any of the other canopy dominant species  
367 ( $F_{7,137} = 7.96$ ,  $p < 0.001$ ) and had also increased much more than any other species relative to  
368 pre-disturbance growth rates (Fig. 6;  $F_{7,137} = 5.21$ ,  $p < 0.001$ ).

369 Neighborhood scale (5m radius subplot) analysis of the predictors of BAI for large  
370 subcanopy stems ( $\geq 4$  cm DBH) indicated that species composition and structural/environmental  
371 conditions (i.e., canopy openness and light transmittance) combined to predict the response of  
372 the subcanopy to moderate severity disturbance. Specific response variables are defined in  
373 section 2.3 *Data analysis* above and referenced by corresponding numbers. Full lists of all  
374 potential candidate models are included in Appendix C. Post-disturbance growth rates (response  
375 variable #6) were most strongly predicted by subcanopy oak dominance and pre-disturbance  
376 subcanopy BA (Table 2). The model containing subcanopy oak dominance and pre-disturbance  
377 subcanopy BA as predictors was the most highly supported and accounted for a large proportion

378 of the weighting in the model set (70%). Growth of subcanopy trees was higher in locations with  
379 high oak dominance and high pre-disturbance subcanopy BA (Fig. 7). The post-disturbance  
380 response of subcanopy BA growth (response variable #7) at the neighborhood scale was most  
381 strongly predicted by subcanopy oak dominance and the fraction of above-canopy  
382 photosynthetically active radiation (fAPAR) available at 5m above the ground surface (Table 2).  
383 The model containing these two predictors was more highly supported than any other model in  
384 the set (all  $\Delta AICc > 2.0$ ) and accounted for 59% of the weighting in the model set. Growth  
385 response in subcanopy trees was highest with greater oak dominance and greater transmittance of  
386 PAR into the upper part of the subcanopy (Fig. 7).

387

### 388 *3.3 Community-level subcanopy biomass and productivity – Objective 3*

389 In general, subcanopy productivity and response to disturbance at the plot scale were  
390 predicted by subcanopy community composition and not disturbance severity. By far the  
391 strongest predictor of post-disturbance subcanopy productivity (response variable #1) was  
392 subcanopy community composition (Axis 1 from NMS ordination;  $r = 0.74$ ; Appendix B). Post-  
393 disturbance subcanopy productivity was not related to disturbance severity ( $r = 0.24$ ; Appendix  
394 B), but was somewhat negatively correlated with pre-disturbance canopy productivity ( $r = -0.57$ ;  
395 Appendix B), indicating that the subcanopy was less productive in highly productive sites,  
396 presumably because of greater competition from canopy trees (Fig.7). The most highly supported  
397 models from multiple regression of post-disturbance subcanopy productivity included  
398 combinations of pre-disturbance canopy productivity, subcanopy composition, and canopy oak  
399 dominance (Table 3, full model lists in Appendix C). The three most highly supported models  
400 that included combinations of these predictors accounted for 14-32% of the weighting in the full

401 set of models. Post-disturbance subcanopy productivity was highest in plots with northern red  
402 oak dominance in the subcanopy and in areas with lower pre-disturbance canopy productivity  
403 and lower canopy dominance by northern red oak (Fig. 7).

404 Absolute response in subcanopy productivity following disturbance (response variable #2)  
405 was most strongly predicted by a combination of subcanopy species composition (Axis 2 from  
406 NMS ordination;  $r = -0.75$ ; Appendix B), pre-disturbance canopy productivity ( $r = -0.56$ ), and  
407 canopy dominance by oaks ( $r = -0.59$ ) and late-successional hardwoods ( $r = 0.74$ ). The model  
408 containing these predictors was more highly supported than any other model in the set (all  $\Delta AICc$   
409  $> 2.0$ ) and accounted for 60% of the weighting in the model set (Table 3). Relative response in  
410 subcanopy productivity (response variable #3) was most strongly predicted by subcanopy  
411 composition (Axis 2 from NMS ordination;  $r = -0.51$ ) and late-successional hardwood  
412 dominance in the canopy layer ( $r = 0.42$ ). The model containing these predictors was more  
413 highly supported than any other model in the set (all  $\Delta AICc > 2.0$ ) and accounted for 47% of the  
414 weighting in the model set (Table 3). Both absolute and relative responsiveness were much lower  
415 in locations with subcanopy dominance by pines (Fig. 2b) and canopy dominance by oaks  
416 relative to late-successional hardwoods (Fig. 7).

417

#### 418 *3.4 Subcanopy contribution to overall productivity and disturbance response – Objective 4*

419 The subcanopy made a significant contribution to overall (subcanopy + canopy) wood  
420 production in the period following peak disturbance (from 2010-2012) averaging 16.2% across  
421 all plots within the moderately disturbed landscape. At the plot scale, subcanopy contribution to  
422 total wood production ranged between 5.3% in plot D3 and 28.8% in plot C2, which had low  
423 pre-disturbance canopy productivity and high disturbance severity (Table 1). Subcanopy

424 contribution to productivity (response variable #4) was most strongly predicted by a combination  
425 of subcanopy species composition (Axis 1 from NMS ordination;  $r = 0.68$ ; Appendix B), canopy  
426 species composition (Axis 2 from NMS ordination;  $r = 0.64$ ), pre-disturbance canopy  
427 productivity ( $r = -0.64$ ), and subcanopy species diversity ( $r = 0.14$ ). The model containing these  
428 predictors was more highly supported than any other model in the set (all  $\Delta AICc > 2.0$ ) and  
429 accounted for 67% of the weighting in the model set (Table 4). Subcanopy contribution to total  
430 productivity was greater in areas with low pre-disturbance canopy productivity where the canopy  
431 was dominated by sugar maple and basswood and the subcanopy by northern red oak.

432 The contribution of the subcanopy to overall productivity response (response variable #5)  
433 was substantial, averaging 54.1% across all plots. At the plot level, this contribution varied  
434 between 17% in plot F2 and 100% in the 6 plots which had a decline in canopy layer  
435 productivity (Table 1). Subcanopy contribution to plot scale growth response was most strongly  
436 predicted by pre-disturbance canopy productivity ( $r = -0.54$ ; Appendix B) and subcanopy  
437 composition (Axis 2 from NMS ordination;  $r = -0.52$ ). The models including combinations of  
438 these predictors were all more highly supported than other models in the set (all  $\Delta AICc > 2.0$ )  
439 and accounted for 18-28% of the weighting in the model set (Table 4). Subcanopy contribution  
440 to response was greater in areas with low pre-disturbance canopy productivity and where the  
441 subcanopy was pine dominated.

442

## 443 **4. Discussion**

### 444 *4.1 Subcanopy composition and ecosystem resilience*

445 The subcanopy played a central role in forest ecosystem wood productivity response to  
446 moderate severity disturbance, and this role was strongly mediated by subcanopy community

447 composition and the local light environment. Enhanced productivity in forest subcanopies has  
448 been shown to contribute significantly to overall ecosystem productivity response following  
449 disturbance (Romme *et al.*, 1986; Veblen *et al.*, 1989; Foster *et al.*, 1997; Dyer *et al.*, 2010). This  
450 was the case in the FASET disturbance as well, as subcanopy productivity was stimulated by the  
451 disturbance and associated redistribution of growth-limiting resources to undisturbed vegetation  
452 (Gough *et al.*, 2013; Stuart-Haëntjens *et al.*, 2015). In our analysis, the subcanopy accounted for  
453 greater than half of the total wood production response (change in productivity relative to pre-  
454 disturbance levels) following disturbance.

455       Although productivity response in the subcanopy was generally positive across the  
456 experimental disturbance (in 20 of 21 plots; Table 1), there was considerable variation among  
457 plots in both production and relative growth response following disturbance (approximately one  
458 order of magnitude in both cases) owing to subcanopy composition and environmental variation.  
459 The neighborhood-scale effects of moderate severity disturbance on resource distribution can be  
460 highly heterogeneous (Woods, 2004), and variation in the contribution of the subcanopy to  
461 ecosystem response has been associated with spatial variation in disturbance severity (Stuart-  
462 Haëntjens *et al.*, 2015). However, in this analysis subcanopy species composition outweighed  
463 disturbance severity and pre-disturbance canopy productivity (i.e., site fertility) as factors  
464 predicting responsiveness of growth to disturbance at the plot-scale. Community composition  
465 and diversity in vegetated ecosystems have previously been shown to be strong drivers of  
466 productivity (Chapin *et al.*, 1997; Tilman *et al.*, 2012) and response to disturbance (Spies and  
467 Franklin, 1989; Frelich and Reich, 1999), including at our study site (Gough *et al.*, 2010). Our  
468 results illustrate how fine-scale variation in community composition and resource availability

469 within a disturbed area can affect the response of forest ecosystems to disturbance, specifically  
470 through variable subcanopy response.

471 At a neighborhood scale, subcanopy light availability, which is associated with disturbance  
472 severity (Stuart-Haëntjens *et al.*, 2015), interacted strongly with species composition to drive  
473 growth response following moderate disturbance severity. Neighborhood disturbance severity,  
474 expressed as canopy basal area loss at the 5 m radius subplot-scale, was not strongly related to  
475 growth response in subcanopy trees. However, we found that variation in subcanopy light  
476 availability, specifically the proportion of photosynthetically active radiation that was  
477 transmitted (fAPAR) to the approximate top of the subcanopy layer (5 m height), was related to  
478 patterns in subcanopy tree growth at a neighborhood-scale. Subcanopy light availability is  
479 related to variation in canopy structure, which was driven partially by substantial fine-scale  
480 variation in disturbance severity (Stuart-Haëntjens *et al.*, 2015). The growth response of  
481 subcanopy trees was not mediated by light transmittance alone, though, as species composition  
482 was a strong and independent predictor of post-disturbance growth response. However, the  
483 response of species was not straightforwardly linked to perceived shade tolerance. Northern red  
484 oak, which is mid-tolerant of shade (Crow, 1988), was able to respond across a wide range of  
485 light environments and across the gradient in disturbance severity, not just in the most open sites.

486 The factors most strongly associated with spatial variation in subcanopy response to the  
487 landscape-scale moderate severity disturbance depended on the scale of the analysis and the  
488 proxies that were used to quantify disturbance severity. The ecological correlates of response to  
489 heterogeneous disturbance may vary across spatial scales (e.g., light transmittance vs. basal  
490 area), but some of our findings may also be explained by methodological factors associated with  
491 analysis scales. Plot-scale results suggested that the variable effects of moderate severity

492 disturbance on productivity were more strongly associated with pre-existing variation in  
493 subcanopy and canopy species assemblages and site fertility than the direct effects of differing  
494 disturbance severity (Frelich and Reich, 1999; Reyes and Kneeshaw, 2008). However, at the  
495 neighborhood scale the effect of disturbance-associated heterogeneity in resource conditions was  
496 apparent (in this case light transmittance, although other resources and edaphic factors such as N,  
497 water availability, and temperature may also be important; Nave *et al.*, 2011; Matheny *et al.*,  
498 2014). This may be a more appropriate scale at which to assess subcanopy tree growth response  
499 to such a fine-grained heterogeneous disturbance (Frelich and Reich, 1999). However, although  
500 light transmittance may have a stronger direct functional link to subcanopy tree growth than  
501 basal area loss, light transmittance was not directly correlated with disturbance severity  
502 (correlation between BA loss and fAPAR was  $r = 0.59$  at the subplot scale) and is less  
503 meaningful when averaged across subplots to the plot scale. Basal area loss can be a useful proxy  
504 for disturbance severity at the plot or stand scale (Stuart-Haëntjens *et al.*, 2015), but may not be  
505 as suitable at the neighborhood scale, due to the high weighting to individual trees in such a  
506 small area (0.008ha). The emphasis given to basal area loss and light transmittance at the plot  
507 and neighborhood scale reflects that they are likely the best proxies at these scales for matching  
508 data availability and ecological processes driving subcanopy growth/productivity response  
509 (Canham *et al.*, 2006).

510       The specifics of ecosystem response to moderate severity disturbance are likely to depend  
511 greatly on pre-disturbance forest structure and disturbance type (Turner *et al.*, 1997; Millward  
512 and Kraft, 2004; Cohen *et al.*, 2016). In forests with more homogenous species composition the  
513 relative effects of resource and environmental heterogeneity on disturbance growth response may  
514 be much more pronounced (Romme *et al.*, 1986; Veblen *et al.*, 1989). Similarly, in more patchy

515 moderate severity disturbances (e.g., with different spatial grain of disturbance patches) gradients  
516 in disturbance severity and therefore resource redistribution may be even more pronounced and  
517 may, thus, have a greater effect on community response (Woods, 2004; Reyes *et al.*, 2010; Fahey  
518 *et al.*, 2015). Our findings illustrate the need to understand the specific compositional and  
519 structural characteristics of forests to predict ecosystem response to disturbance (Hicke *et al.*,  
520 2012). However, our results also suggest that the response of forest ecosystems to moderate  
521 disturbance is likely to be an interactive relationship between site, community, and disturbance  
522 patterns, which may affect growth responses to disturbance differently at different spatial scales.

523

#### 524 *4.2 Subcanopy compositional trajectories*

525 Our results support the view that moderate severity disturbance, including those related to  
526 senescence of early-successional species, could play an important role in maintaining northern  
527 red oak in eastern deciduous forests (Hibbs, 1983; Crow, 1988). Patterns of species-level  
528 productivity and individual tree growth rates suggest greater future importance of northern red  
529 oak as a component of the FASET ecosystem following disturbance, matching results from  
530 studies on other natural and anthropogenic moderate severity disturbances (Oliver and Stephens,  
531 1977; Lorimer, 1983; Buckley *et al.*, 1998; Kaelke *et al.*, 2001; Parker and Dey, 2008). Northern  
532 red oak is generally described as mid-tolerant of shade and is associated with post-disturbance  
533 environments and edges (Crow, 1988). However, the trend toward oak dominance was not  
534 limited to areas with higher disturbance severity, as there was a very strong response from this  
535 species across the disturbance gradient. In this study, northern red oak appeared to be  
536 competitive with red maple in locations where the species co-occurred. This result may be  
537 considered somewhat surprising given the widespread and ongoing replacement of oaks by



538 maples and other mesophytic species across eastern North America (Lorimer, 1984; Nowacki *et*  
539 *al.*, 1990; Nowacki and Abrams, 2008), as well as previous work indicating that northern red oak  
540 is outcompeted by red maple in all but the most open conditions and even under oak canopy  
541 (Hibbs, 1982; Hartman *et al.*, 2005). However, the specific site conditions represented in the  
542 FASET experiment are also likely to have affected these patterns. For example, the UMBS area  
543 has sandy soils and relatively low N deposition and is fairly representative of the Upper Great  
544 Lakes region (Nave *et al.*, 2009). Previous studies have indicated that northern red oak can have  
545 a competitive advantage over shade tolerant species under such conditions (Kim *et al.*, 1996;  
546 Kaelke *et al.*, 2001). Ecosystems that do not have these characteristics may not be as suitable for  
547 northern red oak even with the influence of moderate severity disturbance (Crow, 1988).

548       The FASET disturbance appears to have created a patchy mosaic lacking widespread  
549 dominance by shade-tolerant, late-successional species. This finding matches predictions of  
550 theoretic and mechanistic models (Roxburgh *et al.*, 2004; Papaik and Canham, 2006) and  
551 patterns shown for other forests following moderate severity disturbance (Woods, 2004; Reyes *et*  
552 *al.*, 2010). Early results from this site do not indicate a trend toward accelerated dominance by  
553 shade-tolerant, late-successional species as has been illustrated elsewhere (Webb and Scanga,  
554 2001; Allen *et al.*, 2012). Overall, compositional trajectories suggest a potential “tri-furcation” of  
555 the community from uniform pre-disturbance aspen dominance to northern red oak, sugar maple,  
556 and beech dominated patches (Fig. 3). There was greater variation in species composition among  
557 plots (higher  $\beta$  diversity) in the subcanopy than the canopy layer, suggesting that the future  
558 canopy may be even more heterogeneous. However, within the disturbance area variation among  
559 plots in compositional trajectory was not related to variable disturbance severity but rather to  
560 differences in pre-disturbance canopy productivity. This finding suggests that compositional

561 transitions related to the FASET disturbance are being mediated by variation in site fertility,  
562 rather than directly by variable disturbance impacts (Reyes and Kneeshaw, 2008). The patterns  
563 presented here support the potential for moderate severity disturbance to maintain species  
564 diversity and create spatial heterogeneity in species composition (Hanson and Lorimer, 2007),  
565 but suggest that these effects may be limited to certain site conditions (e.g., heterogeneous sites,  
566 diverse communities) (Webb and Scanga, 2001; Reyes and Kneeshaw, 2008).

567

#### 568 *4.3 Management Implications and Conclusion*

569 Emulating moderate severity disturbances has been a major focus of ecological forestry in  
570 recent decades (Franklin *et al.*, 2002; Seymour and White, 2002; Puettmann *et al.*, 2009). The  
571 results of the FASET experiment provide some support for the utility of moderate severity  
572 disturbance in managing for biological diversity and structural heterogeneity, but also some  
573 insight on the potential limitations of such treatments. One primary goal of ecological forestry  
574 has been creation of structural complexity in forest stands and landscapes (Puettmann *et al.*,  
575 2009) and moderate severity disturbances have been shown to be highly effective at producing  
576 such conditions (Hanson and Lorimer, 2007; Fahey *et al.*, 2015). The structural complexity  
577 produced by the FASET disturbance has been strongly implicated in the resistance of the system  
578 to declines in productivity (Hardiman *et al.*, 2011; Gough *et al.*, 2013). The development of a  
579 robust subcanopy appears to be an important component of this current complexity and will also  
580 affect the future structure of the forest.

581 Another major goal of ecological forestry is to promote species and functional diversity,  
582 often through management focused on under-represented mid-tolerant species (Poznanovic *et al.*,  
583 2013). The FASET disturbance has promoted some species that are considered mid-tolerant and

584 under-represented at a regional scale (e.g., northern red oak). However, pines were not positively  
585 affected by the disturbance and community trajectories indicate a potential reduction in pine  
586 importance in the future at our site. Such a shift, if widespread, would represent a decrease in the  
587 primary conifer component in the system and the further reduction of an important component of  
588 functional diversity that has already declined at a regional scale following broad deforestation  
589 more than a century ago (Schulte *et al.*, 2007). The effects of moderate severity disturbances in  
590 promoting diversity are highly dependent on pre-existing conditions, as newly established  
591 seedlings are unlikely to be competitive with advance regeneration or sprouts (Webster and  
592 Lorimer, 2005; Dietze and Clark, 2008). Thus reintroduction or enrichment plantings focused on  
593 mid-tolerant species (possibly in combination with understory management) may be needed to  
594 take advantage of canopy disturbance in some cases (Dey *et al.*, 2012; Fahey and Lorimer,  
595 2013). Also, relying on moderate severity disturbance *per se* may not be sufficient to promote  
596 some species or functional types that require conditions associated with specific disturbance  
597 types. For example, to maintain a pine component, a combination of prescribed surface fire and  
598 moderate canopy disturbance may be needed (Frelich, 2002).

599 Our findings highlight the fundamental linkages between community composition and  
600 ecosystem structural and functional response to disturbance (Chapin *et al.*, 1997; Tilman *et al.*,  
601 2012). Efforts to predict the resilience of forest ecosystems to broad-scale, emerging threats such  
602 as global climate change and exotic pest outbreaks must account for the interaction of  
603 communities and ecosystem processes (Flower and Gonzalez-Meler, 2015; Thom and Seidl,  
604 2015). In addition to illustrating the potential response of forest ecosystems to canopy  
605 senescence related to successional transition, the results of this study could be applicable to  
606 understanding the response of forest ecosystems to ongoing exotic pest and pathogen outbreaks

607 (Flower *et al.*, 2013). Our work also illustrates the potential for diverse and heterogeneous forest  
608 communities to promote resilience to forest disturbance or successional change (Tilman *et al.*,  
609 2012). Understanding specific community responses to disturbance and landscape change will be  
610 essential to effectively modeling future forest landscapes and their ecological functioning  
611 (Duveneck *et al.*, 2014b).

612

### 613 **Acknowledgements**

614 Funding for the UMBS AmeriFlux core site was provided by the U.S. Department of Energy's  
615 (DOE) Office of Science and by the DOE Climate and Environmental Sciences Division, Office  
616 of Science (Award Nos. DE-SC0006708, DE-SC0006708) with additional support from the  
617 National Oceanic and Atmospheric Administration [NA11OAR4310190]. A. De La Cruz and E.  
618 Stockton were supported by the UMBS Summer Research Experience for Undergraduates (REU)  
619 program through NSF [AGS-0851421]. We acknowledge the University of Michigan Biological  
620 Station for facilities and material support and specifically J. Tallant for assistance with data  
621 management and spatial data layers.

622

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844

845 Table 1. Characteristics of sample plots within the FASET treatment area, including proportion  
 846 of basal area lost to girdling, which is used as a proxy for disturbance severity, and pre-  
 847 disturbance ANPP<sub>w</sub>, which was used as a proxy for site fertility.

Plot	Basal area		Subcanopy biomass (kg C ha <sup>-1</sup> )			Subcanopy ANPP <sub>w</sub> (kg C ha <sup>-1</sup> yr <sup>-1</sup> )				Prod. ratio <sup>^</sup>	Resp. ratio <sup>^</sup>
	loss <sup>#</sup>	Pre-disturbance canopy ANPP <sub>w</sub>	2006	2010	2012	Pre	Post	All	Response		
A1*	0.58	1369.9	1822.8	2196.4	3309.5	93.4	556.6	247.8	463.2	22.6	46.4
A2*	0.54	1286.1	1685.3	2116.3	3383.2	107.8	633.4	283.0	525.7	25.3	47.6
A3	0.37	1250.0	2430.9	1654.8	2746.2	-194.0	545.7	52.5	739.7	26.5	73.7
B1*	0.47	1722.6	1408.2	2559.5	3750.7	287.8	595.6	390.4	307.7	22.8	50.9
B2	0.38	1011.8	1745.3	1945.2	2998.4	50.0	526.6	208.8	476.6	27.7	56.9
B3	0.23	1766.9	3267.3	2499.9	3231.1	-191.8	365.6	-6.0	557.4	10.7	30.3
C1*	0.64	2546.5	2416.9	2758.5	3663.5	85.4	452.5	207.8	367.1	21.2	100
C2	0.64	852.0	906.5	1195.4	2054.5	72.2	429.5	191.3	357.3	28.8	62.9
C3	0.30	1486.4	3198.3	2631.3	3295.3	-141.8	332.0	16.2	473.7	15.2	56.5
D1*	0.38	1900.8	863.2	992.7	1666.6	32.4	336.9	133.9	304.6	17.5	100
D2*	0.67	1634.3	750.2	719.7	1200.0	-7.6	240.1	75.0	247.7	18.7	100
D3	0.45	2099.0	844.9	918.8	1129.2	18.5	105.2	47.4	86.7	5.3	100
E1	0.49	1790.0	3242.3	3352.9	4008.5	27.6	327.8	127.7	300.1	15.4	97.3
E2	0.10	1411.9	3223.0	3062.8	3737.5	-40.0	337.3	85.7	377.4	17.0	62.2
E3	0.22	2394.7	2232.6	2963.0	3277.1	182.6	157.0	174.1	-25.6	7.3	6.2
F1*	0.13	1627.3	1405.0	1144.6	1720.2	-65.1	287.8	52.5	352.9	13.7	66.1
F2	0.09	1702.7	1126.2	1085.7	1377.0	-10.1	145.6	41.8	155.7	5.6	17.3
F3	0.12	1729.5	1798.8	2075.0	2520.6	69.0	222.8	120.3	153.8	8.9	22.1
G1	0.69	3214.1	1338.4	1075.3	1335.2	-65.8	129.9	-0.5	195.7	8.3	100
G2	0.19	1431.9	1289.1	1197.2	1754.8	-23.0	278.8	77.6	301.8	14.5	59.2
G3	0.53	2180.3	798.5	928.9	1224.4	32.6	147.8	71.0	115.2	6.9	100

848 <sup>#</sup> Proportion of total pre-disturbance basal area that was removed by stem girdling treatment

- 849 \* Plots where measurements of subcanopy tree height and diameter growth were made on nested subplots.
- 850 ^ Prod. ratio = post-disturbance subcanopy ANPP<sub>w</sub>/total post-disturbance ANPP<sub>w</sub>, Resp. ratio = post-disturbance
- 851 subcanopy ANPP<sub>w</sub> response/total post-disturbance ANPP<sub>w</sub> response.



852 Table 2. Model lists for multiple regression modeling of basal area growth of large subcanopy  
 853 stems at the subplot scale.

Model*	k	AIC <sub>c</sub>	ΔAIC <sub>c</sub>	w <sub>i</sub>
<i>Post-peak BAI</i>				
SCPreBA SCOakDom <sup>%</sup>	4	-271.8	0.0	0.70
SCPreBA SCNMS1	4	-269.4	2.4	0.21
SCPreBA SCNMS1 OpennessGround	5	-267.6	4.2	0.09
SCOakDom	3	-261.7	10.1	0.00
SCOakDom OpennessGround	5	-260.2	11.6	0.00
<i>BAI Response</i>				
SCOakDom FracfAPAR5m <sup>&amp;</sup>	4	12.1	0.0	0.59
SCNMS1 FracfAPAR5m	4	14.1	2.0	0.22
SPDistBA SCOakDom FracfAPAR5m	5	15.8	3.7	0.09
SPDistBA SCNMS1 FracfAPAR5m	5	17.8	5.7	0.03
SPDistBA FracfAPAR5m	4	18.3	6.2	0.03

854 \* SCNMS1 = axis from ordination of subcanopy composition at the subplot-scale,

855 SCOakDom = relative basal area of northern red oak in subcanopy layer,

856 FracfAPAR5m = fraction of above canopy photosynthetically active radiation at 5m

857 above ground, SPDistBA = basal area removed by experimental disturbance at the

858 subplot level, OpennessGround = percent canopy openness at ground level,

859 SCPreBA = pre-disturbance subcanopy basal area at the subplot scale.

860 <sup>%</sup> Model parameters and standard errors: Intercept -0.0008 (0.0005), SCPreBA 0.23

861 (0.03), SCOakDom 0.00004 (0.000001).

862 <sup>&</sup> Model parameters and standard errors: Intercept -0.0008 (0.0005), SCOakDom 0.008

863 (0.002), FracfAPAR5m 1.15 (0.48).

864

865

866 Table 3. Model lists for multiple regression modeling of subcanopy productivity and productivity  
 867 response at the plot scale.

Model*	k	AIC <sub>c</sub>	ΔAIC <sub>c</sub>	w <sub>i</sub>
<i>Post-peak Productivity</i>				
CanPreNPP SubCanNMS1 CanOakDom <sup>%</sup>	5	185.1	0.0	0.32
CanPreNPP SubCanNMS1	4	185.5	0.4	0.26
SubCanNMS1 CanOakDom	4	186.7	1.6	0.14
SubCanNMS1	3	188.2	3.1	0.07
CanPreNPP SubCanNMS1 CanOakDom CanNMS2	6	188.6	3.5	0.05
<i>Absolute change in productivity</i>				
CanPreNPP SubCanNMS2 CanOakDom CanLSHDom <sup>&amp;</sup>	6	180.5	0.0	0.60
SubCanNMS2 CanOakDom CanLSHDom	5	183.5	3.0	0.13
CanPreNPP CanOakDom CanLSHDom	5	184.6	4.1	0.08
CanPreNPP SubCanNMS2 CanOakDom CanLSHDom SubCanD	7	184.8	4.3	0.07
CanPreNPP SubCanNMS2	4	186.5	6.0	0.03
<i>Relative change in productivity</i>				
SubCanNMS2 CanD <sup>§</sup>	4	14.1	0.0	0.47
SubCanNMS2 CanNMS2 CanD	5	17.5	3.4	0.09
SubCanNMS2 CanD CanLSHDom	4	17.6	3.5	0.08
CanNMS2 CanD	4	17.9	3.8	0.07
SubCanNMS2	3	18.3	4.2	0.06

868 \* CanPreNPP = pre-disturbance ANPP<sub>w</sub> for canopy, SubCanNMS1 & 2 = axes from  
 869 ordination of subcanopy composition at the plot-scale, CanNMS1 & 2 = axes from  
 870 ordination of canopy composition at the plot-scale, CanOakDom & CanLSHDom =  
 871 relative basal area of northern red oak and late-successional hardwoods (sugar  
 872 maple, basswood, beech) in canopy layer, CanD = canopy species diversity  
 873 (Simpson's diversity index).

874 <sup>%</sup> Model parameters and standard errors: Intercept 94.97 (13.21), CanPreNPP -0.015  
875 (0.007), SubCanNMS1 22.39 (7.04), CanOakDom -31.23 (16.88).

876 <sup>&</sup> Model parameters and standard errors: Intercept 88.09 (11.22), CanPreNPP -0.015  
877 (0.006), SubCanNMS2 -17.99 (6.52), CanOakDom -34.68 (13.14), CanLSHDom  
878 47.00 (18.89)

879 <sup>§</sup> Model parameters and standard errors: Intercept 0.44 (0.20), SubCanNMS2 -0.31  
880 (0.10), CanD 0.93 (0.34).

881

882 Table 4. Model lists for multiple regression modeling of subcanopy contribution to total  
 883 productivity and response of productivity to disturbance.

Model*	k	AIC <sub>c</sub>	ΔAIC <sub>c</sub>	w <sub>i</sub>
<i>Subcanopy contribution to overall productivity</i>				
CanPreNPP SubCanNMS1 CanNMS2 SubCanD <sup>%</sup>	6	58.4	0.0	0.67
CanPreNPP SubCanNMS1 CanOakDom CanNMS2 SubCanD	7	60.5	2.1	0.23
CanPreNPP CanOakDom CanNMS2 SubCanD	6	64.0	5.6	0.04
CanPreNPP SubCanNMS1 CanNMS2	5	65.0	6.6	0.02
CanPreNPP CanNMS2 SubCanD	5	66.9	8.5	0.01
<i>Subcanopy contribution to overall response</i>				
CanPreNPP	3	191.9	0.0	0.28
SubCanNMS2	3	192.1	0.2	0.25
CanPreNPP SubCanNMS2 <sup>&amp;</sup>	4	192.8	0.9	0.18
SubCanNMS2 CanNMS2	4	194.4	2.5	0.08
CanNMS2	3	194.5	2.6	0.08

884 \* CanPreNPP = pre-disturbance ANPP<sub>w</sub> for canopy, SCNMS1 & 2 = axes from  
 885 ordination of subcanopy composition at the plot-scale, CanNMS2 = axis from  
 886 ordination of canopy composition at the plot-scale, CanOakDom = relative basal area  
 887 of northern red oak in canopy layer, SubCanD – species diversity (Simpson’s  
 888 diversity index) of subcanopy layer.

889 <sup>%</sup> Model parameters and standard errors: Intercept 3.24 (0.78), CanPreNPP -0.001  
 890 (0.0002), SubCanNMS1 0.85 (0.27), CanNMS2 1.12 (0.22), SubCanD 4.35 (1.08).

891 <sup>&</sup> Model parameters and standard errors: Intercept 33.4 (14.44), CanPreNPP -0.013  
 892 (0.008), SubCanNMS2 -10.73 (7.04).

893  
 894  
 895

896 Figure Captions

897

898 Figure 1. Mean plot-level subcanopy biomass by species in 2006, 2010, and 2012 – error bars  
899 indicate standard error. Letters indicate significant differences among years for species based on  
900 repeated measures ANOVA ( $F_{24,480} = 5.283$ ,  $p < 0.001$ ). Species acronyms as follows: ACPE –  
901 *Acer pensylvanicum*, ACRU - *Acer rubrum*, ACSA - *Acer saccharum*, AMEL - *Amelanchier*  
902 *spp.*, BEPA - *Betula papyrifera*, FAGR - *Fagus grandifolia*, OSVI - *Ostrya virginiana*, PIRE –  
903 *Pinus resinosa*, PIST - *Pinus strobus*, POGR - *Populus grandidentata*, POTR - *Populus*  
904 *tremuloides*, QURU - *Quercus rubra*.

905

906 Figure 2. Non-metric multidimensional scaling ordination of subcanopy composition with (a)  
907 species centroids and (b) plot symbols scaled by post-disturbance subcanopy productivity  
908 (ANPP<sub>w</sub>). Species acronyms can be found in Fig. 1 caption.

909

910 Figure 3. Non-metric multidimensional scaling ordination for combined canopy and subcanopy  
911 in 2012 showing difference in composition between layers in each plot. Species acronyms can be  
912 found in Fig. 1 caption.

913

914 Figure 4. Height increment trend by species for small saplings (< 4cm dbh) showing (a) mean  
915 individual height increment (error bars indicate standard error) and (b) total height increment  
916 summed across all individuals. Species acronyms can be found in Fig. 1 caption.

917

918 Figure 5. Basal area increment trends over study period by species for large saplings (> 4cm dbh)  
919 showing (a) mean individual basal area increment (error bars indicate standard error) and (b)  
920 total basal area increment summed across all individuals. Species acronyms can be found in Fig.  
921 1 caption.

922

923 Figure 6. Relative growth (total basal area increment divided by total basal area) of subcanopy  
924 tree species before and after peak disturbance in 2010. Error bars represent standard error.  
925 Species acronyms can be found in Fig. 1 caption.

926

927 Figure 7. Relationships of subcanopy (a) productivity (post-disturbance ANPP<sub>w</sub> in kg C ha<sup>-1</sup> yr),  
928 (b) productivity response ((post-disturbance ANPP<sub>w</sub> – pre-disturbance ANPP<sub>w</sub>)/pre-disturbance  
929 ANPP<sub>w</sub>), (c) post-disturbance basal area increment (in m<sup>2</sup> ha<sup>-1</sup> yr), and (d) response in basal area  
930 increment ((post-disturbance BAI – pre-disturbance BAI)/pre-disturbance BAI) with strongest  
931 predictors from multiple regression analysis.

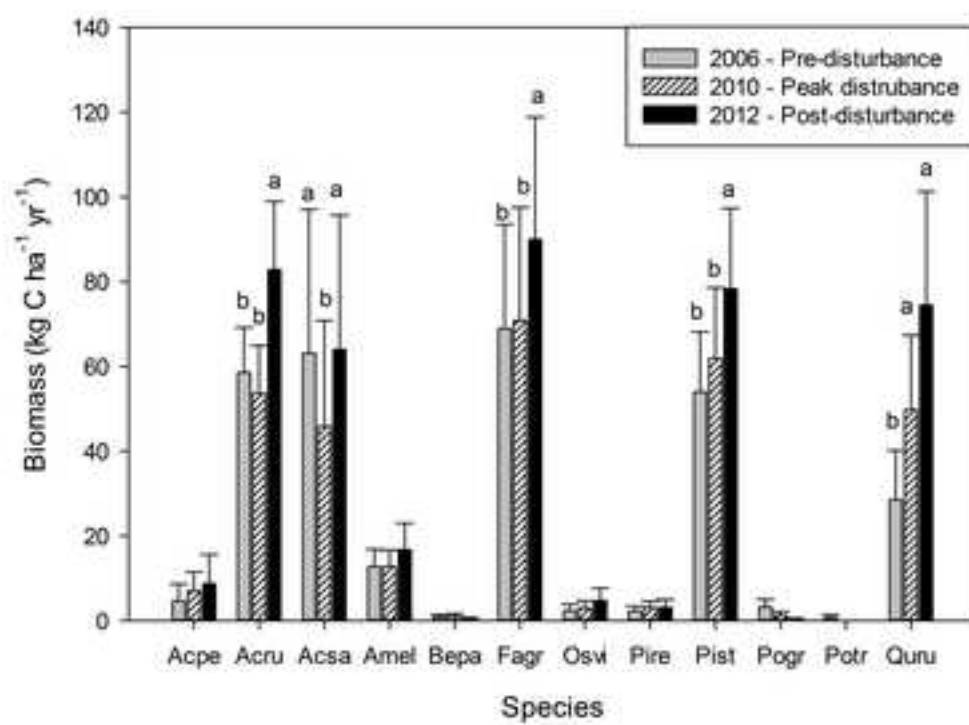
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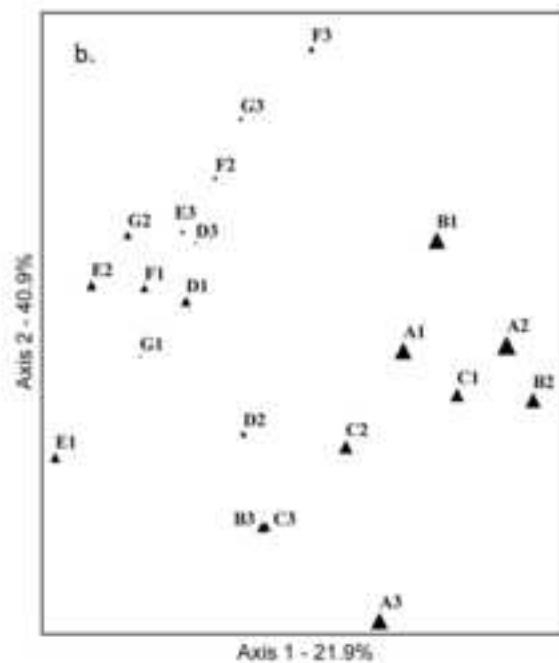
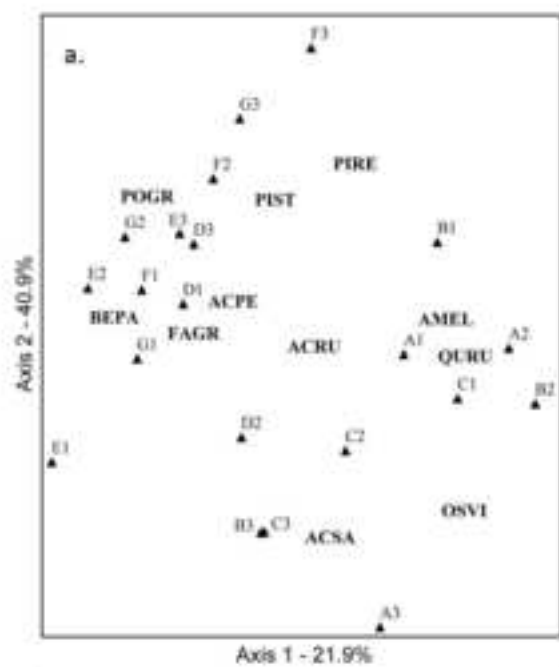
Highlights:

- Subcanopy contributed substantially to forest growth response to moderate disturbance
- Variable subcanopy response was mediated by composition not disturbance severity
- Moderate severity disturbance promoted growth of mid-tolerant northern red oak

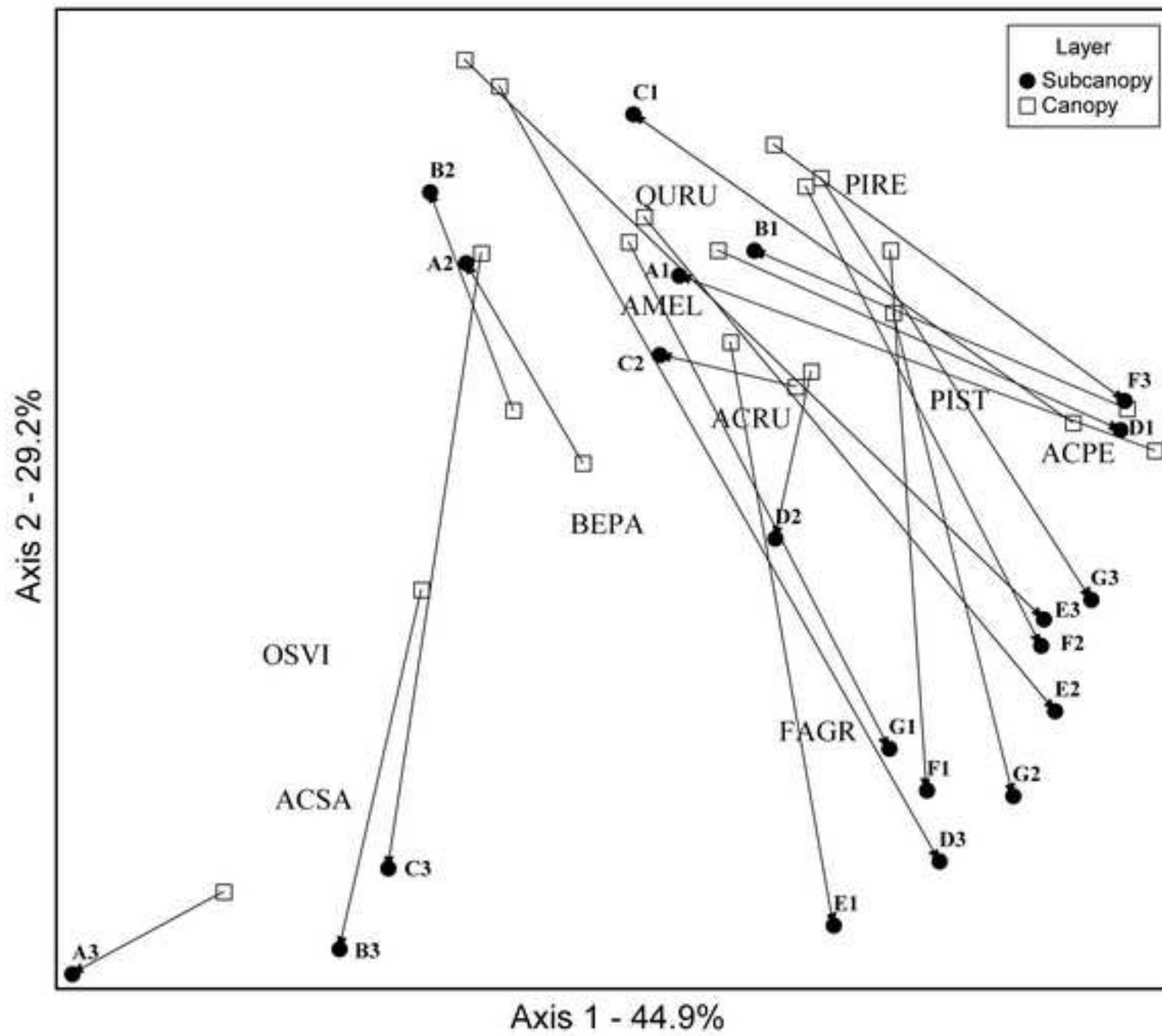
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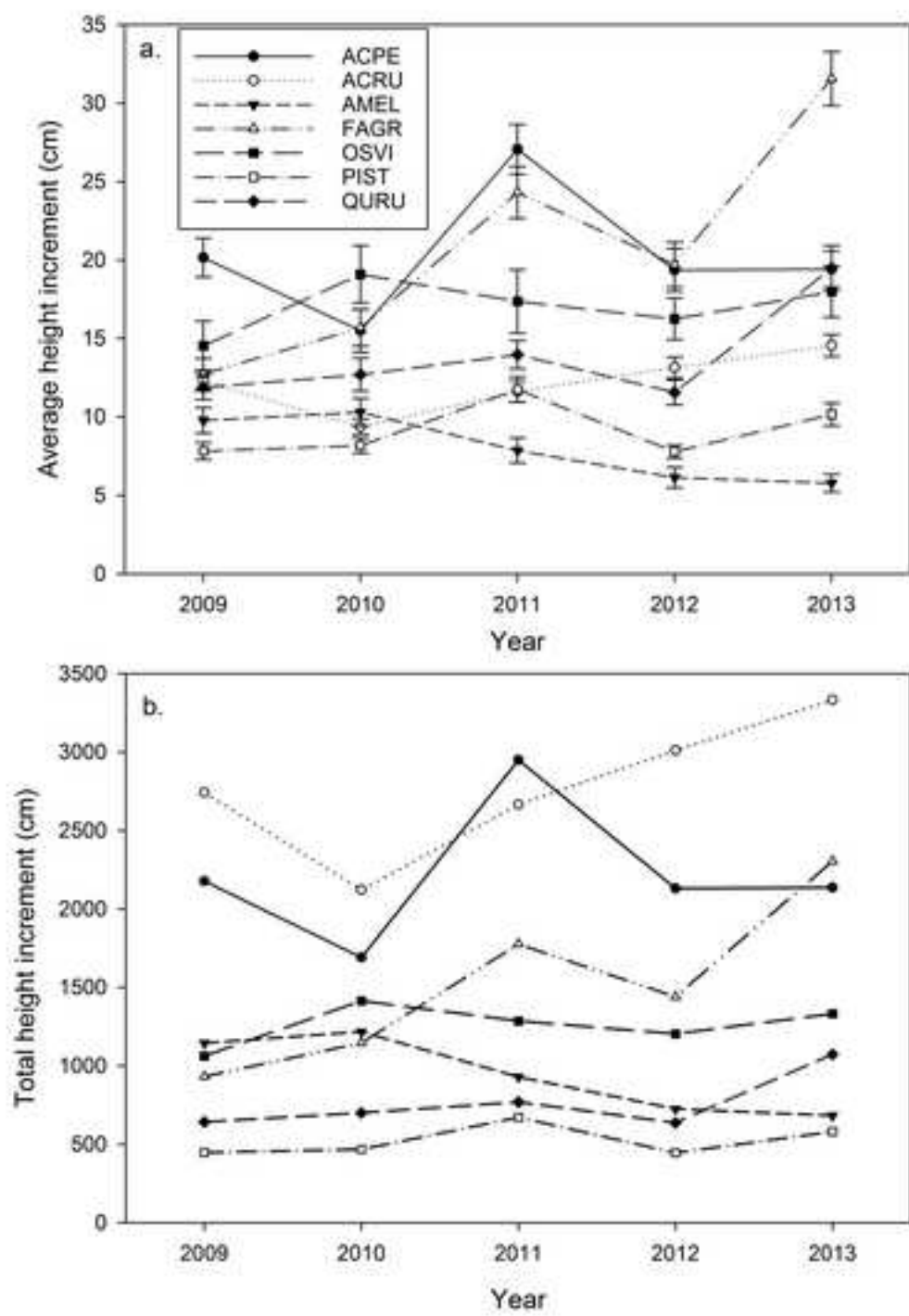




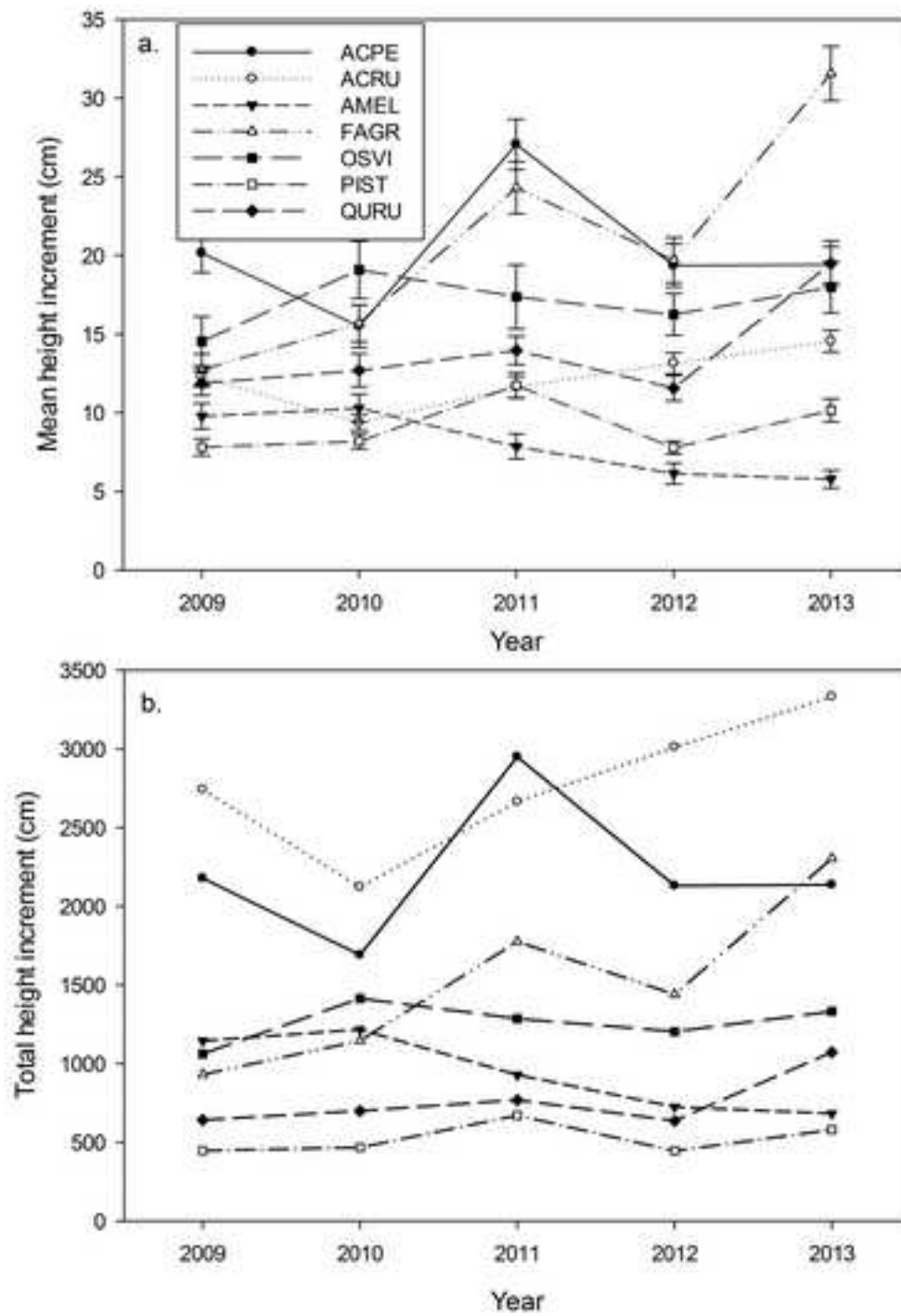
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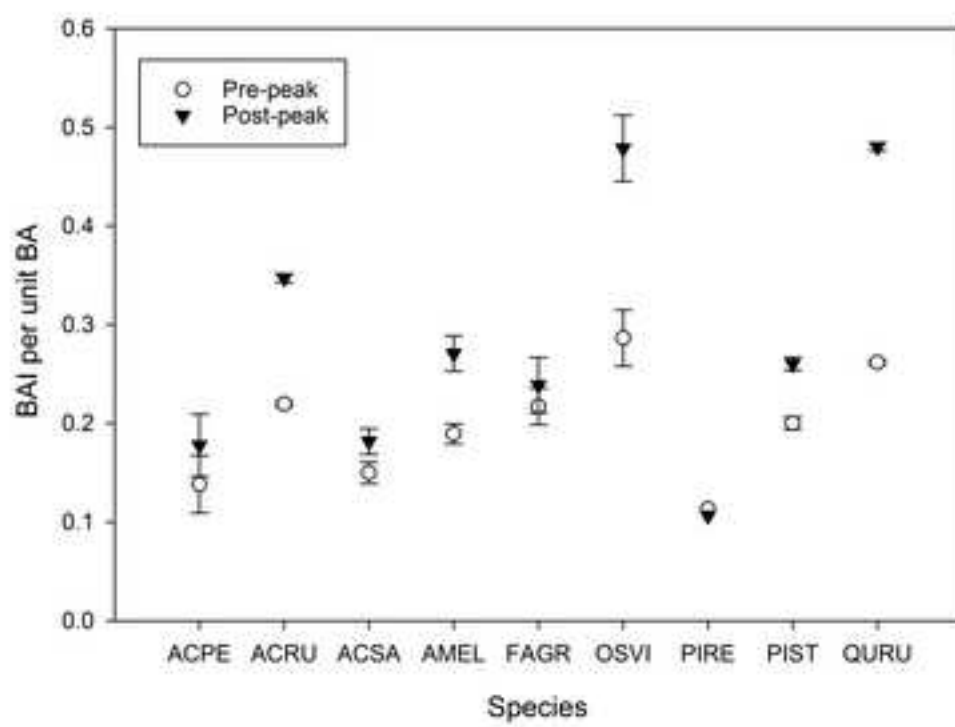
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